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# Convex Analysis and Monotone Operator Theory in Hilbert Spaces

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# Convex Analysis and Monotone Operator Theory in Hilbert Spaces

 Springer

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*Für Steffi, Andrea & Kati*

*À ma famille*



# Foreword

This self-contained book offers a modern unifying presentation of three basic areas of nonlinear analysis, namely convex analysis, monotone operator theory, and the fixed point theory of nonexpansive mappings.

This turns out to be a judicious choice. Showing the rich connections and interplay between these topics gives a strong coherence to the book. Moreover, these particular topics are at the core of modern optimization and its applications.

Choosing to work in Hilbert spaces offers a wide range of applications, while keeping the mathematics accessible to a large audience. Each topic is developed in a self-contained fashion, and the presentation often draws on recent advances.

The organization of the book makes it accessible to a large audience. Each chapter is illustrated by several exercises, which makes the monograph an excellent textbook. In addition, it offers deep insights into algorithmic aspects of optimization, especially splitting algorithms, which are important in theory and applications.

Let us point out the high quality of the writing and presentation. The authors combine an uncompromising demand for rigorous mathematical statements and a deep concern for applications, which makes this book remarkably accomplished.

Montpellier (France), October 2010

*Hédy Attouch*





# Preface

Three important areas of nonlinear analysis emerged in the early 1960s: convex analysis, monotone operator theory, and the theory of nonexpansive mappings. Over the past four decades, these areas have reached a high level of maturity, and an increasing number of connections have been identified between them. At the same time, they have found applications in a wide array of disciplines, including mechanics, economics, partial differential equations, information theory, approximation theory, signal and image processing, game theory, optimal transport theory, probability and statistics, and machine learning.

The purpose of this book is to present a largely self-contained account of the main results of convex analysis, monotone operator theory, and the theory of nonexpansive operators in the context of Hilbert spaces. Authoritative monographs are already available on each of these topics individually. A novelty of this book, and indeed, its central theme, is the tight interplay among the key notions of convexity, monotonicity, and nonexpansiveness. We aim at making the presentation accessible to a broad audience, and to reach out in particular to the applied sciences and engineering communities, where these tools have become indispensable. We chose to cast our exposition in the Hilbert space setting. This allows us to cover many applications of interest to practitioners in infinite-dimensional spaces and yet to avoid the technical difficulties pertaining to general Banach space theory that would exclude a large portion of our intended audience. We have also made an attempt to draw on recent developments and modern tools to simplify the proofs of key results, exploiting for instance heavily the concept of a Fitzpatrick function in our exposition of monotone operators, the notion of Fejér monotonicity to unify the convergence proofs of several algorithms, and that of a proximity operator throughout the second half of the book.

The book is organized in 29 chapters. Chapters 1 and 2 provide background material. Chapters 3 to 7 cover set convexity and nonexpansive operators. Various aspects of the theory of convex functions are discussed in Chapters 8 to 19. Chapters 20 to 25 are dedicated to monotone operator the-

ory. In addition to these basic building blocks, we also address certain themes from different angles in several places. Thus, optimization theory is discussed in Chapters 11, 19, 26, and 27. Best approximation problems are discussed in Chapters 3, 19, 27, 28, and 29. Algorithms are also present in various parts of the book: fixed point and convex feasibility algorithms in Chapter 5, proximal-point algorithms in Chapter 23, monotone operator splitting algorithms in Chapter 25, optimization algorithms in Chapter 27, and best approximation algorithms in Chapters 27 and 29. More than 400 exercises are distributed throughout the book, at the end of each chapter.

Preliminary drafts of this book have been used in courses in our institutions and we have benefited from the input of postdoctoral fellows and many students. To all of them, many thanks. In particular, HHB thanks Liangjin Yao for his helpful comments. We are grateful to Hédÿ Attouch, Jon Borwein, Stephen Simons, Jon Vanderwerff, Shawn Wang, and Isao Yamada for helpful discussions and pertinent comments. PLC also thanks Oscar Wesler. Finally, we thank the Natural Sciences and Engineering Research Council of Canada, the Canada Research Chair Program, and France's Agence Nationale de la Recherche for their support.

Kelowna (Canada)  
Paris (France)  
October 2010

*Heinz H. Bauschke*  
*Patrick L. Combettes*

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# Chapter 1

## Background

This chapter reviews basic definitions, facts, and notation from set-valued analysis, topology, and metric spaces that will be used throughout the book.

### 1.1 Sets in Vector Spaces

Let  $\mathcal{X}$  be a real vector space, let  $C$  and  $D$  be subsets of  $\mathcal{X}$ , and let  $z \in \mathcal{X}$ . Then  $C + D = \{x + y \mid x \in C, y \in D\}$ ,  $C - D = \{x - y \mid x \in C, y \in D\}$ ,  $z + C = \{z\} + C$ ,  $C - z = C - \{z\}$ , and, for every  $\lambda \in \mathbb{R}$ ,  $\lambda C = \{\lambda x \mid x \in C\}$ . If  $A$  is a nonempty subset of  $\mathbb{R}$ , then  $\lambda C = \bigcup_{\lambda \in A} \lambda C$  and  $Az = A\{z\} = \{\lambda z \mid \lambda \in A\}$ . In particular,  $C$  is a *cone* if

$$C = \mathbb{R}_{++}C, \quad (1.1)$$

where  $\mathbb{R}_{++} = \{\lambda \in \mathbb{R} \mid \lambda > 0\}$ . Moreover,  $C$  is an *affine subspace* if

$$C \neq \emptyset \quad \text{and} \quad (\forall \lambda \in \mathbb{R}) \quad C = \lambda C + (1 - \lambda)C. \quad (1.2)$$

Suppose that  $C \neq \emptyset$ . The intersection of all the linear subspaces of  $\mathcal{X}$  containing  $C$ , i.e., the smallest linear subspace of  $\mathcal{X}$  containing  $C$ , is denoted by  $\text{span } C$ ; its closure is the smallest closed linear subspace of  $\mathcal{X}$  containing  $C$  and it is denoted by  $\overline{\text{span}} C$ . Likewise, the intersection of all the affine subspaces of  $\mathcal{X}$  containing  $C$ , i.e., the smallest affine subspace of  $\mathcal{X}$  containing  $C$ , is denoted by  $\text{aff } C$  and called the *affine hull* of  $C$ . If  $C$  is an affine subspace, then  $V = C - C$  is the *linear subspace parallel to  $C$*  and  $(\forall x \in C) \quad C = x + V$ .

The four types of *line segments* between two points  $x$  and  $y$  in  $\mathcal{X}$  are

$$[x, y] = \{(1 - \alpha)x + \alpha y \mid 0 \leq \alpha \leq 1\}, \quad (1.3)$$

$]x, y[ = \{(1 - \alpha)x + \alpha y \mid 0 < \alpha < 1\}$ ,  $[x, y[ = \{(1 - \alpha)x + \alpha y \mid 0 \leq \alpha < 1\}$ , and  $]x, y] = [y, x[$ .

## 1.2 Operators

Let  $\mathcal{X}$ ,  $\mathcal{Y}$ , and  $\mathcal{Z}$  be nonempty sets, and let  $2^{\mathcal{Y}}$  be the *power set* of  $\mathcal{Y}$ , i.e., the family of all subsets of  $\mathcal{Y}$ . The notation  $T: \mathcal{X} \rightarrow \mathcal{Y}$  means that the operator (also called mapping)  $T$  maps every point  $x$  in  $\mathcal{X}$  to a point  $Tx$  in  $\mathcal{Y}$ . Thus, the notation  $A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$  means that  $A$  is a *set-valued operator* from  $\mathcal{X}$  to  $\mathcal{Y}$ , i.e.,  $A$  maps every point  $x \in \mathcal{X}$  to a set  $Ax \subset \mathcal{Y}$ . Let  $A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$ . Then  $A$  is characterized by its *graph*

$$\text{gra } A = \{(x, u) \in \mathcal{X} \times \mathcal{Y} \mid u \in Ax\}. \quad (1.4)$$

If  $C$  is a subset of  $\mathcal{X}$ , then  $A(C) = \bigcup_{x \in C} Ax$ . Given  $B: \mathcal{Y} \rightarrow 2^{\mathcal{Z}}$ , the *composition*  $B \circ A$  is

$$B \circ A: \mathcal{X} \rightarrow 2^{\mathcal{Z}}: x \mapsto B(Ax) = \bigcup_{y \in Ax} By. \quad (1.5)$$

The *domain* and the *range* of  $A$  are

$$\text{dom } A = \{x \in \mathcal{X} \mid Ax \neq \emptyset\} \quad \text{and} \quad \text{ran } A = A(\mathcal{X}), \quad (1.6)$$

respectively. If  $\mathcal{X}$  is a topological space, the closure of  $\text{dom } A$  is denoted by  $\overline{\text{dom } A}$ ; likewise, if  $\mathcal{Y}$  is a topological space, the closure of  $\text{ran } A$  is denoted by  $\overline{\text{ran } A}$ . The *inverse* of  $A$ , denoted by  $A^{-1}$ , is defined through its graph

$$\text{gra } A^{-1} = \{(u, x) \in \mathcal{Y} \times \mathcal{X} \mid (x, u) \in \text{gra } A\}. \quad (1.7)$$

Thus, for every  $(x, u) \in \mathcal{X} \times \mathcal{Y}$ ,  $u \in Ax \Leftrightarrow x \in A^{-1}u$ . Moreover,  $\text{dom } A^{-1} = \text{ran } A$  and  $\text{ran } A^{-1} = \text{dom } A$ . If  $\mathcal{Y}$  is a vector space, the set of zeros of  $A$  is

$$\text{zer } A = A^{-1}0 = \{x \in \mathcal{X} \mid 0 \in Ax\}. \quad (1.8)$$

When, for every  $x \in \text{dom } A$ ,  $Ax$  is a singleton, say  $Ax = \{Tx\}$ , then  $A$  is said to be *at most single-valued* from  $\mathcal{X}$  to  $\mathcal{Y}$  and it can be identified with an operator  $T: \text{dom } A \rightarrow \mathcal{Y}$ . Conversely, if  $D \subset \mathcal{X}$ , an operator  $T: D \rightarrow \mathcal{Y}$  can be identified with an at most single-valued operator from  $\mathcal{X}$  to  $\mathcal{Y}$ , namely

$$A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}: Ax = \begin{cases} \{Tx\}, & \text{if } x \in D; \\ \emptyset, & \text{otherwise.} \end{cases} \quad (1.9)$$

A *selection* of a set-valued operator  $A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$  is an operator  $T: \text{dom } A \rightarrow \mathcal{Y}$  such that  $(\forall x \in \text{dom } A) Tx \in Ax$ . Now let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ , let  $C \subset \mathcal{X}$ , and let  $D \subset \mathcal{Y}$ . Then  $T(C) = \{Tx \mid x \in C\}$  and  $T^{-1}(D) = \{x \in \mathcal{X} \mid Tx \in D\}$ .

Suppose that  $\mathcal{Y}$  is a real vector space, let  $A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$ , let  $B: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$ , and let  $\lambda \in \mathbb{R}$ . Then

$$A + \lambda B: \mathcal{X} \rightarrow 2^{\mathcal{Y}}: x \mapsto Ax + \lambda Bx. \quad (1.10)$$

Thus,  $\text{gra}(A + \lambda B) = \{(x, u + \lambda v) \mid (x, u) \in \text{gra } A, (x, v) \in \text{gra } B\}$  and  $\text{dom}(A + \lambda B) = \text{dom } A \cap \text{dom } B$ . Now suppose that  $\mathcal{X}$  is a real vector space and let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ . Then  $T$  is *positively homogeneous* if

$$(\forall x \in \mathcal{X})(\forall \lambda \in \mathbb{R}_{++}) \quad T(\lambda x) = \lambda Tx, \quad (1.11)$$

and  $T$  is *affine* if

$$(\forall x \in \mathcal{X})(\forall y \in \mathcal{X})(\forall \lambda \in \mathbb{R}) \quad T(\lambda x + (1 - \lambda)y) = \lambda Tx + (1 - \lambda)Ty. \quad (1.12)$$

Note that  $T$  is affine if and only if  $x \mapsto Tx - T0$  is linear.

Finally, suppose that  $\mathcal{X}$  is a real vector space and let  $A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$ . Then the *translation* of  $A$  by  $y \in \mathcal{X}$  is  $\tau_y A: x \mapsto A(x - y)$  and the *reversal* of  $A$  is  $A^\vee: x \mapsto A(-x)$ .

## 1.3 Order

Let  $A$  be a nonempty set and let  $\preccurlyeq$  be a binary relation on  $A \times A$ . Consider the following statements:

- ①  $(\forall a \in A) \quad a \preccurlyeq a$ .
- ②  $(\forall a \in A)(\forall b \in A)(\forall c \in A) \quad [a \preccurlyeq b \text{ and } b \preccurlyeq c] \Rightarrow a \preccurlyeq c$ .
- ③  $(\forall a \in A)(\forall b \in A)(\exists c \in A) \quad a \preccurlyeq c \text{ and } b \preccurlyeq c$ .
- ④  $(\forall a \in A)(\forall b \in A) \quad [a \preccurlyeq b \text{ and } b \preccurlyeq a] \Rightarrow a = b$ .
- ⑤  $(\forall a \in A)(\forall b \in A) \quad a \preccurlyeq b \text{ or } b \preccurlyeq a$ .

If ①, ②, and ③ are satisfied, then  $(A, \preccurlyeq)$  is a *directed set*. If ①, ②, and ④ hold, then  $(A, \preccurlyeq)$  is a *partially ordered set*. If  $(A, \preccurlyeq)$  is a partially ordered set such that ⑤ holds, then  $(A, \preccurlyeq)$  is a *totally ordered set*. Unless mentioned otherwise, nonempty subsets of  $\mathbb{R}$  will be totally ordered and directed by  $\leq$ . A totally ordered subset of a partially ordered set is often called a *chain*. Let  $(A, \preccurlyeq)$  be a partially ordered set and let  $B \subset A$ . Then  $a \in A$  is an *upper bound* of  $B$  if  $(\forall b \in B) \quad b \preccurlyeq a$ , and a *lower bound* of  $B$  if  $(\forall b \in B) \quad a \preccurlyeq b$ . Furthermore,  $b \in B$  is the *least element* of  $B$  if  $(\forall c \in B) \quad b \preccurlyeq c$ . Finally,  $a \in A$  is a *maximal element* of  $A$  if  $(\forall c \in A) \quad a \preccurlyeq c \Rightarrow c = a$ .

**Fact 1.1 (Zorn's lemma)** *Let  $A$  be a partially ordered set such that every chain in  $A$  has an upper bound. Then  $A$  contains a maximal element.*

## 1.4 Nets

Let  $(A, \preceq)$  be a directed set. For  $a$  and  $b$  in  $A$ , the notation  $b \succcurlyeq a$  means  $a \preceq b$ . Let  $\mathcal{X}$  be a nonempty set. A *net* (or *generalized sequence*) in  $\mathcal{X}$  indexed by  $A$  is an operator from  $A$  to  $\mathcal{X}$  and it is denoted by  $(x_a)_{a \in A}$ . Let  $\mathbb{N} = \{0, 1, \dots\}$ . Since  $(\mathbb{N}, \leq)$  is a directed set, every sequence is a net;  $(a)_{a \in ]0, 1[}$  is an example of a net that is not a sequence.

Let  $(x_a)_{a \in A}$  be a net in  $\mathcal{X}$  and let  $\mathcal{Y} \subset \mathcal{X}$ . Then  $(x_a)_{a \in A}$  is *eventually* in  $\mathcal{Y}$  if

$$(\exists c \in A)(\forall a \in A) \quad a \succcurlyeq c \quad \Rightarrow \quad x_a \in \mathcal{Y}, \quad (1.13)$$

and it is *frequently* in  $\mathcal{Y}$  if

$$(\forall c \in A)(\exists a \in A) \quad a \succcurlyeq c \quad \text{and} \quad x_a \in \mathcal{Y}. \quad (1.14)$$

A net  $(y_b)_{b \in B}$  is a *subnet* of  $(x_a)_{a \in A}$  via  $k: B \rightarrow A$  if

$$(\forall b \in B) \quad y_b = x_{k(b)} \quad (1.15)$$

and

$$(\forall a \in A)(\exists d \in B)(\forall b \in B) \quad b \succcurlyeq d \Rightarrow k(b) \succcurlyeq a. \quad (1.16)$$

The notations  $(x_{k(b)})_{b \in B}$  and  $(x_{k_b})_{b \in B}$  will also be used for subnets. Thus,  $(y_b)_{b \in B}$  is a subsequence of  $(x_a)_{a \in A}$  when  $A = B = \mathbb{N}$  and  $(y_b)_{b \in B}$  is a subnet of  $(x_a)_{a \in A}$  via some strictly increasing function  $k: \mathbb{N} \rightarrow \mathbb{N}$ .

**Remark 1.2** A subnet of a sequence  $(x_n)_{n \in \mathbb{N}}$  need not be a subsequence. For instance, let  $B$  be a nonempty subset of  $\mathbb{R}$  that is unbounded above and suppose that the function  $k: B \rightarrow \mathbb{N}$  satisfies  $k(b) \rightarrow +\infty$  as  $b \rightarrow +\infty$ . Then  $(x_{k(b)})_{b \in B}$  is a subnet of  $(x_n)_{n \in \mathbb{N}}$ . However, if  $B$  is uncountable, then  $(x_{k(b)})_{b \in B}$  is not a subsequence. Likewise, if  $B = \mathbb{N}$  and  $k$  is not strictly increasing, e.g.,  $k: b \mapsto b(2 + (-1)^b)$ , then  $(x_{k(b)})_{b \in B}$  is not a subsequence.

## 1.5 The Extended Real Line

One obtains the *extended real line*  $[-\infty, +\infty] = \mathbb{R} \cup \{-\infty\} \cup \{+\infty\}$  by adjoining the elements  $-\infty$  and  $+\infty$  to the real line  $\mathbb{R}$  and extending the order via  $(\forall \xi \in \mathbb{R}) \quad -\infty < \xi < +\infty$ . Arithmetic rules are extended to elements of  $[-\infty, +\infty]$  in the usual fashion, leaving expressions such as  $+\infty + (-\infty)$ ,  $0 \cdot (+\infty)$ , and  $+\infty / +\infty$  undefined, unless mentioned otherwise. Given  $\xi \in \mathbb{R}$ ,  $]\xi, +\infty] = ]\xi, +\infty[ \cup \{+\infty\}$ ; the other extended intervals are defined similarly.

**Remark 1.3** Throughout this book, we use the following terminology.

- (i) For extended real numbers, *positive* means  $\geq 0$ , *strictly positive* means  $> 0$ , *negative* means  $\leq 0$ , and *strictly negative* means  $< 0$ . Moreover,

$\mathbb{R}_+ = [0, +\infty[ = \{\xi \in \mathbb{R} \mid \xi \geq 0\}$  and  $\mathbb{R}_{++} = ]0, +\infty[ = \{\xi \in \mathbb{R} \mid \xi > 0\}$ . Likewise, if  $N$  is a strictly positive integer, the *positive orthant* is  $\mathbb{R}_+^N = [0, +\infty[^N$  and the *strictly positive orthant* is  $\mathbb{R}_{++}^N = ]0, +\infty[^N$ . The sets  $\mathbb{R}_-$  and  $\mathbb{R}_{--}$ , as well as the *negative orthants*,  $\mathbb{R}_-^N$  and  $\mathbb{R}_{--}^N$ , are defined similarly.

- (ii) Let  $D \subset \mathbb{R}$ . Then a function  $f: D \rightarrow [-\infty, +\infty]$  is *increasing* if, for every  $\xi$  and  $\eta$  in  $D$  such that  $\xi < \eta$ , we have  $f(\xi) \leq f(\eta)$  (*strictly increasing* if  $f(\xi) < f(\eta)$ ). Applying these definitions to  $-f$  yields the notions of a *decreasing* and of a *strictly decreasing* function, respectively.

Let  $S \subset [-\infty, +\infty]$ . A number  $\gamma \in [-\infty, +\infty]$  is the (necessarily unique) *infimum* (or the greatest lower bound) of  $S$  if it is a lower bound of  $S$  and if, for every lower bound  $\delta$  of  $S$ , we have  $\delta \leq \gamma$ . This number is denoted by  $\inf S$ , and by  $\min S$  when  $\inf S \in S$ . The *supremum* (least upper bound) of  $S$  is  $\sup S = -\inf \{-\alpha \mid \alpha \in S\}$ . This number is denoted by  $\max S$  when  $\sup S \in S$ . The set  $S$  always admits an infimum and a supremum. Note that  $\inf \emptyset = +\infty$  and  $\sup \emptyset = -\infty$ .

The *limit inferior* of a net  $(\xi_a)_{a \in A}$  in  $[-\infty, +\infty]$  is

$$\underline{\lim} \xi_a = \sup_{a \in A} \inf_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b \quad (1.17)$$

and its *limit superior* is

$$\overline{\lim} \xi_a = \inf_{a \in A} \sup_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b. \quad (1.18)$$

It is clear that  $\underline{\lim} \xi_a \leq \overline{\lim} \xi_a$ .

## 1.6 Functions

Let  $\mathcal{X}$  be a nonempty set.

**Definition 1.4** Let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ . The *domain* of  $f$  is

$$\text{dom } f = \{x \in \mathcal{X} \mid f(x) < +\infty\}, \quad (1.19)$$

the *graph* of  $f$  is

$$\text{gra } f = \{(x, \xi) \in \mathcal{X} \times \mathbb{R} \mid f(x) = \xi\}, \quad (1.20)$$

the *epigraph* of  $f$  is

$$\text{epi } f = \{(x, \xi) \in \mathcal{X} \times \mathbb{R} \mid f(x) \leq \xi\}, \quad (1.21)$$

the *lower level set* of  $f$  at height  $\xi \in \mathbb{R}$  is

$$\text{lev}_{\leq \xi} f = \{x \in \mathcal{X} \mid f(x) \leq \xi\}, \quad (1.22)$$

and the *strict lower level set* of  $f$  at height  $\xi \in \mathbb{R}$  is

$$\text{lev}_{< \xi} f = \{x \in \mathcal{X} \mid f(x) < \xi\}. \quad (1.23)$$

The function  $f$  is *proper* if  $-\infty \notin f(\mathcal{X})$  and  $\text{dom } f \neq \emptyset$ . In addition, the closures of  $\text{dom } f$  and  $\text{epi } f$  are respectively denoted by  $\overline{\text{dom } f}$  and  $\overline{\text{epi } f}$ .

### Remark 1.5

- (i) Strictly speaking,  $\text{dom } f$  in (1.19) does not correspond to the domain of  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  viewed as an operator (which is  $\mathcal{X}$  in this case in light of our conventions) and it is sometimes called the *effective domain*. However, it is customary in convex analysis to call it simply the domain of  $f$  and to denote it still by  $\text{dom } f$ .
- (ii) Let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  and  $g: \mathcal{X} \rightarrow [-\infty, +\infty]$ . Then the sum  $f + g: \mathcal{X} \rightarrow [-\infty, +\infty]$  is defined pointwise using the convention  $+\infty + (-\infty) = +\infty$ , which yields  $\text{dom}(f + g) = \text{dom } f \cap \text{dom } g$ .

**Lemma 1.6** *Let  $(f_i)_{i \in I}$  be a family of functions from  $\mathcal{X}$  to  $[-\infty, +\infty]$ . Then the following hold:*

- (i)  $\text{epi} \left( \sup_{i \in I} f_i \right) = \bigcap_{i \in I} \text{epi } f_i$ .
- (ii) *If  $I$  is finite, then  $\text{epi} \left( \min_{i \in I} f_i \right) = \bigcup_{i \in I} \text{epi } f_i$ .*

*Proof.* Take  $(x, \xi) \in \mathcal{X} \times \mathbb{R}$ .

(i):  $(x, \xi) \in \text{epi} \left( \sup_{i \in I} f_i \right) \Leftrightarrow \sup_{i \in I} f_i(x) \leq \xi \Leftrightarrow (\forall i \in I) f_i(x) \leq \xi \Leftrightarrow (\forall i \in I) (x, \xi) \in \text{epi } f_i \Leftrightarrow (x, \xi) \in \bigcap_{i \in I} \text{epi } f_i$ .

(ii):  $(x, \xi) \in \text{epi} \left( \min_{i \in I} f_i \right) \Leftrightarrow \min_{i \in I} f_i(x) \leq \xi \Leftrightarrow (\exists i \in I) f_i(x) \leq \xi \Leftrightarrow (\exists i \in I) (x, \xi) \in \text{epi } f_i \Leftrightarrow (x, \xi) \in \bigcup_{i \in I} \text{epi } f_i$ .  $\square$

**Definition 1.7** Let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  and let  $C \subset \mathcal{X}$ . The *infimum* of  $f$  over  $C$  is  $\inf f(C)$ ; it is also denoted by  $\inf_{x \in C} f(x)$ . Moreover,  $f$  achieves its infimum over  $C$  if there exists  $y \in C$  such that  $f(y) = \inf f(C)$ . In this case, we write  $f(y) = \min f(C)$  or  $f(y) = \min_{x \in C} f(x)$  and call  $\min f(C)$  the minimum of  $f$  over  $C$ . Likewise, the *supremum* of  $f$  over  $C$  is  $\sup f(C)$ ; it is also denoted by  $\sup_{y \in C} f(y)$ . Moreover,  $f$  achieves its supremum over  $C$  if there exists  $x \in C$  such that  $f(x) = \sup f(C)$ . In this case, we write  $f(x) = \max f(C)$  or  $f(x) = \max_{y \in C} f(y)$  and call  $\max f(C)$  the maximum of  $f$  over  $C$ .

**Definition 1.8** Let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{dom } f$ . Then  $(x_n)_{n \in \mathbb{N}}$  is a *minimizing sequence* of  $f$  if  $f(x_n) \rightarrow \inf f(\mathcal{X})$ .



## 1.7 Topological Spaces

Let  $\mathcal{X}$  be a set and let  $\mathcal{T}$  be a family of subsets of  $\mathcal{X}$  that contains  $\mathcal{X}$ ,  $\emptyset$ , as well as all arbitrary unions and finite intersections of its elements. Then  $\mathcal{T}$  is a *topology* and  $(\mathcal{X}, \mathcal{T})$ —or simply  $\mathcal{X}$  if a topology is assumed—is a *topological space*. The elements of  $\mathcal{T}$  are called *open* sets and their complements in  $\mathcal{X}$  *closed* sets. A *neighborhood* of  $x \in \mathcal{X}$  is a subset  $V$  of  $\mathcal{X}$  such that  $x \in U \subset V$  for some  $U \in \mathcal{T}$ . The family of all neighborhoods of  $x$  is denoted by  $\mathcal{V}(x)$ . A subfamily  $\mathcal{B}$  of  $\mathcal{T}$  is a *base* of  $\mathcal{T}$  if, for every  $x \in \mathcal{X}$  and every  $V \in \mathcal{V}(x)$ , there exists a set  $B \in \mathcal{B}$  such that  $x \in B \subset V$ . If  $\mathcal{B}$  is a base of  $\mathcal{T}$ , then every set in  $\mathcal{T}$  can be written as a union of elements in  $\mathcal{B}$ . Now let  $C$  be a subset of  $\mathcal{X}$ . Then the *interior* of  $C$  is the largest open set that is contained in  $C$ ; it is denoted by  $\text{int } C$ . A point  $x \in \mathcal{X}$  belongs to  $\text{int } C$  if and only if  $(\exists V \in \mathcal{V}(x)) V \subset C$ . The *closure* of  $C$  is the smallest closed set that contains  $C$ ; it is denoted by  $\overline{C}$ . If  $\overline{C} = \mathcal{X}$ , then  $C$  is *dense* in  $\mathcal{X}$ . A point  $x \in \mathcal{X}$  belongs to  $\overline{C}$  if and only if  $(\forall V \in \mathcal{V}(x)) V \cap C \neq \emptyset$ . The *boundary* of  $C$  is  $\text{bdry } C = \overline{C} \setminus (\text{int } C)$ . If  $\mathcal{X}_1$  and  $\mathcal{X}_2$  are topological spaces with respective bases  $\mathcal{B}_1$  and  $\mathcal{B}_2$ , the Cartesian product  $\mathcal{X}_1 \times \mathcal{X}_2$  will be considered as a topological space equipped with the *product topology*, i.e., the topology that admits

$$\mathcal{B} = \{B_1 \times B_2 \mid B_1 \in \mathcal{B}_1 \text{ and } B_2 \in \mathcal{B}_2\} \quad (1.24)$$

as a base.

The topological space  $\mathcal{X}$  is a *Hausdorff space* if, for any two distinct points  $x_1$  and  $x_2$  in  $\mathcal{X}$ , there exist  $V_1 \in \mathcal{V}(x_1)$  and  $V_2 \in \mathcal{V}(x_2)$  such that  $V_1 \cap V_2 = \emptyset$ . Let  $\mathcal{X}$  be a Hausdorff space. A subset  $C$  of  $\mathcal{X}$  is *compact* if, whenever  $C$  is contained in the union of a family of open sets, it is also contained in the union of a finite subfamily from that family. Now let  $(x_a)_{a \in A}$  be a net in  $\mathcal{X}$ . Then  $(x_a)_{a \in A}$  *converges* to a (necessarily unique) limit point  $x \in \mathcal{X}$ , in symbols,  $x_a \rightarrow x$  or  $\lim x_a = x$ , if  $(x_a)_{a \in A}$  lies eventually in every neighborhood of  $x$ , i.e.,

$$(\forall V \in \mathcal{V}(x))(\exists b \in A)(\forall a \in A) \quad a \succ b \quad \Rightarrow \quad x_a \in V. \quad (1.25)$$

**Fact 1.9** *Let  $(x_a)_{a \in A}$  be a net in a Hausdorff space  $\mathcal{X}$  that converges to a point  $x \in \mathcal{X}$  and let  $(x_{k(b)})_{b \in B}$  be a subnet of  $(x_a)_{a \in A}$ . Then  $x_{k(b)} \rightarrow x$ .*

A point  $x \in \mathcal{X}$  is a *cluster point* of  $(x_a)_{a \in A}$  if  $(x_a)_{a \in A}$  lies frequently in every neighborhood of  $x$ , i.e.,

$$(\forall V \in \mathcal{V}(x))(\forall b \in A)(\exists a \in A) \quad a \succ b \quad \text{and} \quad x_a \in V. \quad (1.26)$$

Alternatively,  $x$  is a cluster point of  $(x_a)_{a \in A}$  if  $(x_a)_{a \in A}$  possesses a subnet that converges to  $x$ . If a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{X}$  possesses a subsequence that converges to a point  $x \in \mathcal{X}$ , then  $x$  is a *sequential cluster point* of  $(x_n)_{n \in \mathbb{N}}$ .

Topological notions can be conveniently characterized using nets.

**Lemma 1.10** *Let  $C$  be a subset of a Hausdorff space  $\mathcal{X}$  and let  $x \in \mathcal{X}$ . Then  $x \in \overline{C}$  if and only if there exists a net in  $C$  that converges to  $x$ .*

*Proof.* First, suppose that  $x \in \overline{C}$  and direct  $A = \mathcal{V}(x)$  via  $(\forall a \in A)(\forall b \in A) a \preceq b \Leftrightarrow a \supset b$ . For every  $a \in \mathcal{V}(x)$ , there exists  $x_a \in C \cap a$ . The net  $(x_a)_{a \in A}$  lies in  $C$  and, by (1.25), it converges to  $x$ . Conversely, let  $(x_a)_{a \in A}$  be a net in  $C$  such that  $x_a \rightarrow x$  and let  $V \in \mathcal{V}(x)$ . Then  $(x_a)_{a \in A}$  is eventually in  $V$  and therefore  $C \cap V \neq \emptyset$ . Thus,  $x \in \overline{C}$ .  $\square$

Let  $C$  be a subset of  $\mathcal{X}$ . It follows from Lemma 1.10 that  $C$  is closed if and only if the limit of every convergent net that lies in  $C$  belongs to  $C$ . Likewise,  $C$  is open if and only if, for every point  $x \in C$ , every net in  $\mathcal{X}$  that converges to  $x$  is eventually in  $C$ .

**Fact 1.11** *Let  $C$  be a subset of a Hausdorff space  $\mathcal{X}$ . Then the following are equivalent:*

- (i)  $C$  is compact.
- (ii)  $C \cap \bigcap_{j \in J} C_j \neq \emptyset$  for every family  $(C_j)_{j \in J}$  of closed subsets of  $\mathcal{X}$  such that, for every finite subset  $I$  of  $J$ ,  $C \cap \bigcap_{i \in I} C_i \neq \emptyset$ .
- (iii) Every net in  $C$  has a cluster point in  $C$ .
- (iv) Every net in  $C$  has a subnet that converges to a point in  $C$ .

**Lemma 1.12** *Let  $C$  be a compact subset of a Hausdorff space  $\mathcal{X}$ . Then  $C$  is closed, and every closed subset of  $C$  is compact.*

*Proof.* Let  $(x_a)_{a \in A}$  be a net in  $C$  that converges to a point  $x \in \mathcal{X}$ . By Fact 1.11, there exists a subnet  $(x_{k(b)})_{b \in B}$  of  $(x_a)_{a \in A}$  that converges to a point  $y \in C$ . Therefore  $x = y \in C$  and  $C$  is closed. Now let  $D$  be a closed subset of  $C$  and let  $(x_a)_{a \in A}$  be a net in  $D$ . Then  $(x_a)_{a \in A}$  lies in  $C$  and, as above, there exists a subnet  $(x_{k(b)})_{b \in B}$  of  $(x_a)_{a \in A}$  that converges to a point  $y \in C$ . However, since  $D$  is closed,  $y \in D$ . We conclude that  $D$  is compact.  $\square$

**Remark 1.13** Let  $C$  be a compact subset of a Hausdorff space  $\mathcal{X}$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $C$ . By Fact 1.11,  $(x_n)_{n \in \mathbb{N}}$  possesses a convergent subnet. However, it may happen that no subsequence of  $(x_n)_{n \in \mathbb{N}}$  converges (see [101, Chapter 13] for an example).

**Lemma 1.14** *Let  $C$  be a compact subset of a Hausdorff space  $\mathcal{X}$  and suppose that  $(x_a)_{a \in A}$  is a net in  $C$  that admits a unique cluster point  $x \in \mathcal{X}$ . Then  $x_a \rightarrow x$ .*

*Proof.* Suppose that  $x_a \not\rightarrow x$ . Then it follows from (1.25) that there exist a subnet  $(x_{k(a)})_{a \in A}$  of  $(x_a)_{a \in A}$  and an open set  $V \in \mathcal{V}(x)$  such that  $(x_{k(a)})_{a \in A}$  lies in  $C \setminus V$ . Since  $C \setminus V$  is compact by Lemma 1.12,  $(x_{k(a)})_{a \in A}$  possesses a cluster point  $y \in C \setminus V$ . Hence,  $y \neq x \in V$  and  $y$  is a cluster point of  $(x_a)_{a \in A}$ , which contradicts the assumption that  $x$  is the unique cluster point of  $(x_a)_{a \in A}$ .  $\square$

## 1.8 Two-Point Compactification of the Real Line

Unless stated otherwise, the real line  $\mathbb{R}$  will always be equipped with the usual topology, a base of which is the family of open intervals. With this topology,  $\mathbb{R}$  is a Hausdorff space that is not compact. The extended real line  $[-\infty, +\infty]$  equipped with the topology that admits as a base the open real intervals and the intervals of the form  $[-\infty, \xi[$  and  $]\xi, +\infty]$ , where  $\xi \in \mathbb{R}$ , is a compact space.

**Fact 1.15** *Let  $(\xi_a)_{a \in A}$  be a net in  $[-\infty, +\infty]$ . Then the following hold:*

- (i) *The nets  $\left(\inf_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b\right)_{a \in A}$  and  $\left(\sup_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b\right)_{a \in A}$  converge to  $\underline{\lim} \xi_a$  and  $\overline{\lim} \xi_a$ , respectively.*
- (ii) *The net  $(\xi_a)_{a \in A}$  possesses subnets that converge to  $\underline{\lim} \xi_a$  and  $\overline{\lim} \xi_a$ , respectively.*
- (iii) *The net  $(\xi_a)_{a \in A}$  converges if and only if  $\underline{\lim} \xi_a = \overline{\lim} \xi_a$ , in which case  $\lim \xi_a = \underline{\lim} \xi_a = \overline{\lim} \xi_a$ .*

Moreover, if  $(\xi_a)_{a \in A}$  is a sequence, then (ii) remains true if subnets are replaced by subsequences.

**Lemma 1.16** *Let  $(\xi_a)_{a \in A}$  and  $(\eta_a)_{a \in A}$  be nets in  $[-\infty, +\infty]$  such that  $\underline{\lim} \xi_a > -\infty$  and  $\underline{\lim} \eta_a > -\infty$ . Then  $\underline{\lim} \xi_a + \underline{\lim} \eta_a \leq \underline{\lim}(\xi_a + \eta_a)$ .*

*Proof.* Let  $a \in A$ . Clearly,

$$(\forall c \in A) \quad a \preccurlyeq c \quad \Rightarrow \quad \inf_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b + \inf_{\substack{b \in A \\ a \preccurlyeq b}} \eta_b \leq \xi_c + \eta_c. \quad (1.27)$$

Hence

$$\inf_{\substack{b \in A \\ a \preccurlyeq b}} \xi_b + \inf_{\substack{b \in A \\ a \preccurlyeq b}} \eta_b \leq \inf_{\substack{c \in A \\ a \preccurlyeq c}} (\xi_c + \eta_c). \quad (1.28)$$

In view of Fact 1.15(i), the result follows by taking limits over  $a$  in (1.28).  $\square$

## 1.9 Continuity

**Definition 1.17** Let  $(\mathcal{X}, \mathcal{T}_{\mathcal{X}})$  and  $(\mathcal{Y}, \mathcal{T}_{\mathcal{Y}})$  be topological spaces and let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ . Then  $T$  is *continuous* at  $x \in \mathcal{X}$  if

$$(\forall W \in \mathcal{V}(Tx))(\exists V \in \mathcal{V}(x)) \quad T(V) \subset W. \quad (1.29)$$

Moreover,  $T$  is continuous if it is continuous at every point in  $\mathcal{X}$ .

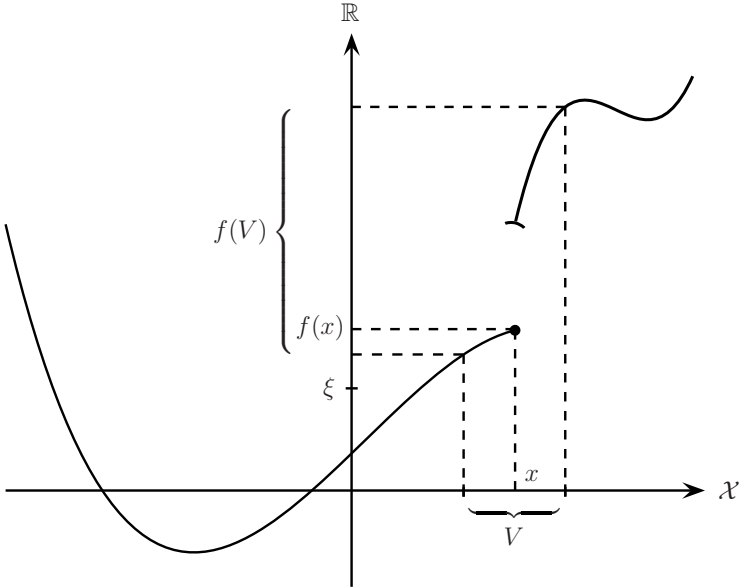
**Fact 1.18** *Let  $(\mathcal{X}, \mathcal{T}_{\mathcal{X}})$  and  $(\mathcal{Y}, \mathcal{T}_{\mathcal{Y}})$  be topological spaces, let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ , and suppose that  $\mathcal{B}_{\mathcal{Y}}$  is a base of  $\mathcal{T}_{\mathcal{Y}}$ . Then  $T$  is continuous if and only if  $(\forall B \in \mathcal{B}_{\mathcal{Y}}) T^{-1}(B) \in \mathcal{T}_{\mathcal{X}}$ .*

**Fact 1.19** Let  $\mathcal{X}$  and  $\mathcal{Y}$  be Hausdorff spaces, let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ , and let  $x \in \mathcal{X}$ . Then  $T$  is continuous at  $x$  if and only if  $Tx_a \rightarrow Tx$  for every net  $(x_a)_{a \in A}$  in  $\mathcal{X}$  that converges to  $x$ .

**Lemma 1.20** Let  $\mathcal{X}$  and  $\mathcal{Y}$  be Hausdorff spaces, let  $T: \mathcal{X} \rightarrow \mathcal{Y}$  be continuous, and let  $C \subset \mathcal{X}$  be compact. Then  $T(C)$  is compact.

*Proof.* Let  $(y_a)_{a \in A}$  be a net in  $T(C)$ . Then there exists a net  $(x_a)_{a \in A}$  in  $C$  such that  $(\forall a \in A) y_a = Tx_a$ . By Fact 1.11, we can find a subnet  $(x_{k(b)})_{b \in B}$  of  $(x_a)_{a \in A}$  that converges to a point  $x \in C$ . Fact 1.19 implies that  $Tx_{k(b)} \rightarrow Tx \in T(C)$ . Therefore, the claim follows from Fact 1.11.  $\square$

## 1.10 Lower Semicontinuity



**Fig. 1.1** The function  $f$  is lower semicontinuous at  $x$ : for every  $\xi \in ]-\infty, f(x)[$ , we can find a neighborhood  $V$  of  $x$  such that  $f(V) \subset [\xi, +\infty]$ .

**Definition 1.21** Let  $\mathcal{X}$  be a Hausdorff space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ , and let  $x \in \mathcal{X}$ . Then  $f$  is *lower semicontinuous* at  $x$  if, for every net  $(x_a)_{a \in A}$  in  $\mathcal{X}$ ,

$$x_a \rightarrow x \Rightarrow f(x) \leq \underline{\lim} f(x_a) \quad (1.30)$$

or, equivalently, if (see Figure 1.1)

$$(\forall \xi \in ]-\infty, f(x)[)(\exists V \in \mathcal{V}(x)) \quad f(V) \subset ]\xi, +\infty]. \quad (1.31)$$

*Upper semicontinuity* of  $f$  at  $x \in \mathcal{X}$  holds if  $-f$  is lower semicontinuous at  $x$ , i.e., if  $x_a \rightarrow x \Rightarrow \underline{\lim} f(x_a) \leq f(x)$  for every net  $(x_a)_{a \in A}$  in  $\mathcal{X}$ ; if  $f$  is lower and upper semicontinuous at  $x$ , it is *continuous* at  $x$ , i.e.,  $x_a \rightarrow x \Rightarrow f(x_a) \rightarrow f(x)$  for every net  $(x_a)_{a \in A}$  in  $\mathcal{X}$ . Finally, the *domain of continuity* of  $f$  is the set

$$\text{cont } f = \{x \in \mathcal{X} \mid f(x) \in \mathbb{R} \text{ and } f \text{ is continuous at } x\}. \quad (1.32)$$

Note that  $\text{cont } f \subset \text{int dom } f$ , since  $f$  cannot be both continuous and real-valued on the boundary of its domain. Let us also observe that a continuous function may take on the values  $-\infty$  and  $+\infty$ .

**Example 1.22** Let  $\mathcal{X} = \mathbb{R}$  and set

$$f: \mathbb{R} \rightarrow [-\infty, +\infty] : x \mapsto \begin{cases} 1/x, & \text{if } x > 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (1.33)$$

Then  $f$  is continuous on  $\mathbb{R}$  and  $\text{cont } f = \mathbb{R}_{++}$  is a proper open subset of  $\mathcal{X}$ .

Let  $\mathcal{X}$  be a Hausdorff space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ , and let  $x \in \mathcal{X}$ . Then

$$\underline{\lim}_{y \rightarrow x} f(y) = \sup_{V \in \mathcal{V}(x)} \inf f(V). \quad (1.34)$$

**Lemma 1.23** Let  $\mathcal{X}$  be a Hausdorff space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ , let  $x \in \mathcal{X}$ , and let  $N(x)$  be the set of all nets in  $\mathcal{X}$  that converge to  $x$ . Then

$$\underline{\lim}_{y \rightarrow x} f(y) = \min_{(x_a)_{a \in A} \in N(x)} \underline{\lim} f(x_a). \quad (1.35)$$

*Proof.* Let  $(x_a)_{a \in A} \in N(x)$  and set  $(\forall a \in A) \mu_a = \inf \{f(x_b) \mid a \preccurlyeq b\}$ . By Fact 1.15(i),  $\lim \mu_a = \underline{\lim} f(x_a)$ . For every  $V \in \mathcal{V}(x)$ , there exists  $a_V \in A$  such that  $a \succcurlyeq a_V \Rightarrow x_a \in V$ . Thus

$$(\forall a \succcurlyeq a_V) \quad \mu_a = \inf \{f(x_b) \mid a \preccurlyeq b\} \geq \inf f(V). \quad (1.36)$$

Hence for every  $V \in \mathcal{V}(x)$ ,  $\underline{\lim} f(x_a) = \lim \mu_a = \lim_{a \succcurlyeq a_V} \mu_a \geq \inf f(V)$ , which implies that

$$\inf \{\underline{\lim} f(x_a) \mid x_a \rightarrow x\} \geq \sup_{V \in \mathcal{V}(x)} \inf f(V). \quad (1.37)$$

Set  $B = \{(y, V) \mid y \in V \in \mathcal{V}(x)\}$  and direct  $B$  via  $(y, V) \preccurlyeq (z, W) \Leftrightarrow W \subset V$ . For every  $b = (y, V) \in B$ , set  $x_b = y$ . Then  $x_b \rightarrow x$  and

$$\begin{aligned}
\liminf f(x_b) &= \sup_{b \in B} \inf \{f(x_c) \mid b \preccurlyeq c\} \\
&= \sup_{y \in V \in \mathcal{V}(x)} \inf \{f(z) \mid z \in V\} \\
&= \sup_{V \in \mathcal{V}(x)} \inf f(V).
\end{aligned} \tag{1.38}$$

Altogether, (1.34), (1.37), and (1.38) yield (1.35).  $\square$

**Lemma 1.24** *Let  $\mathcal{X}$  be a Hausdorff space and let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ . Then the following are equivalent:*

- (i)  *$f$  is lower semicontinuous, i.e.,  $f$  is lower semicontinuous at every point in  $\mathcal{X}$ .*
- (ii)  *$\text{epi } f$  is closed in  $\mathcal{X} \times \mathbb{R}$ .*
- (iii) *For every  $\xi \in \mathbb{R}$ ,  $\text{lev}_{\leq \xi} f$  is closed in  $\mathcal{X}$ .*

*Proof.* (i) $\Rightarrow$ (ii): Let  $(x_a, \xi_a)_{a \in A}$  be a net in  $\text{epi } f$  that converges to  $(x, \xi)$  in  $\mathcal{X} \times \mathbb{R}$ . Then  $f(x) \leq \liminf f(x_a) \leq \liminf \xi_a = \xi$  and, hence,  $(x, \xi) \in \text{epi } f$ .

(ii) $\Rightarrow$ (iii): Fix  $\xi \in \mathbb{R}$  and assume that  $(x_a)_{a \in A}$  is a net in  $\text{lev}_{\leq \xi} f$  that converges to  $x$ . Then the net  $(x_a, \xi)_{a \in A}$  lies in  $\text{epi } f$  and converges to  $(x, \xi)$ . Since  $\text{epi } f$  is closed, we deduce that  $(x, \xi) \in \text{epi } f$  and, hence, that  $x \in \text{lev}_{\leq \xi} f$ .

(iii) $\Rightarrow$ (i): Fix  $x \in \mathcal{X}$ , let  $(x_a)_{a \in A}$  be a net in  $\mathcal{X}$  that converges to  $x$ , and set  $\mu = \liminf f(x_a)$ . Then it suffices to show that  $f(x) \leq \mu$ . When  $\mu = +\infty$ , the inequality is clear and we therefore assume that  $\mu < +\infty$ . By Fact 1.15(ii), there exists a subnet  $(x_{k(b)})_{b \in B}$  of  $(x_a)_{a \in A}$  such that  $f(x_{k(b)}) \rightarrow \mu$ . Now fix  $\xi \in ]\mu, +\infty[$ . Then  $(f(x_{k(b)}))_{b \in B}$  is eventually in  $[-\infty, \xi]$  and, therefore, there exists  $c \in B$  such that  $\{x_{k(b)} \mid c \preccurlyeq b \in B\} \subset \text{lev}_{\leq \xi} f$ . Since  $x_{k(b)} \rightarrow x$  and since  $\text{lev}_{\leq \xi} f$  is closed, we deduce that  $x \in \text{lev}_{\leq \xi} f$ , i.e.,  $f(x) \leq \xi$ . Letting  $\xi \downarrow \mu$ , we conclude that  $f(x) \leq \mu$ .  $\square$

**Example 1.25** The *indicator function* of a set  $C \subset \mathcal{X}$ , i.e., the function

$$\iota_C: \mathcal{X} \rightarrow [-\infty, +\infty]: x \mapsto \begin{cases} 0, & \text{if } x \in C; \\ +\infty, & \text{otherwise,} \end{cases} \tag{1.39}$$

is lower semicontinuous if and only if  $C$  is closed.

*Proof.* Take  $\xi \in \mathbb{R}$ . Then  $\text{lev}_{\leq \xi} \iota_C = \emptyset$  if  $\xi < 0$ , and  $\text{lev}_{\leq \xi} \iota_C = C$  otherwise. Hence, the result follows from Lemma 1.24.  $\square$

**Lemma 1.26** *Let  $\mathcal{X}$  be a Hausdorff space and let  $(f_i)_{i \in I}$  be a family of lower semicontinuous functions from  $\mathcal{X}$  to  $[-\infty, +\infty]$ . Then  $\sup_{i \in I} f_i$  is lower semicontinuous. If  $I$  is finite, then  $\min_{i \in I} f_i$  is lower semicontinuous.*

*Proof.* A direct consequence of Lemma 1.6 and Lemma 1.24.  $\square$

**Lemma 1.27** *Let  $\mathcal{X}$  be a Hausdorff space, let  $(f_i)_{i \in I}$  be a finite family of lower semicontinuous functions from  $\mathcal{X}$  to  $]-\infty, +\infty]$ , and let  $(\alpha_i)_{i \in I}$  be in  $\mathbb{R}_{++}$ . Then  $\sum_{i \in I} \alpha_i f_i$  is lower semicontinuous.*

*Proof.* It is clear that, for every  $i \in I$ ,  $\alpha_i f_i$  is lower semicontinuous. Let  $f$  and  $g$  be lower semicontinuous functions from  $\mathcal{X}$  to  $]-\infty, +\infty]$ , and let  $(x_a)_{a \in A}$  be a net that converges to some point  $x \in \mathcal{X}$ . By Lemma 1.16,

$$\begin{aligned} (f + g)(x) &= f(x) + g(x) \leq \underline{\lim} f(x_a) + \underline{\lim} g(x_a) \\ &\leq \underline{\lim} (f(x_a) + g(x_a)) \\ &= \underline{\lim} (f + g)(x_a), \end{aligned} \tag{1.40}$$

which yields the result when  $I$  contains two elements. The general case follows by induction on the number of elements in  $I$ .  $\square$

The classical Weierstrass theorem states that a continuous function defined on a compact set achieves its minimum and its maximum on that set. The following refinement is a fundamental tool in proving the existence of solutions to minimization problems.

**Theorem 1.28 (Weierstrass)** *Let  $\mathcal{X}$  be a Hausdorff space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  be lower semicontinuous, and let  $C$  be a compact subset of  $\mathcal{X}$ . Suppose that  $C \cap \text{dom } f \neq \emptyset$ . Then  $f$  achieves its infimum over  $C$ .*

*Proof.* By definition of  $\inf f(C)$ , there exists a minimizing sequence of  $f + \iota_C$ . By Fact 1.11 and the compactness of  $C$ , we can extract a subnet  $(x_{k(b)})_{b \in B}$  that converges to a point  $x \in C$ . Therefore  $f(x_{k(b)}) \rightarrow \inf f(C) \leq f(x)$ . On the other hand, by lower semicontinuity, we get  $f(x) \leq \underline{\lim} f(x_{k(b)}) = \lim f(x_{k(b)}) = \inf f(C)$ . Altogether,  $f(x) = \inf f(C)$ .  $\square$

**Lemma 1.29** *Let  $\mathcal{X}$  be a Hausdorff space, let  $C$  be a compact Hausdorff space, and let  $\varphi: \mathcal{X} \times C \rightarrow [-\infty, +\infty]$  be lower semicontinuous. Then the marginal function*

$$f: \mathcal{X} \rightarrow [-\infty, +\infty] : x \mapsto \inf \varphi(x, C) \tag{1.41}$$

*is lower semicontinuous and  $(\forall x \in \mathcal{X}) f(x) = \min \varphi(x, C)$ .*

*Proof.* First let us note that, for every  $x \in \mathcal{X}$ ,  $\{x\} \times C$  is compact and hence Theorem 1.28 implies that  $f(x) = \min \varphi(x, C)$ . Now fix  $\xi \in \mathbb{R}$  and let  $(x_a)_{a \in A}$  be a net in  $\text{lev}_{\leq \xi} f$  that converges to some point  $x \in \mathcal{X}$ . Then there exists a net  $(y_a)_{a \in A}$  in  $C$  such that  $(\forall a \in A) f(x_a) = \varphi(x_a, y_a)$ . Since  $C$  is compact, Fact 1.11 yields the existence of a subnet  $(y_{k(b)})_{b \in B}$  that converges to a point  $y \in C$ . It follows that  $(x_{k(b)}, y_{k(b)}) \rightarrow (x, y)$  and, by lower semicontinuity of  $\varphi$ , that  $f(x) \leq \varphi(x, y) \leq \underline{\lim} \varphi(x_{k(b)}, y_{k(b)}) = \underline{\lim} f(x_{k(b)}) \leq \xi$ . Thus  $x \in \text{lev}_{\leq \xi} f$  and  $f$  is therefore lower semicontinuous by Lemma 1.24.  $\square$

**Definition 1.30** Let  $\mathcal{X}$  be a Hausdorff space. The *lower semicontinuous envelope* of  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$  is

$$\bar{f} = \sup \{g: \mathcal{X} \rightarrow [-\infty, +\infty] \mid g \leq f \text{ and } g \text{ is lower semicontinuous}\}. \quad (1.42)$$

**Lemma 1.31** Let  $\mathcal{X}$  be a Hausdorff space and let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ . Then the following hold:

- (i)  $\bar{f}$  is the largest lower semicontinuous function majorized by  $f$ .
- (ii)  $\text{epi } \bar{f}$  is closed.
- (iii)  $\text{dom } f \subset \text{dom } \bar{f} \subset \overline{\text{dom } f}$ .
- (iv)  $(\forall x \in \mathcal{X}) \bar{f}(x) = \underline{\lim}_{y \rightarrow x} f(y)$ .
- (v) Let  $x \in \mathcal{X}$ . Then  $f$  is lower semicontinuous at  $x$  if and only if  $\bar{f}(x) = f(x)$ .
- (vi)  $\text{epi } \bar{f} = \overline{\text{epi } f}$ .

*Proof.* (i): This follows from (1.42) and Lemma 1.26.

(ii): Since  $\bar{f}$  is lower semicontinuous by (i), the closedness of  $\text{epi } \bar{f}$  follows from Lemma 1.24.

(iii): Since  $\bar{f} \leq f$ , we have  $\text{dom } f \subset \text{dom } \bar{f}$ . Now set

$$g: \mathcal{X} \rightarrow [-\infty, +\infty]: x \mapsto \begin{cases} \bar{f}(x), & \text{if } x \in \overline{\text{dom } f}; \\ +\infty, & \text{if } x \notin \overline{\text{dom } f}. \end{cases} \quad (1.43)$$

It follows from (ii) that  $\text{epi } g = \text{epi } \bar{f} \cap (\overline{\text{dom } f} \times \mathbb{R})$  is closed and hence from Lemma 1.24 that  $g$  is lower semicontinuous. On the other hand, for every  $x \in \mathcal{X}$ ,  $g(x) = \bar{f}(x) \leq f(x)$  if  $x \in \overline{\text{dom } f}$ , and  $g(x) = f(x) = +\infty$  if  $x \notin \overline{\text{dom } f}$ . Hence,  $g \leq f$  and thus  $g = \bar{g} \leq \bar{f}$ . We conclude that  $\text{dom } \bar{f} \subset \text{dom } g \subset \overline{\text{dom } f}$ .

(iv): Set  $\tilde{f}: x \mapsto \underline{\lim}_{y \rightarrow x} f(y)$  and let  $x \in \mathcal{X}$ . We first show that  $\tilde{f}$  is lower semicontinuous. To this end, suppose that  $\tilde{f}(x) > -\infty$  and fix  $\xi \in ]-\infty, \tilde{f}(x)[$ . In view of (1.34), there exists  $V \in \mathcal{V}(x)$  such that  $\xi < \inf f(V)$ . Now let  $U$  be an open set such that  $x \in U \subset V$ . Then  $\xi < \inf f(U)$  and  $(\forall y \in U) U \in \mathcal{V}(y)$ . Hence,  $(\forall y \in U) \xi < \sup_{W \in \mathcal{V}(y)} \inf f(W) = \tilde{f}(y)$  and therefore  $\tilde{f}(U) \subset ]\xi, +\infty]$ . It follows from (1.31) that  $\tilde{f}$  is lower semicontinuous at  $x$ . Thus, since (1.34) yields  $\tilde{f} \leq f$ , we derive from (i) that  $\tilde{f} \leq \bar{f}$ . Next, let  $g: \mathcal{X} \rightarrow [-\infty, +\infty]$  be a lower semicontinuous function such that  $g \leq f$ . Then, in view of (1.42), it remains to show that  $g \leq \tilde{f}$  to prove that  $\bar{f} \leq \tilde{f}$ , and hence conclude that  $\bar{f} = \tilde{f}$ . To this end, suppose that  $g(x) > -\infty$  and let  $\eta \in ]-\infty, g(x)[$ . By (1.31), there exists  $V \in \mathcal{V}(x)$  such that  $g(V) \subset ]\eta, +\infty]$ . Hence, (1.34) yields  $\eta \leq \inf g(V) \leq \underline{\lim}_{y \rightarrow x} g(y)$ . Letting  $\eta \uparrow g(x)$ , we obtain  $g(x) \leq \underline{\lim}_{y \rightarrow x} g(y)$ . Thus,

$$g(x) \leq \underline{\lim}_{y \rightarrow x} g(y) = \sup_{W \in \mathcal{V}(x)} \inf g(W) \leq \sup_{W \in \mathcal{V}(x)} \inf f(W) = \tilde{f}(x). \quad (1.44)$$



(v): Suppose that  $f$  is lower semicontinuous at  $x$ . Then it follows from (iv) that  $\bar{f}(x) \leq f(x) \leq \underline{\lim}_{y \rightarrow x} f(y) = \bar{f}(x)$ . Therefore,  $\bar{f}(x) = f(x)$ . Conversely, suppose that  $\bar{f}(x) = f(x)$ . Then  $f(x) = \underline{\lim}_{y \rightarrow x} f(y)$  by (iv) and therefore  $f$  is lower semicontinuous at  $x$ .

(vi): Since  $\bar{f} \leq f$ ,  $\text{epi } f \subset \text{epi } \bar{f}$ , and therefore (ii) yields  $\overline{\text{epi } f} \subset \overline{\text{epi } \bar{f}} = \text{epi } \bar{f}$ . Conversely, let  $(x, \xi) \in \text{epi } \bar{f}$  and fix a neighborhood of  $(x, \xi)$  of the form  $W = V \times [\xi - \varepsilon, \xi + \varepsilon]$ , where  $V \in \mathcal{V}(x)$  and  $\varepsilon \in \mathbb{R}_{++}$ . To show that  $(x, \xi) \in \overline{\text{epi } f}$ , it is enough to show that  $W \cap \text{epi } f \neq \emptyset$ . To this end, note that (1.34) and (iv) imply that  $\bar{f}(x) \geq \inf f(V)$ . Hence, there exists  $y \in V$  such that  $f(y) \leq \bar{f}(x) + \varepsilon \leq \xi + \varepsilon$  and therefore  $(y, \xi + \varepsilon) \in W \cap \text{epi } f$ .  $\square$

## 1.11 Sequential Topological Notions

Let  $\mathcal{X}$  be a Hausdorff space and let  $C$  be a subset of  $\mathcal{X}$ . Then  $C$  is *sequentially closed* if the limit of every convergent sequence  $(x_n)_{n \in \mathbb{N}}$  that lies in  $C$  is also in  $C$ . A closed set is sequentially closed, but the converse is false (see Example 3.31), which shows that, in general, sequences are not adequate to describe topological notions.

**Definition 1.32** A subset  $C$  of a Hausdorff space  $\mathcal{X}$  is *sequentially compact* if every sequence in  $C$  has a sequential cluster point in  $C$ , i.e., if every sequence in  $C$  has a subsequence that converges to a point in  $C$ .

The notions of *sequential continuity* and *sequential lower (upper) semicontinuity* are obtained by replacing nets by sequences in Fact 1.19 and Definition 1.21, respectively. Thus, by replacing nets by sequences and subnets by subsequences in the proofs of Lemma 1.12, Lemma 1.14, and Lemma 1.24, we obtain the following sequential versions.

**Lemma 1.33** *Let  $C$  be a sequentially compact subset of a Hausdorff space  $\mathcal{X}$ . Then  $C$  is sequentially closed, and every sequentially closed subset of  $C$  is sequentially compact.*

**Lemma 1.34** *Let  $C$  be a sequentially compact subset of a Hausdorff space  $\mathcal{X}$  and suppose that  $(x_n)_{n \in \mathbb{N}}$  is a sequence in  $C$  that admits a unique sequential cluster point  $x$ . Then  $x_n \rightarrow x$ .*

**Lemma 1.35** *Let  $\mathcal{X}$  be a Hausdorff space and let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ . Then the following are equivalent:*

- (i)  $f$  is sequentially lower semicontinuous, i.e.,  $f$  is sequentially lower semicontinuous at every point in  $\mathcal{X}$ .
- (ii)  $\text{epi } f$  is sequentially closed in  $\mathcal{X} \times \mathbb{R}$ .
- (iii) For every  $\xi \in \mathbb{R}$ ,  $\text{lev}_{\leq \xi} f$  is sequentially closed in  $\mathcal{X}$ .

**Remark 1.36** Let  $\mathcal{X}$  be a topological space. Then  $\mathcal{X}$  is called *sequential* if every sequentially closed subset of  $\mathcal{X}$  is closed, i.e., if the notions of closedness and sequential closedness coincide. It follows that in sequential spaces the notions of lower semicontinuity and sequential lower semicontinuity are equivalent. Alternatively,  $\mathcal{X}$  is sequential if, for every topological space  $\mathcal{Y}$  and every operator  $T: \mathcal{X} \rightarrow \mathcal{Y}$ , the notions of continuity and sequential continuity coincide. Note, however, that in Hausdorff sequential spaces, the notions of compactness and sequential compactness need not coincide (see [243, Counterexample 43]).

## 1.12 Metric Spaces

Let  $\mathcal{X}$  be a metric space with *distance* (or *metric*)  $d$ . The *diameter* of a subset  $C$  of  $\mathcal{X}$  is  $\text{diam } C = \sup_{(x,y) \in C \times C} d(x,y)$ . The *distance to a set*  $C \subset \mathcal{X}$  is the function

$$d_C: \mathcal{X} \rightarrow [0, +\infty]: x \mapsto \inf d(x, C). \quad (1.45)$$

Note that if  $C = \emptyset$  then  $d_C \equiv +\infty$ . The *closed* and *open balls* of center  $x \in \mathcal{X}$  and radius  $\rho \in \mathbb{R}_{++}$  in  $\mathcal{X}$  are defined as  $B(x; \rho) = \{y \in \mathcal{X} \mid d(x, y) \leq \rho\}$  and  $\{y \in \mathcal{X} \mid d(x, y) < \rho\}$ , respectively. The *metric topology* of  $\mathcal{X}$  is the topology that admits the family of all open balls as a base. A topological space is *metrizable* if its topology coincides with a metric topology.

A sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{X}$  converges to a point  $x \in \mathcal{X}$  if  $d(x_n, x) \rightarrow 0$ . Moreover,  $\mathcal{X}$  is a sequential Hausdorff space and thus, as seen in Remark 1.36, the notions of closedness, continuity, and lower semicontinuity are equivalent to their sequential counterparts.

**Fact 1.37** *Let  $\mathcal{X}$  be a metric space, let  $\mathcal{Y}$  be a Hausdorff space, and let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ . Then  $T$  is continuous if and only if it is sequentially continuous.*

In addition, in metric spaces, the notions of compactness and sequential compactness are equivalent.

**Fact 1.38** *Let  $C$  be a subset of a metric space  $\mathcal{X}$ . Then  $C$  is compact if and only if it is sequentially compact.*

**Lemma 1.39** *Let  $C$  be a subset of a metric space  $\mathcal{X}$  such that  $(\forall n \in \mathbb{N}) C \cap B(0; n)$  is closed. Then  $C$  is closed.*

*Proof.* Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $C$  that converges to a point  $x$ . Since  $(x_n)_{n \in \mathbb{N}}$  is bounded, there exists  $m \in \mathbb{N}$  such that  $(x_n)_{n \in \mathbb{N}}$  lies in  $C \cap B(0; m)$ . The hypothesis implies that  $x \in C \cap B(0; m)$  and hence that  $x \in C$ .  $\square$

**Lemma 1.40** *Let  $C$  be a compact subset of a metric space  $\mathcal{X}$ . Then  $C$  is closed and bounded.*

*Proof.* Closedness follows from Lemma 1.12. Now suppose that  $C$  is not bounded. Then it contains a sequence  $(x_n)_{n \in \mathbb{N}}$  such that  $d(x_0, x_n) \rightarrow +\infty$ . Clearly,  $(x_n)_{n \in \mathbb{N}}$  has no convergent subsequence and  $C$  is therefore not sequentially compact. In view of Fact 1.38,  $C$  is not compact.  $\square$

**Lemma 1.41** *Let  $\mathcal{X}$  be a metric space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ , let  $x \in \mathcal{X}$ , and let  $S(x)$  be the set of all sequences in  $\mathcal{X}$  that converge to  $x$ . Then*

$$\liminf_{y \rightarrow x} f(y) = \min_{(x_n)_{n \in \mathbb{N}} \in S(x)} \liminf f(x_n). \quad (1.46)$$

*Proof.* Let  $(\varepsilon_n)_{n \in \mathbb{N}}$  be a strictly decreasing sequence in  $\mathbb{R}_{++}$  such that  $\varepsilon_n \downarrow 0$ , and set  $(\forall n \in \mathbb{N}) V_n = \{y \in \mathcal{X} \mid d(x, y) < \varepsilon_n\}$  and  $\sigma = \sup_{n \in \mathbb{N}} \inf f(V_n)$ . By Lemma 1.23,

$$\inf \left\{ \liminf_{n \in \mathbb{N}} f(x_n) \mid x_n \rightarrow x \right\} \geq \liminf_{y \rightarrow x} f(y) = \sup_{V \in \mathcal{V}(x)} \inf f(V) = \sigma. \quad (1.47)$$

Since  $\inf f(V_n) \uparrow \sigma$ , it suffices to provide a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{X}$  such that  $x_n \rightarrow x$  and  $\liminf f(x_n) \leq \sigma$ .

If  $\sigma = +\infty$ , then (1.47) implies that  $f(x_n) \rightarrow \sigma$  for every sequence  $(x_n)_{n \in \mathbb{N}}$  that converges to  $x$ .

Now assume that  $\sigma \in \mathbb{R}$  and that  $(x_n)_{n \in \mathbb{N}}$  is a sequence such that  $(\forall n \in \mathbb{N}) x_n \in V_n$  and  $f(x_n) \leq \varepsilon_n + \inf f(V_n)$ . Then  $\liminf f(x_n) \leq \overline{\lim} f(x_n) \leq \overline{\lim} \varepsilon_n + \liminf f(V_n) = \sigma$ .

Finally, assume that  $\sigma = -\infty$ . Then, for every  $n \in \mathbb{N}$ ,  $\inf f(V_n) = -\infty$ , and we take  $x_n \in V_n$  such that  $f(x_n) \leq -n$ . Then  $\liminf f(x_n) = -\infty = \sigma$ , and the proof is complete.  $\square$

A sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{X}$  is a *Cauchy sequence* if  $d(x_m, x_n) \rightarrow 0$  as  $m, n \rightarrow +\infty$ . The metric space  $\mathcal{X}$  is *complete* if every Cauchy sequence in  $\mathcal{X}$  converges to a point in  $\mathcal{X}$ .

**Lemma 1.42 (Cantor)** *Let  $\mathcal{X}$  be a complete metric space and let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of nonempty closed sets in  $\mathcal{X}$  such that  $(\forall n \in \mathbb{N}) C_n \supset C_{n+1}$  and  $\text{diam } C_n \rightarrow 0$ . Then  $\bigcap_{n \in \mathbb{N}} C_n$  is a singleton.*

*Proof.* Set  $C = \bigcap_{n \in \mathbb{N}} C_n$ . For every  $n \in \mathbb{N}$ , fix  $x_n \in C_n$  and set  $A_n = \{x_m\}_{m \geq n} \subset C_n$ . Then  $\text{diam } A_n \rightarrow 0$ , i.e.,  $(x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence. By completeness, there exists  $x \in \mathcal{X}$  such that  $x_n \rightarrow x$ . Hence, for every  $n \in \mathbb{N}$ ,  $C_n \ni x_{n+p} \rightarrow x$  as  $p \rightarrow +\infty$  and, by closedness of  $C_n$ , we get  $x \in C_n$ . Thus,  $x \in C$  and  $\text{diam } C \leq \text{diam } C_n \rightarrow 0$ . Altogether,  $C = \{x\}$ .  $\square$

**Lemma 1.43 (Ursescu)** *Let  $\mathcal{X}$  be a complete metric space. Then the following hold:*

- (i) *Suppose that  $(C_n)_{n \in \mathbb{N}}$  is a sequence of closed subsets of  $\mathcal{X}$ . Then  $\overline{\bigcup_{n \in \mathbb{N}} \text{int } C_n} = \text{int } \bigcup_{n \in \mathbb{N}} C_n$ .*

- (ii) Suppose that  $(C_n)_{n \in \mathbb{N}}$  is a sequence of open subsets of  $\mathcal{X}$ . Then  $\text{int } \bigcap_{n \in \mathbb{N}} \overline{C_n} = \text{int } \bigcap_{n \in \mathbb{N}} C_n$ .

*Proof.* For any subset  $C$  of  $\mathcal{X}$ , one has  $\mathcal{X} \setminus \overline{C} = \text{int}(\mathcal{X} \setminus C)$  and  $\mathcal{X} \setminus \text{int } C = \overline{\mathcal{X} \setminus C}$ . This and De Morgan's laws imply that (i) and (ii) are equivalent.

(ii): The inclusion  $\text{int } \bigcap_{n \in \mathbb{N}} \overline{C_n} \subset \text{int } \bigcap_{n \in \mathbb{N}} C_n$  is clear. To show the reverse inclusion, let us fix

$$z \in \text{int } \bigcap_{n \in \mathbb{N}} \overline{C_n} \quad \text{and} \quad \varepsilon \in \mathbb{R}_{++}. \quad (1.48)$$

Using strong (also known as complete) induction, we shall construct sequences  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{X}$  and  $(\varepsilon_n)_{n \in \mathbb{N}}$  in  $\mathbb{R}_{++}$  such that  $x_0 = z$  and, for every  $n \in \mathbb{N}$ ,

$$B(x_{n+1}; \varepsilon_{n+1}) \subset B(x_n; \varepsilon_n) \cap C_n \quad \text{with} \quad \varepsilon_n \in ]0, \varepsilon/2^n[. \quad (1.49)$$

For every  $n \in \mathbb{N}$ , let us denote by  $U_n$  the open ball of center  $x_n$  and radius  $\varepsilon_n$ . First, set  $x_0 = z$  and let  $\varepsilon_0 \in ]0, \varepsilon[$  be such that

$$B(x_0; \varepsilon_0) \subset \bigcap_{n \in \mathbb{N}} \overline{C_n}. \quad (1.50)$$

Since  $x_0 \in \overline{C_0}$ , the set  $U_0 \cap C_0$  is nonempty and open. Thus, there exist  $x_1 \in \mathcal{X}$  and  $\varepsilon_1 \in ]0, \varepsilon/2[$  such that

$$B(x_1; \varepsilon_1) \subset U_0 \cap C_0 \subset B(x_0; \varepsilon_0) \cap C_0. \quad (1.51)$$

Now assume that  $(x_k)_{0 \leq k \leq n}$  and  $(\varepsilon_k)_{0 \leq k \leq n}$  are already constructed. Then, using (1.50), we obtain

$$x_n \in U_n \subset B(x_n; \varepsilon_n) \subset B(x_0; \varepsilon_0) \subset \overline{C_n}. \quad (1.52)$$

Hence, there exists  $x_{n+1} \in C_n$  such that  $d(x_n, x_{n+1}) < \varepsilon_n/2$ . Moreover,  $x_{n+1}$  belongs to the open set  $U_n \cap C_n$ . As required, there exists  $\varepsilon_{n+1} \in ]0, \varepsilon_n/2[ \subset ]0, \varepsilon/2^{n+1}[$  such that

$$B(x_{n+1}; \varepsilon_{n+1}) \subset U_n \cap C_n \subset B(x_n; \varepsilon_n) \cap C_n. \quad (1.53)$$

Since the sequence  $(B(x_n; \varepsilon_n))_{n \in \mathbb{N}}$  is decreasing and  $\varepsilon_n \rightarrow 0$ , we have  $\text{diam } B(x_n; \varepsilon_n) = 2\varepsilon_n \rightarrow 0$ . Therefore, Lemma 1.42 yields a point  $z_\varepsilon \in \mathcal{X}$  such that

$$\bigcap_{n \in \mathbb{N}} B(x_n; \varepsilon_n) = \{z_\varepsilon\}. \quad (1.54)$$

Combining (1.49) and (1.54), we deduce that  $z_\varepsilon \in B(z; \varepsilon) \cap \bigcap_{n \in \mathbb{N}} C_n$ . Letting  $\varepsilon \downarrow 0$ , we conclude that  $z \in \overline{\bigcap_{n \in \mathbb{N}} C_n}$ .  $\square$

A countable intersection of open sets in a Hausdorff space is called a  $G_\delta$  set.

**Corollary 1.44** *Let  $\mathcal{X}$  be a complete metric space, and let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of dense open subsets of  $\mathcal{X}$ . Then  $\bigcap_{n \in \mathbb{N}} C_n$  is a dense  $G_\delta$  subset of  $\mathcal{X}$ .*

*Proof.* It is clear that  $C = \bigcap_{n \in \mathbb{N}} C_n$  is a  $G_\delta$ . Using Lemma 1.43(ii), we obtain  $\mathcal{X} = \text{int} \bigcap_{n \in \mathbb{N}} \overline{C_n} = \text{int} \overline{\bigcap_{n \in \mathbb{N}} C_n} = \text{int} \overline{C} \subset \overline{C} \subset \mathcal{X}$ . Hence  $\overline{C} = \mathcal{X}$ .  $\square$

**Theorem 1.45 (Ekeland)** *Let  $(\mathcal{X}, d)$  be a complete metric space, let  $f: \mathcal{X} \rightarrow ]-\infty, +\infty]$  be proper, lower semicontinuous, and bounded below, let  $\alpha \in \mathbb{R}_{++}$ , let  $\beta \in \mathbb{R}_{++}$ , and suppose that  $y \in \text{dom } f$  satisfies  $f(y) \leq \alpha + \inf f(\mathcal{X})$ . Then there exists  $z \in \mathcal{X}$  such that the following hold:*

- (i)  $f(z) + (\alpha/\beta)d(y, z) \leq f(y)$ .
- (ii)  $d(y, z) \leq \beta$ .
- (iii)  $(\forall x \in \mathcal{X} \setminus \{z\}) f(z) < f(x) + (\alpha/\beta)d(x, z)$ .

*Proof.* We fix  $x_0 \in \mathcal{X}$  and define inductively sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(C_n)_{n \in \mathbb{N}}$  as follows. Given  $x_n \in \mathcal{X}$ , where  $n \in \mathbb{N}$ , set

$$C_n = \{x \in \mathcal{X} \mid f(x) + (\alpha/\beta)d(x_n, x) \leq f(x_n)\} \quad (1.55)$$

and take  $x_{n+1} \in \mathcal{X}$  such that

$$x_{n+1} \in C_n \quad \text{and} \quad f(x_{n+1}) \leq \frac{1}{2}f(x_n) + \frac{1}{2}\inf f(C_n). \quad (1.56)$$

Since  $x_{n+1} \in C_n$ , we have  $(\alpha/\beta)d(x_n, x_{n+1}) \leq f(x_n) - f(x_{n+1})$ . Thus,

$$(f(x_n))_{n \in \mathbb{N}} \text{ is decreasing and bounded below, hence convergent,} \quad (1.57)$$

and

$$(\forall n \in \mathbb{N})(\forall m \in \mathbb{N}) \quad n \leq m \Rightarrow (\alpha/\beta)d(x_n, x_m) \leq f(x_n) - f(x_m). \quad (1.58)$$

Combining (1.57) and (1.58), we see that  $(x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence. Set  $z = \lim x_n$ . Since  $f$  is lower semicontinuous at  $z$ , it follows from (1.57) that

$$f(z) \leq \lim f(x_n). \quad (1.59)$$

Letting  $m \rightarrow +\infty$  in (1.58), we deduce that

$$(\forall n \in \mathbb{N}) \quad (\alpha/\beta)d(x_n, z) \leq f(x_n) - f(z). \quad (1.60)$$

Recalling that  $x_0 = y$  and setting  $n = 0$  in (1.60), we obtain (i). In turn, (i) implies that  $f(z) + (\alpha/\beta)d(y, z) \leq f(y) \leq \alpha + \inf f(\mathcal{X}) \leq \alpha + f(z)$ . Thus (ii) holds. Now assume that (iii) is false. Then there exists  $x \in \mathcal{X} \setminus \{z\}$  such that

$$f(x) \leq f(z) - (\alpha/\beta)d(x, z) < f(z). \quad (1.61)$$

In view of (1.60), we get

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad f(x) &\leq f(z) - (\alpha/\beta)d(x, z) \\ &\leq f(x_n) - (\alpha/\beta)(d(x, z) + d(x_n, z)) \\ &\leq f(x_n) - (\alpha/\beta)d(x, x_n). \end{aligned} \quad (1.62)$$

Thus  $x \in \bigcap_{n \in \mathbb{N}} C_n$ . Using (1.56), we deduce that  $(\forall n \in \mathbb{N}) \ 2f(x_{n+1}) - f(x_n) \leq f(x)$ . Hence  $\lim f(x_n) \leq f(x)$ . This, (1.61), and (1.59) imply that  $\lim f(x_n) \leq f(x) < f(z) \leq \lim f(x_n)$ , which is impossible. Therefore, (iii) holds.  $\square$

**Definition 1.46** Let  $(\mathcal{X}_1, d_1)$  and  $(\mathcal{X}_2, d_2)$  be metric spaces, let  $T: \mathcal{X}_1 \rightarrow \mathcal{X}_2$ , and let  $C$  be a subset of  $\mathcal{X}_1$ . Then  $T$  is *Lipschitz continuous* with constant  $\beta \in \mathbb{R}_+$  if

$$(\forall x \in \mathcal{X}_1)(\forall y \in \mathcal{X}_1) \quad d_2(Tx, Ty) \leq \beta d_1(x, y), \quad (1.63)$$

*locally Lipschitz continuous* near a point  $x \in \mathcal{X}_1$  if there exists  $\rho \in \mathbb{R}_{++}$  such that the operator  $T|_{B(x; \rho)}$  is Lipschitz continuous, and locally Lipschitz continuous on  $C$  if it is locally Lipschitz continuous near every point in  $C$ . Finally,  $T$  is *Lipschitz continuous relative to  $C$*  with constant  $\beta \in \mathbb{R}_+$  if

$$(\forall x \in C)(\forall y \in C) \quad d_2(Tx, Ty) \leq \beta d_1(x, y). \quad (1.64)$$

**Example 1.47** Let  $C$  be a nonempty subset of a metric space  $(\mathcal{X}, d)$ . Then

$$(\forall x \in \mathcal{X})(\forall y \in \mathcal{X}) \quad |d_C(x) - d_C(y)| \leq d(x, y). \quad (1.65)$$

*Proof.* Take  $x$  and  $y$  in  $\mathcal{X}$ . Then  $(\forall z \in \mathcal{X}) \ d(x, z) \leq d(x, y) + d(y, z)$ . Taking the infimum over  $z \in C$  yields  $d_C(x) \leq d(x, y) + d_C(y)$ , hence  $d_C(x) - d_C(y) \leq d(x, y)$ . Interchanging  $x$  and  $y$ , we obtain  $d_C(y) - d_C(x) \leq d(x, y)$ . Altogether,  $|d_C(x) - d_C(y)| \leq d(x, y)$ .  $\square$

The following result is known as the Banach–Picard fixed point theorem. The set of fixed points of an operator  $T: \mathcal{X} \rightarrow \mathcal{X}$  is denoted by  $\text{Fix } T$ , i.e.,

$$\text{Fix } T = \{x \in \mathcal{X} \mid Tx = x\}. \quad (1.66)$$

**Theorem 1.48 (Banach–Picard)** *Let  $(\mathcal{X}, d)$  be a complete metric space and let  $T: \mathcal{X} \rightarrow \mathcal{X}$  be Lipschitz continuous with constant  $\beta \in [0, 1[$ . Given  $x_0 \in \mathcal{X}$ , set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Tx_n. \quad (1.67)$$

*Then there exists  $x \in \mathcal{X}$  such that the following hold:*

- (i)  $x$  is the unique fixed point of  $T$ .
- (ii)  $(\forall n \in \mathbb{N}) \ d(x_{n+1}, x) \leq \beta d(x_n, x)$ .

- (iii)  $(\forall n \in \mathbb{N}) \ d(x_n, x) \leq \beta^n d(x_0, x)$  (hence  $(x_n)_{n \in \mathbb{N}}$  converges linearly to  $x$ ).
- (iv) *A priori error estimate:*  $(\forall n \in \mathbb{N}) \ d(x_n, x) \leq \beta^n d(x_0, x_1)/(1 - \beta)$ .
- (v) *A posteriori error estimate:*  $(\forall n \in \mathbb{N}) \ d(x_n, x) \leq d(x_n, x_{n+1})/(1 - \beta)$ .
- (vi)  $d(x_0, x_1)/(1 + \beta) \leq d(x_0, x) \leq d(x_0, x_1)/(1 - \beta)$ .

*Proof.* The triangle inequality and (1.67) yield

$$\begin{aligned}
 (\forall m \in \mathbb{N})(\forall n \in \mathbb{N}) \quad d(x_n, x_{n+m}) &\leq d(x_n, x_{n+1}) + \cdots + d(x_{n+m-1}, x_{n+m}) \\
 &\leq (1 + \beta + \cdots + \beta^{m-1})d(x_n, x_{n+1}) \\
 &= \frac{1 - \beta^m}{1 - \beta} d(x_n, x_{n+1}) \tag{1.68}
 \end{aligned}$$

$$\leq \frac{\beta^n}{1 - \beta} d(x_0, x_1). \tag{1.69}$$

It follows that  $(x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence in  $\mathcal{X}$  and therefore that it converges to some point  $x \in \mathcal{X}$ . In turn, since  $T$  is continuous,  $Tx = T(\lim x_n) = \lim Tx_n = \lim x_{n+1} = x$  and thus  $x \in \text{Fix } T$ . Now let  $y \in \text{Fix } T$ . Then  $d(x, y) = d(Tx, Ty) \leq \beta d(x, y)$  and hence  $y = x$ . This establishes (i).

(ii): Observe that  $(\forall n \in \mathbb{N}) \ d(x_{n+1}, x) = d(Tx_n, Tx) \leq \beta d(x_n, x)$ .

(ii) $\Rightarrow$ (iii): Clear.

(iv)&(v): Let  $m \rightarrow +\infty$  in (1.69) and (1.68), respectively.

(vi): Since  $d(x_0, x_1) \leq d(x_0, x) + d(x, x_1) \leq (1 + \beta)d(x_0, x)$ , the first inequality holds. The second inequality follows from (iv) or (v).  $\square$

We close this chapter with a variant of the Banach–Picard theorem.

**Theorem 1.49** *Let  $(\mathcal{X}, d)$  be a complete metric space and let  $T: \mathcal{X} \rightarrow \mathcal{X}$  be such that there exists a summable sequence  $(\beta_n)_{n \in \mathbb{N}}$  in  $\mathbb{R}_+$  such that*

$$(\forall x \in \mathcal{X})(\forall y \in \mathcal{X})(\forall n \in \mathbb{N}) \quad d(T^n x, T^n y) \leq \beta_n d(x, y), \tag{1.70}$$

where  $T^n$  denotes the  $n$ -fold composition of  $T$  if  $n > 0$ , and  $T^0 = \text{Id}$ . Let  $x_0 \in \mathcal{X}$  and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Tx_n \quad \text{and} \quad \alpha_n = \sum_{k=n}^{+\infty} \beta_k. \tag{1.71}$$

Then there exists  $x \in \mathcal{X}$  such that the following hold:

- (i)  $x$  is the unique fixed point of  $T$ .
- (ii)  $x_n \rightarrow x$ .
- (iii)  $(\forall n \in \mathbb{N}) \ d(x_n, x) \leq \alpha_n d(x_0, x_1)$ .

*Proof.* We deduce from (1.71) that

$$(\forall m \in \mathbb{N})(\forall n \in \mathbb{N}) \quad d(x_n, x_{n+m}) \leq \sum_{k=n}^{n+m-1} d(x_k, x_{k+1})$$

$$\begin{aligned}
&= \sum_{k=n}^{n+m-1} d(T^k x_0, T^k x_1) \\
&\leq \sum_{k=n}^{n+m-1} \beta_k d(x_0, x_1) \\
&\leq \alpha_n d(x_0, x_1). \tag{1.72}
\end{aligned}$$

(i)&(ii): Since  $(\beta_n)_{n \in \mathbb{N}}$  is summable, we have  $\alpha_n \rightarrow 0$ . Thus, (1.72) implies that  $(x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence. Thus, it converges to some point  $x \in \mathcal{X}$ . It follows from the continuity of  $T$  that  $Tx = T(\lim x_n) = \lim Tx_n = \lim x_{n+1} = x$  and therefore that  $x \in \text{Fix } T$ . Now let  $y \in \text{Fix } T$ . Then (1.70) yields  $(\forall n \in \mathbb{N}) \ d(x, y) = d(T^n x, T^n y) \leq \beta_n d(x, y)$ . Since  $\beta_n \rightarrow 0$ , we conclude that  $y = x$ .

(iii): Let  $m \rightarrow +\infty$  in (1.72). □

## Exercises

**Exercise 1.1** Let  $\mathcal{X}$  and  $\mathcal{Y}$  be two nonempty sets and let  $T: \mathcal{X} \rightarrow \mathcal{Y}$ .

- (i) Let  $C \subset \mathcal{X}$  and  $D \subset \mathcal{Y}$ . Show that  $C \subset T^{-1}(T(C))$  and  $T(T^{-1}(D)) \subset D$ . Provide an example of strict inclusion in both cases.
- (ii) Let  $D \subset \mathcal{Y}$  and let  $(D_i)_{i \in I}$  be a family of subsets of  $\mathcal{Y}$ . Show the following:
  - (a)  $T^{-1}(\mathcal{Y} \setminus D) = \mathcal{X} \setminus T^{-1}(D)$ .
  - (b)  $T^{-1}(\bigcap_{i \in I} D_i) = \bigcap_{i \in I} T^{-1}(D_i)$ .
  - (c)  $T^{-1}(\bigcup_{i \in I} D_i) = \bigcup_{i \in I} T^{-1}(D_i)$ .

(iii) Prove Fact 1.18.

**Exercise 1.2** Let  $C$  and  $D$  be two arbitrary subsets of a topological space  $\mathcal{X}$ . Show the following:

- (i)  $\mathcal{X} \setminus \text{int } C = \overline{\mathcal{X} \setminus C}$  and  $\mathcal{X} \setminus \overline{C} = \text{int}(\mathcal{X} \setminus C)$ .
- (ii)  $\text{int}(C \cap D) = (\text{int } C) \cap (\text{int } D)$  and  $\overline{C \cup D} = \overline{C} \cup \overline{D}$ .
- (iii)  $\text{int}(C \cup D) \neq (\text{int } C) \cup (\text{int } D)$  and  $\overline{C \cap D} \neq \overline{C} \cap \overline{D}$ .

**Exercise 1.3** Let  $A = \mathbb{Z}$  be directed by  $\leq$  and define a net  $(x_a)_{a \in A}$  in  $\mathbb{R}$  by

$$(\forall a \in A) \quad x_a = \begin{cases} a, & \text{if } a \leq 0; \\ 1/a, & \text{if } a > 0. \end{cases} \tag{1.73}$$

Show that  $(x_a)_{a \in A}$  is unbounded and that it converges.

**Exercise 1.4** In this exercise,  $\mathbb{N}$  designates the unordered and undirected set of positive integers. Let  $A = \mathbb{N}$  be directed by  $\leq$ , and let  $B = \mathbb{N}$  be directed



by the relation  $\preccurlyeq$  that coincides with  $\leq$  on  $\{2n\}_{n \in \mathbb{N}}$  and on  $\{2n+1\}_{n \in \mathbb{N}}$ , and satisfies  $(\forall m \in \mathbb{N})(\forall n \in \mathbb{N}) 2m+1 \preccurlyeq 2n$ . Now let  $(x_a)_{a \in A}$  be a sequence in  $\mathbb{R}$ .

- (i) Show that  $(\forall a \in A)(\forall b \in B) b \succcurlyeq 2a \Rightarrow 3b \geq a$ .
- (ii) Show that  $(x_{3b})_{b \in B}$  is a subnet, but not a subsequence, of  $(x_a)_{a \in A}$ .
- (iii) Set  $(\forall a \in A) x_a = (1 - (-1)^a)a/2$ . Show that the subnet  $(x_{3b})_{b \in B}$  converges to 0 but that the subsequence  $(x_{3a})_{a \in A}$  does not converge.

**Exercise 1.5** Let  $\mathcal{X}$  be a nonempty set and let  $f: \mathcal{X} \rightarrow \mathbb{R}$ . Equip  $\mathcal{X}$  with a topology  $\mathsf{T}$  and  $\mathbb{R}$  with the usual topology. Provide a general condition for  $f$  to be continuous in the following cases:

- (i)  $\mathsf{T} = \{\emptyset, \mathcal{X}\}$ .
- (ii)  $\mathsf{T} = 2^{\mathcal{X}}$ .

**Exercise 1.6** Let  $f: \mathbb{R} \rightarrow [-\infty, +\infty]$  and set

$$(\forall z \in \mathbb{R}) \quad f_z: \mathbb{R} \rightarrow [-\infty, +\infty] : x \mapsto \begin{cases} f(z), & \text{if } x = z; \\ +\infty, & \text{otherwise.} \end{cases} \quad (1.74)$$

Show that the functions  $(f_z)_{z \in \mathbb{R}}$  are lower semicontinuous and conclude that the infimum of an infinite family of lower semicontinuous functions need not be lower semicontinuous (compare with Lemma 1.26).

**Exercise 1.7** Let  $f_1$  and  $f_2$  be functions from  $\mathbb{R}$  to  $\mathbb{R}$ . Show the following:

- (i) If  $f_1$  is lower semicontinuous and  $f_2$  is continuous, then  $f_1 \circ f_2$  is lower semicontinuous.
- (ii) If  $f_1$  is continuous and  $f_2$  is lower semicontinuous, then  $f_1 \circ f_2$  may fail to be lower semicontinuous.

**Exercise 1.8** Let  $\mathcal{X}$  be a Hausdorff space, let  $f: \mathcal{X} \rightarrow [-\infty, +\infty]$ , and let  $g: \mathcal{X} \rightarrow \mathbb{R}$  be continuous. Show that  $\overline{f+g} = \bar{f} + g$ .

**Exercise 1.9** Let  $\mathcal{X}$  be a topological space. Then  $\mathcal{X}$  is *first countable* if, for every  $x \in \mathcal{X}$ , there exists a countable base of neighborhoods, i.e., a family  $\{V_n\}_{n \in \mathbb{N}} \subset \mathcal{V}(x)$  such that  $(\forall V \in \mathcal{V}(x))(\exists n \in \mathbb{N}) V_n \subset V$ .

- (i) Show that, without loss of generality, the sequence of sets  $(V_n)_{n \in \mathbb{N}}$  can be taken to be decreasing in the above definition.
- (ii) Show that if  $\mathcal{X}$  is metrizable, then it is first countable.
- (iii) The space  $\mathcal{X}$  is a *Fréchet space* if, for every subset  $C$  of  $\mathcal{X}$  and every  $x \in \overline{C}$ , there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $C$  such that  $x_n \rightarrow x$ . Show the following:
  - (a) If  $\mathcal{X}$  is first countable, then it is Fréchet.
  - (b) If  $\mathcal{X}$  is Fréchet, then it is sequential.

- (iv) Conclude that the following relations hold: metrizable  $\Rightarrow$  first countable  $\Rightarrow$  Fréchet  $\Rightarrow$  sequential.

**Exercise 1.10** Let  $C$  be a subset of a metric space  $\mathcal{X}$ . Show the following:

- (i)  $\overline{C} = \{x \in \mathcal{X} \mid d_C(x) = 0\}$ .
- (ii)  $\overline{C}$  is a  $G_\delta$ .

**Exercise 1.11** A subset of a Hausdorff space  $\mathcal{X}$  is an  $F_\sigma$  if it is a countable union of closed sets. Show the following:

- (i) The complement of an  $F_\sigma$  is a  $G_\delta$ .
- (ii) If  $\mathcal{X}$  is a metric space, every open set in  $\mathcal{X}$  is an  $F_\sigma$ .

**Exercise 1.12** Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in a Hausdorff space  $\mathcal{X}$  that converges to some point  $x \in \mathcal{X}$ . Show that the set  $\{x_n\}_{n \in \mathbb{N}} \cup \{x\}$  is compact.

**Exercise 1.13** In the compact metric space  $\mathcal{X} = [0, 1]$ , direct the set

$$A = \{a \in ]0, 1[ \mid a \text{ is a rational number}\} \quad (1.75)$$

by  $\leq$  and define  $(\forall a \in A) x_a = a$ . Show that the net  $(x_a)_{a \in A}$  converges to 1, while the set  $\{x_a\}_{a \in A} \cup \{1\}$  is not closed. Compare with Exercise 1.12.

**Exercise 1.14** Find a metric space  $(\mathcal{X}, d)$  such that  $(\forall x \in \mathcal{X}) \text{int } B(x; 1) \neq \{y \in \mathcal{X} \mid d(x, y) < 1\}$ .

**Exercise 1.15** Let  $(x_n)_{n \in \mathbb{N}}$  be a Cauchy sequence in a metric space  $\mathcal{X}$ . Show that  $(x_n)_{n \in \mathbb{N}}$  is bounded and that, if it possesses a sequential cluster point, it converges to that point.

**Exercise 1.16** Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in a complete metric space  $(\mathcal{X}, d)$  such that  $\sum_{n \in \mathbb{N}} d(x_n, x_{n+1}) < +\infty$ . Show that  $(x_n)_{n \in \mathbb{N}}$  converges and that this is no longer true if we merely assume that  $\sum_{n \in \mathbb{N}} d^2(x_n, x_{n+1}) < +\infty$ .

**Exercise 1.17** Show that if in Lemma 1.43 the sequence  $(C_n)_{n \in \mathbb{N}}$  is replaced by an arbitrary family  $(C_a)_{a \in A}$ , then the conclusion fails.

**Exercise 1.18** Provide an example of a function  $f$  from  $\mathbb{R}$  to  $\mathbb{R}$  such that  $f$  is locally Lipschitz continuous on  $\mathbb{R}$  but  $f$  is not Lipschitz continuous.

**Exercise 1.19** Let  $(\mathcal{X}, d)$  be a metric space and let  $(x_n)_{n \in \mathbb{N}}$  be a Cauchy sequence in  $\mathcal{X}$ . Show that for every  $x \in \mathcal{X}$ ,  $(d(x, x_n))_{n \in \mathbb{N}}$  converges, that  $f: \mathcal{X} \rightarrow \mathbb{R}: x \mapsto \lim d(x, x_n)$  is Lipschitz continuous with constant 1, and that  $f(x_n) \rightarrow 0$ .

**Exercise 1.20** Let  $(\mathcal{X}, d)$  be a metric space. Use Theorem 1.45 and Exercise 1.19 to show that  $\mathcal{X}$  is complete if and only if, for every Lipschitz continuous function  $f: \mathcal{X} \rightarrow \mathbb{R}_+$  and for every  $\varepsilon \in ]0, 1[$ , there exists  $z \in \mathcal{X}$  such that  $(\forall x \in \mathcal{X}) f(z) \leq f(x) + \varepsilon d(x, z)$ .

**Exercise 1.21 (Hausdorff)** Let  $(\mathcal{X}, d)$  be a metric space and let  $\mathcal{C}$  be the class of nonempty bounded closed subsets of  $\mathcal{X}$ . Define the Hausdorff distance by

$$H: \mathcal{C} \times \mathcal{C} \rightarrow \mathbb{R}_+: (C, D) \mapsto \max \left\{ \sup_{x \in C} d_D(x), \sup_{x \in D} d_C(x) \right\}. \quad (1.76)$$

The purpose of this exercise is to show that  $H$  is indeed a distance on  $\mathcal{C}$ .

- (i) Let  $x \in \mathcal{X}$  and  $y \in \mathcal{Y}$ . Compute  $H(\{x\}, \{y\})$ .
- (ii) Show that  $(\forall C \in \mathcal{C})(\forall D \in \mathcal{C}) H(C, D) = H(D, C)$ .
- (iii) Show that  $(\forall C \in \mathcal{C})(\forall D \in \mathcal{C}) H(C, D) = 0 \Leftrightarrow C = D$ .
- (iv) Show that  $(\forall C \in \mathcal{C})(\forall D \in \mathcal{C})(\forall x \in \mathcal{X}) d_D(x) \leq d_C(x) + H(C, D)$ .
- (v) Show that  $(\mathcal{C}, H)$  is a metric space.

**Exercise 1.22** Use Theorem 1.45 to prove Theorem 1.48(i).

**Exercise 1.23** Let  $\mathcal{X}$  be a complete metric space, suppose that  $(C_i)_{i \in I}$  is a family of subsets of  $\mathcal{X}$  that are open and dense in  $\mathcal{X}$ , and set  $C = \bigcap_{i \in I} C_i$ . Show the following:

- (i) If  $I$  is finite, then  $C$  is open and dense in  $\mathcal{X}$ .
- (ii) If  $I$  is countably infinite, then  $C$  is dense (hence nonempty), but  $C$  may fail to be open.
- (iii) If  $I$  is uncountable, then  $C$  may be empty.



# Chapter 2

## Hilbert Spaces

Throughout this book,  $\mathcal{H}$  is a real Hilbert space with scalar (or inner) product  $\langle \cdot | \cdot \rangle$ . The associated norm is denoted by  $\| \cdot \|$  and the associated distance by  $d$ , i.e.,

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad \|x\| = \sqrt{\langle x | x \rangle} \quad \text{and} \quad d(x, y) = \|x - y\|. \quad (2.1)$$

The identity operator on  $\mathcal{H}$  is denoted by  $\text{Id}$ .

In this chapter, we derive useful identities and inequalities, and we review examples and basic results from linear and nonlinear analysis in a Hilbert space setting.

### 2.1 Notation and Examples

The *orthogonal complement* of a subset  $C$  of  $\mathcal{H}$  is denoted by  $C^\perp$ , i.e.,

$$C^\perp = \{u \in \mathcal{H} \mid (\forall x \in C) \quad \langle x | u \rangle = 0\}. \quad (2.2)$$

An orthonormal set  $C \subset \mathcal{H}$  is an *orthonormal basis* of  $\mathcal{H}$  if  $\overline{\text{span}} C = \mathcal{H}$ . The space  $\mathcal{H}$  is *separable* if it possesses a countable orthonormal basis. Now let  $(x_i)_{i \in I}$  be a family of vectors in  $\mathcal{H}$  and let  $\mathcal{I}$  be the class of nonempty finite subsets of  $I$ , directed by  $\subset$ . Then  $(x_i)_{i \in I}$  is *summable* if there exists  $x \in \mathcal{H}$  such that the net  $(\sum_{i \in J} x_i)_{J \in \mathcal{I}}$  converges to  $x$ , i.e., by (1.25),

$$(\forall \varepsilon \in \mathbb{R}_{++})(\exists K \in \mathcal{I})(\forall J \in \mathcal{I}) \quad J \supset K \Rightarrow \left\| x - \sum_{i \in J} x_i \right\| \leq \varepsilon. \quad (2.3)$$

In this case we write  $x = \sum_{i \in I} x_i$ . For a family  $(\alpha_i)_{i \in I}$  in  $[0, +\infty]$ , we have

$$\sum_{i \in I} \alpha_i = \sup_{J \in \mathcal{I}} \sum_{i \in J} \alpha_i. \quad (2.4)$$

Here are specific real Hilbert spaces that will be used in this book.

**Example 2.1** The *Hilbert direct sum* of a totally ordered family of real Hilbert spaces  $(\mathcal{H}_i, \|\cdot\|_i)_{i \in I}$  is the real Hilbert space

$$\bigoplus_{i \in I} \mathcal{H}_i = \left\{ \mathbf{x} = (x_i)_{i \in I} \in \prod_{i \in I} \mathcal{H}_i \mid \sum_{i \in I} \|x_i\|_i^2 < +\infty \right\} \quad (2.5)$$

equipped with the addition  $(\mathbf{x}, \mathbf{y}) \mapsto (x_i + y_i)_{i \in I}$ , the scalar multiplication  $(\alpha, \mathbf{x}) \mapsto (\alpha x_i)_{i \in I}$ , and the scalar product

$$(\mathbf{x}, \mathbf{y}) \mapsto \sum_{i \in I} \langle x_i \mid y_i \rangle_i \quad (2.6)$$

(when  $I$  is finite, we shall sometimes adopt a common abuse of notation and write  $\prod_{i \in I} \mathcal{H}_i$  instead of  $\bigoplus_{i \in I} \mathcal{H}_i$ ). Now suppose that, for every  $i \in I$ ,  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$ , and that if  $I$  is infinite,  $\inf_{i \in I} f_i \geq 0$ . Then

$$\bigoplus_{i \in I} f_i: \bigoplus_{i \in I} \mathcal{H}_i \rightarrow ]-\infty, +\infty]: (x_i)_{i \in I} \mapsto \sum_{i \in I} f_i(x_i). \quad (2.7)$$

**Example 2.2** If each  $\mathcal{H}_i$  is the Euclidean line  $\mathbb{R}$  in Example 2.1, we obtain  $\ell^2(I) = \bigoplus_{i \in I} \mathbb{R}$ , which is equipped with the scalar product  $(x, y) = ((\xi_i)_{i \in I}, (\eta_i)_{i \in I}) \mapsto \sum_{i \in I} \xi_i \eta_i$ . The standard unit vectors  $(e_i)_{i \in I}$  of  $\ell^2(I)$  are defined by

$$(\forall i \in I) \quad e_i: I \rightarrow \mathbb{R}: j \mapsto \begin{cases} 1, & \text{if } j = i; \\ 0, & \text{otherwise.} \end{cases} \quad (2.8)$$

**Example 2.3** If  $I = \{1, \dots, N\}$  in Example 2.2, we obtain the standard Euclidean space  $\mathbb{R}^N$ .

**Example 2.4** The space of real  $N \times N$  symmetric matrices is denoted by  $\mathbb{S}^N$ . It is a real Hilbert space with scalar product  $(A, B) \mapsto \text{tr}(AB)$ , where  $\text{tr}$  is the trace function.

**Example 2.5** Let  $(\Omega, \mathcal{F}, \mu)$  be a (positive) measure space, let  $(\mathbf{H}, \langle \cdot \mid \cdot \rangle_{\mathbf{H}})$  be a real Hilbert space, and let  $p \in [1, +\infty[$ . Denote by  $L^p((\Omega, \mathcal{F}, \mu); \mathbf{H})$  the space of (equivalence classes of) Borel measurable functions  $x: \Omega \rightarrow \mathbf{H}$  such that  $\int_{\Omega} \|x(\omega)\|_{\mathbf{H}}^p \mu(d\omega) < +\infty$ . Then  $L^2((\Omega, \mathcal{F}, \mu); \mathbf{H})$  is a real Hilbert space with scalar product  $(x, y) \mapsto \int_{\Omega} \langle x(\omega) \mid y(\omega) \rangle_{\mathbf{H}} \mu(d\omega)$ .

**Example 2.6** In Example 2.5, let  $\mathbf{H} = \mathbb{R}$ . Then we obtain the real Banach space  $L^p(\Omega, \mathcal{F}, \mu) = L^p((\Omega, \mathcal{F}, \mu); \mathbb{R})$  and, for  $p = 2$ , the real Hilbert space  $L^2(\Omega, \mathcal{F}, \mu)$ , which is equipped with the scalar product  $(x, y) \mapsto \int_{\Omega} x(\omega)y(\omega)\mu(d\omega)$ .

**Example 2.7** In Example 2.5, let  $T \in \mathbb{R}_{++}$ , set  $\Omega = [0, T]$ , and let  $\mu$  be the Lebesgue measure. Then we obtain the Hilbert space  $L^2([0, T]; \mathbf{H})$ , which is equipped with the scalar product  $(x, y) \mapsto \int_0^T \langle x(t) | y(t) \rangle_{\mathbf{H}} dt$ . In particular, when  $\mathbf{H} = \mathbb{R}$ , we obtain the classical Lebesgue space  $L^2([0, T]) = L^2([0, T]; \mathbb{R})$ .

**Example 2.8** Let  $(\Omega, \mathcal{F}, \mathbf{P})$  be a *probability space*, i.e., a measure space such that  $\mathbf{P}(\Omega) = 1$ . A property that holds  $\mathbf{P}$ -almost everywhere on  $\Omega$  is said to hold *almost surely* (a.s.). A *random variable* (r.v.) is a measurable function  $X: \Omega \rightarrow \mathbb{R}$ , and its expected value is  $\mathbf{E}X = \int_{\Omega} X(\omega) \mathbf{P}(d\omega)$ , provided that the integral exists. In this context, Example 2.6 yields the Hilbert space

$$L^2(\Omega, \mathcal{F}, \mathbf{P}) = \{X \text{ r.v. on } (\Omega, \mathcal{F}) \mid \mathbf{E}|X|^2 < +\infty\} \quad (2.9)$$

of random variables with finite second absolute moment, which is equipped with the scalar product  $(X, Y) \mapsto \mathbf{E}(XY)$ .

**Example 2.9** Let  $T \in \mathbb{R}_{++}$  and let  $(\mathbf{H}, \langle \cdot | \cdot \rangle_{\mathbf{H}})$  be a real Hilbert space. For every  $y \in L^2([0, T]; \mathbf{H})$ , the function  $x: [0, T] \rightarrow \mathbf{H}: t \mapsto \int_0^t y(s) ds$  is differentiable almost everywhere (a.e.) on  $]0, T[$  with  $x'(t) = y(t)$  a.e. on  $]0, T[$ . We say that  $x: [0, T] \rightarrow \mathbf{H}$  belongs to  $W^{1,2}([0, T]; \mathbf{H})$  if there exists  $y \in L^2([0, T]; \mathbf{H})$  such that

$$(\forall t \in [0, T]) \quad x(t) = x(0) + \int_0^t y(s) ds. \quad (2.10)$$

Alternatively,

$$W^{1,2}([0, T]; \mathbf{H}) = \{x \in L^2([0, T]; \mathbf{H}) \mid x' \in L^2([0, T]; \mathbf{H})\}. \quad (2.11)$$

The scalar product of this real Hilbert space is  $(x, y) \mapsto \int_0^T \langle x(t) | y(t) \rangle_{\mathbf{H}} dt + \int_0^T \langle x'(t) | y'(t) \rangle_{\mathbf{H}} dt$ .

## 2.2 Basic Identities and Inequalities

**Fact 2.10 (Cauchy–Schwarz)** *Let  $x$  and  $y$  be in  $\mathcal{H}$ . Then*

$$|\langle x | y \rangle| \leq \|x\| \|y\|. \quad (2.12)$$

*Moreover,  $\langle x | y \rangle = \|x\| \|y\| \Leftrightarrow (\exists \alpha \in \mathbb{R}_+) \ x = \alpha y$  or  $y = \alpha x$ .*

**Lemma 2.11** *Let  $x, y$ , and  $z$  be in  $\mathcal{H}$ . Then the following hold:*

- (i)  $\|x + y\|^2 = \|x\|^2 + 2\langle x | y \rangle + \|y\|^2$ .
- (ii) *Parallelogram identity:*  $\|x + y\|^2 + \|x - y\|^2 = 2\|x\|^2 + 2\|y\|^2$ .
- (iii) *Polarization identity:*  $4\langle x | y \rangle = \|x + y\|^2 - \|x - y\|^2$ .

(iv) *Apollonius's identity*:  $\|x-y\|^2 = 2\|z-x\|^2 + 2\|z-y\|^2 - 4\|z-(x+y)/2\|^2$ .

*Proof.* (i): A simple expansion.

(ii)&(iii): It follows from (i) that

$$\|x-y\|^2 = \|x\|^2 - 2\langle x | y \rangle + \|y\|^2. \quad (2.13)$$

Adding this identity to (i) yields (ii), and subtracting it from (i) yields (iii).

(iv): Apply (ii) to the points  $(z-x)/2$  and  $(z-y)/2$ .  $\square$

**Lemma 2.12** *Let  $x$  and  $y$  be in  $\mathcal{H}$ . Then the following hold:*

(i)  $\langle x | y \rangle \leq 0 \Leftrightarrow (\forall \alpha \in \mathbb{R}_+) \|x\| \leq \|x - \alpha y\| \Leftrightarrow (\forall \alpha \in [0, 1]) \|x\| \leq \|x - \alpha y\|$ .

(ii)  $x \perp y \Leftrightarrow (\forall \alpha \in \mathbb{R}) \|x\| \leq \|x - \alpha y\| \Leftrightarrow (\forall \alpha \in [-1, 1]) \|x\| \leq \|x - \alpha y\|$ .

*Proof.* (i): Observe that

$$(\forall \alpha \in \mathbb{R}) \quad \|x - \alpha y\|^2 - \|x\|^2 = \alpha(\alpha\|y\|^2 - 2\langle x | y \rangle). \quad (2.14)$$

Hence, the forward implications follow immediately. Conversely, if for every  $\alpha \in ]0, 1]$ ,  $\|x\| \leq \|x - \alpha y\|$ , then (2.14) implies that  $\langle x | y \rangle \leq \alpha\|y\|^2/2$ . As  $\alpha \downarrow 0$ , we obtain  $\langle x | y \rangle \leq 0$ .

(ii): A consequence of (i), since  $x \perp y \Leftrightarrow [\langle x | y \rangle \leq 0 \text{ and } \langle x | -y \rangle \leq 0]$ .  $\square$

**Lemma 2.13** *Let  $(x_i)_{i \in I}$  and  $(u_i)_{i \in I}$  be finite families in  $\mathcal{H}$  and let  $(\alpha_i)_{i \in I}$  be a family in  $\mathbb{R}$  such that  $\sum_{i \in I} \alpha_i = 1$ . Then the following hold:*

$$(i) \quad \left\langle \sum_{i \in I} \alpha_i x_i \mid \sum_{j \in I} \alpha_j u_j \right\rangle + \sum_{i \in I} \sum_{j \in I} \alpha_i \alpha_j \langle x_i - x_j \mid u_i - u_j \rangle / 2 \\ = \sum_{i \in I} \alpha_i \langle x_i \mid u_i \rangle.$$

$$(ii) \quad \left\| \sum_{i \in I} \alpha_i x_i \right\|^2 + \sum_{i \in I} \sum_{j \in I} \alpha_i \alpha_j \|x_i - x_j\|^2 / 2 = \sum_{i \in I} \alpha_i \|x_i\|^2.$$

*Proof.* (i): We have

$$2 \left\langle \sum_{i \in I} \alpha_i x_i \mid \sum_{j \in I} \alpha_j u_j \right\rangle \\ = \sum_{i \in I} \sum_{j \in I} \alpha_i \alpha_j (\langle x_i \mid u_j \rangle + \langle x_j \mid u_i \rangle) \\ = \sum_{i \in I} \sum_{j \in I} \alpha_i \alpha_j (\langle x_i \mid u_i \rangle + \langle x_j \mid u_j \rangle - \langle x_i - x_j \mid u_i - u_j \rangle) \\ = 2 \sum_{i \in I} \alpha_i \langle x_i \mid u_i \rangle - \sum_{i \in I} \sum_{j \in I} \alpha_i \alpha_j \langle x_i - x_j \mid u_i - u_j \rangle. \quad (2.15)$$

(ii): This follows from (i) when  $(u_i)_{i \in I} = (x_i)_{i \in I}$ .  $\square$

**Corollary 2.14** *Let  $x \in \mathcal{H}$ , let  $y \in \mathcal{H}$ , and let  $\alpha \in \mathbb{R}$ . Then*

$$\|\alpha x + (1 - \alpha)y\|^2 + \alpha(1 - \alpha)\|x - y\|^2 = \alpha\|x\|^2 + (1 - \alpha)\|y\|^2. \quad (2.16)$$



## 2.3 Linear Operators and Functionals

Let  $\mathcal{X}$  and  $\mathcal{Y}$  be real normed vector spaces. We set

$$\mathcal{B}(\mathcal{X}, \mathcal{Y}) = \{T: \mathcal{X} \rightarrow \mathcal{Y} \mid T \text{ is linear and continuous}\} \quad (2.17)$$

and  $\mathcal{B}(\mathcal{X}) = \mathcal{B}(\mathcal{X}, \mathcal{X})$ . Equipped with the norm

$$(\forall T \in \mathcal{B}(\mathcal{X}, \mathcal{Y})) \quad \|T\| = \sup \|T(B(0; 1))\| = \sup_{\substack{x \in \mathcal{X} \\ \|x\| \leq 1}} \|Tx\|, \quad (2.18)$$

$\mathcal{B}(\mathcal{X}, \mathcal{Y})$  is a normed vector space, and it is a Banach space if  $\mathcal{Y}$  is a Banach space.

**Fact 2.15** *Let  $\mathcal{X}$  and  $\mathcal{Y}$  be real normed vector spaces and let  $T: \mathcal{X} \rightarrow \mathcal{Y}$  be linear. Then  $T$  is continuous at a point in  $\mathcal{X}$  if and only if it is Lipschitz continuous.*

The following result is also known as the *Banach–Steinhaus theorem*.

**Lemma 2.16 (uniform boundedness)** *Let  $\mathcal{X}$  be a real Banach space, let  $\mathcal{Y}$  be a real normed vector space, and let  $(T_i)_{i \in I}$  be a family of operators in  $\mathcal{B}(\mathcal{X}, \mathcal{Y})$  that is pointwise bounded, i.e.,*

$$(\forall x \in \mathcal{X}) \quad \sup_{i \in I} \|T_i x\| < +\infty. \quad (2.19)$$

*Then  $\sup_{i \in I} \|T_i\| < +\infty$ .*

*Proof.* Apply Lemma 1.43(i) to  $(\{x \in \mathcal{X} \mid \sup_{i \in I} \|T_i x\| \leq n\})_{n \in \mathbb{N}}$ . □

The Riesz–Fréchet representation theorem states that every continuous linear functional on the real Hilbert space  $\mathcal{H}$  can be identified with a vector in  $\mathcal{H}$ .

**Fact 2.17 (Riesz–Fréchet representation)** *Let  $f \in \mathcal{B}(\mathcal{H}, \mathbb{R})$ . Then there exists a unique vector  $u \in \mathcal{H}$  such that  $(\forall x \in \mathcal{H}) \ f(x) = \langle x \mid u \rangle$ . Moreover,  $\|f\| = \|u\|$ .*

If  $\mathcal{K}$  is a real Hilbert space and  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , the *adjoint* of  $T$  is the unique operator  $T^* \in \mathcal{B}(\mathcal{K}, \mathcal{H})$  that satisfies

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{K}) \quad \langle Tx \mid y \rangle = \langle x \mid T^* y \rangle. \quad (2.20)$$

**Fact 2.18** *Let  $\mathcal{K}$  be a real Hilbert space, let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $\ker T = \{x \in \mathcal{H} \mid Tx = 0\}$  be the kernel of  $T$ . Then the following hold:*

- (i)  $T^{**} = T$ .
- (ii)  $\|T^*\| = \|T\| = \sqrt{\|T^*T\|}$ .

- (iii)  $(\ker T)^\perp = \overline{\text{ran } T^*}$ .
- (iv)  $(\text{ran } T)^\perp = \ker T^*$ .
- (v)  $\ker T^*T = \ker T$  and  $\overline{\text{ran } TT^*} = \overline{\text{ran } T}$ .

**Fact 2.19** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then  $\text{ran } T$  is closed  $\Leftrightarrow \text{ran } T^*$  is closed  $\Leftrightarrow \text{ran } TT^*$  is closed  $\Leftrightarrow \text{ran } T^*T$  is closed  $\Leftrightarrow (\exists \alpha \in \mathbb{R}_{++})(\forall x \in (\ker T)^\perp) \|Tx\| \geq \alpha\|x\|$ .

Suppose that  $u \in \mathcal{H} \setminus \{0\}$  and let  $\eta \in \mathbb{R}$ . A *closed hyperplane* in  $\mathcal{H}$  is a set of the form

$$\{x \in \mathcal{H} \mid \langle x \mid u \rangle = \eta\}; \quad (2.21)$$

a *closed half-space* with *outer normal*  $u$  is a set of the form

$$\{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \eta\}; \quad (2.22)$$

and an *open half-space* with *outer normal*  $u$  is a set of the form

$$\{x \in \mathcal{H} \mid \langle x \mid u \rangle < \eta\}. \quad (2.23)$$

The distance function to  $C = \{x \in \mathcal{H} \mid \langle x \mid u \rangle = \eta\}$  is (see (1.45))

$$d_C: \mathcal{H} \rightarrow \mathbb{R}_+: x \mapsto \frac{|\langle x \mid u \rangle - \eta|}{\|u\|}. \quad (2.24)$$

We conclude this section with an example of a discontinuous linear functional.

**Example 2.20** Assume that  $\mathcal{H}$  is infinite-dimensional and let  $H$  be a Hamel basis of  $\mathcal{H}$ , i.e., a maximally linearly independent subset. Then  $H$  is uncountable. Indeed, if  $\mathcal{H} = \bigcup_{n \in \mathbb{N}} \text{span}\{h_k\}_{0 \leq k \leq n}$  for some Hamel basis  $H = \{h_n\}_{n \in \mathbb{N}}$ , then Lemma 1.43(i) implies that some finite-dimensional linear subspace  $\text{span}\{h_k\}_{0 \leq k \leq n}$  has nonempty interior, which is absurd. The Gram–Schmidt orthonormalization procedure thus guarantees the existence of an orthonormal set  $B = \{e_n\}_{n \in \mathbb{N}}$  and an uncountable set  $C = \{c_a\}_{a \in A}$  such that  $B \cup C$  is a Hamel basis of  $\mathcal{H}$ . Thus, every point in  $\mathcal{H}$  is a (finite) linear combination of elements in  $B \cup C$  and, therefore, the function

$$f: \mathcal{H} \rightarrow \mathbb{R}: x = \sum_{n \in \mathbb{N}} \xi_n e_n + \sum_{a \in A} \eta_a c_a \mapsto \sum_{n \in \mathbb{N}} \xi_n \quad (2.25)$$

is well defined and linear. Now take  $(\alpha_n)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N}) \setminus \ell^1(\mathbb{N})$  (e.g.,  $(\forall n \in \mathbb{N}) \alpha_n = 1/(n+1)$ ) and set

$$(\forall n \in \mathbb{N}) \quad x_n = \sum_{k=0}^n \alpha_k e_k. \quad (2.26)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to some point  $z \in \mathcal{H}$  and  $f(x_n) \rightarrow +\infty$ . This proves that  $f$  is discontinuous at  $z$  and hence discontinuous everywhere

by Fact 2.15. Now set  $(\forall n \in \mathbb{N}) y_n = (x_n - f(x_n)e_0)/\max\{f(x_n), 1\}$ . Then  $(y_n)_{n \in \mathbb{N}}$  lies in  $C = \{x \in \mathcal{H} \mid f(x) = 0\}$  and  $y_n \rightarrow -e_0$ . On the other hand,  $-e_0 \notin C$ , since  $f(-e_0) = -1$ . As a result, the hyperplane  $C$  is not closed. In fact, as will be proved in Example 8.33,  $C$  is dense in  $\mathcal{H}$ .

## 2.4 Strong and Weak Topologies

The metric topology of  $(\mathcal{H}, d)$  is called the *strong topology* (or *norm topology*) of  $\mathcal{H}$ . Thus, a net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  converges strongly to a point  $x$  if  $\|x_a - x\| \rightarrow 0$ ; in symbols,  $x_a \rightarrow x$ . When used without modifiers, topological notions in  $\mathcal{H}$  (closedness, openness, neighborhood, continuity, compactness, convergence, etc.) will always be understood with respect to the strong topology.

**Fact 2.21** *Let  $U$  and  $V$  be closed linear subspaces of  $\mathcal{H}$  such that  $V$  has finite dimension or finite codimension. Then  $U + V$  is a closed linear subspace.*

In addition to the strong topology, a very important alternative topology can be introduced.

**Definition 2.22** The family of all finite intersections of open half-spaces of  $\mathcal{H}$  forms the base of the *weak topology* of  $\mathcal{H}$ ;  $\mathcal{H}^{\text{weak}}$  denotes the resulting topological space.

**Lemma 2.23**  $\mathcal{H}^{\text{weak}}$  is a Hausdorff space.

*Proof.* Let  $x$  and  $y$  be distinct points in  $\mathcal{H}$ . Set  $u = x - y$  and  $w = (x + y)/2$ . Then  $\{z \in \mathcal{H} \mid \langle z - w \mid u \rangle > 0\}$  and  $\{z \in \mathcal{H} \mid \langle z - w \mid u \rangle < 0\}$  are disjoint weak neighborhoods of  $x$  and  $y$ , respectively.  $\square$

A subset of  $\mathcal{H}$  is weakly open if it is a union of finite intersections of open half-spaces. If  $\mathcal{H}$  is infinite-dimensional, nonempty intersections of finitely many open half-spaces are unbounded and, therefore, nonempty weakly open sets are unbounded. A net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  converges weakly to a point  $x \in \mathcal{H}$  if, for every  $u \in \mathcal{H}$ ,  $\langle x_a \mid u \rangle \rightarrow \langle x \mid u \rangle$ ; in symbols,  $x_a \rightharpoonup x$ . Moreover (see Section 1.7), a subset  $C$  of  $\mathcal{H}$  is *weakly closed* if the weak limit of every weakly convergent net in  $C$  is also in  $C$ , and *weakly compact* if every net in  $C$  has a weak cluster point in  $C$ . Likewise (see Section 1.11), a subset  $C$  of  $\mathcal{H}$  is *weakly sequentially closed* if the weak limit of every weakly convergent sequence in  $C$  is also in  $C$ , and *weakly sequentially compact* if every convergent sequence in  $C$  has a weak sequential cluster point in  $C$ . Finally, let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $\mathcal{K}$  be a real Hilbert space, let  $T: D \rightarrow \mathcal{K}$ , and let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $T$  is *weakly continuous* if it is continuous with respect to the weak topologies on  $\mathcal{H}$  and  $\mathcal{K}$ , i.e., if, for every net  $(x_a)_{a \in A}$  in  $D$  such that  $x_a \rightharpoonup x \in D$ , we have  $Tx_a \rightharpoonup Tx$ . Likewise,  $f$  is *weakly lower semicontinuous* at  $x \in \mathcal{H}$  if, for every net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  such that  $x_a \rightharpoonup x$ , we have  $f(x) \leq \underline{\lim} f(x_a)$ .

**Remark 2.24** Strong and weak convergence of a net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  to a point  $x$  in  $\mathcal{H}$  can be interpreted in geometrical terms:  $x_a \rightarrow x$  means that  $d_{\{x\}}(x_a) \rightarrow 0$  whereas, by (2.24),  $x_a \rightharpoonup x$  means that  $d_C(x_a) \rightarrow 0$  for every closed hyperplane  $C$  containing  $x$ .

**Example 2.25** Suppose that  $\mathcal{H}$  is infinite-dimensional, let  $(x_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ , and let  $u$  be a point in  $\mathcal{H}$ . Bessel's inequality yields  $\sum_{n \in \mathbb{N}} |\langle x_n | u \rangle|^2 \leq \|u\|^2$ , hence  $\langle x_n | u \rangle \rightarrow 0$ . Thus  $x_n \rightharpoonup 0$ . However,  $\|x_n\| \equiv 1$  and therefore  $x_n \not\rightarrow 0$ . Actually,  $(x_n)_{n \in \mathbb{N}}$  has no Cauchy subsequence since, for any two distinct positive integers  $n$  and  $m$ , we have  $\|x_n - x_m\|^2 = \|x_n\|^2 + \|x_m\|^2 = 2$ . This also shows that the unit sphere  $\{x \in \mathcal{H} \mid \|x\| = 1\}$  is closed but not weakly sequentially closed.

Suppose that  $\mathcal{H}$  is infinite-dimensional. As seen in Example 2.25, an orthonormal sequence in  $\mathcal{H}$  has no strongly convergent subsequence. Hence, it follows from Fact 1.38 that the closed unit ball of  $\mathcal{H}$  is not compact. This property characterizes infinite-dimensional Hilbert spaces.

**Fact 2.26** *The following are equivalent:*

- (i)  $\mathcal{H}$  is finite-dimensional.
- (ii) The closed unit ball  $B(0; 1)$  of  $\mathcal{H}$  is compact.
- (iii) The weak topology of  $\mathcal{H}$  coincides with its strong topology.
- (iv) The weak topology of  $\mathcal{H}$  is metrizable.

In striking contrast, compactness of closed balls always holds in the weak topology. This fundamental and deep result is known as the *Banach–Alaoglu theorem*.

**Fact 2.27 (Banach–Alaoglu)** *The closed unit ball  $B(0; 1)$  of  $\mathcal{H}$  is weakly compact.*

**Fact 2.28** *The weak topology of the closed unit ball  $B(0; 1)$  of  $\mathcal{H}$  is metrizable if and only if  $\mathcal{H}$  is separable.*

**Lemma 2.29** *Let  $C$  be a subset of  $\mathcal{H}$ . Then  $C$  is weakly compact if and only if it is weakly closed and bounded.*

*Proof.* First, suppose that  $C$  is weakly compact. Then Lemma 1.12 and Lemma 2.23 assert that  $C$  is weakly closed. Now set  $\mathcal{C} = \{\langle x | \cdot \rangle\}_{x \in C} \subset \mathcal{B}(\mathcal{H}, \mathbb{R})$  and take  $u \in \mathcal{H}$ . Then  $\langle \cdot | u \rangle$  is weakly continuous. By Lemma 1.20,  $\{\langle x | u \rangle\}_{x \in C}$  is a compact subset of  $\mathbb{R}$ , and it is therefore bounded by Lemma 1.40. Hence,  $\mathcal{C}$  is pointwise bounded, and Lemma 2.16 implies that  $\sup_{x \in C} \|x\| < +\infty$ , i.e., that  $C$  is bounded. Conversely, suppose that  $C$  is weakly closed and bounded, say  $C \subset B(0; \rho)$  for some  $\rho \in \mathbb{R}_{++}$ . By Fact 2.27,  $B(0; \rho)$  is weakly compact. Using Lemma 1.12 in  $\mathcal{H}^{\text{weak}}$ , we deduce that  $C$  is weakly compact.  $\square$

The following important fact states that weak compactness and weak sequential compactness coincide.

**Fact 2.30 (Eberlein–Šmulian)** *Let  $C$  be a subset of  $\mathcal{H}$ . Then  $C$  is weakly compact if and only if it is weakly sequentially compact.*

**Corollary 2.31** *Let  $C$  be a subset of  $\mathcal{H}$ . Then the following are equivalent:*

- (i)  $C$  is weakly compact.
- (ii)  $C$  is weakly sequentially compact.
- (iii)  $C$  is weakly closed and bounded.

*Proof.* Combine Lemma 2.29 and Fact 2.30. □

**Lemma 2.32** *Let  $C$  be a bounded subset of  $\mathcal{H}$ . Then  $C$  is weakly closed if and only if it is weakly sequentially closed.*

*Proof.* If  $C$  is weakly closed, it is weakly sequentially closed. Conversely, suppose that  $C$  is weakly sequentially closed. By assumption, there exists  $\rho \in \mathbb{R}_{++}$  such that  $C \subset B(0; \rho)$ . Since  $B(0; \rho)$  is weakly sequentially compact by Fact 2.27 and Fact 2.30, it follows from Lemma 2.23 and Lemma 1.33 that  $C$  is weakly sequentially compact. In turn, appealing once more to Fact 2.30, we obtain the weak compactness of  $C$  and therefore its weak closedness by applying Lemma 1.12 in  $\mathcal{H}^{\text{weak}}$ . □

**Remark 2.33** As will be seen in Example 3.31, weakly sequentially closed sets need not be weakly closed.

**Lemma 2.34** *Let  $T: \mathcal{H} \rightarrow \mathcal{K}$  be a continuous affine operator. Then  $T$  is weakly continuous.*

*Proof.* Set  $L: x \mapsto Tx - T0$ , let  $x \in \mathcal{H}$ , let  $y \in \mathcal{K}$ , and let  $(x_a)_{a \in A}$  be a net in  $\mathcal{H}$  such that  $x_a \rightarrow x$ . Then  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  and  $\langle x_a \mid L^*y \rangle \rightarrow \langle x \mid L^*y \rangle$ . Hence,  $\langle Lx_a \mid y \rangle \rightarrow \langle Lx \mid y \rangle$ , i.e.,  $Lx_a \rightarrow Lx$ . We conclude that  $Tx_a = T0 + Lx_a \rightarrow T0 + Lx = Tx$ . □

**Lemma 2.35** *The norm of  $\mathcal{H}$  is weakly lower semicontinuous, i.e., for every net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  and every  $x$  in  $\mathcal{H}$ , we have*

$$x_a \rightarrow x \quad \Rightarrow \quad \|x\| \leq \underline{\lim} \|x_a\|. \quad (2.27)$$

*Proof.* Take a net  $(x_a)_{a \in A}$  in  $\mathcal{H}$  and a point  $x$  in  $\mathcal{H}$  such that  $x_a \rightarrow x$ . Then, by Cauchy–Schwarz,  $\|x\|^2 = \lim |\langle x_a \mid x \rangle| \leq \underline{\lim} \|x_a\| \|x\|$ . □

**Lemma 2.36** *Let  $(x_a)_{a \in A}$  and  $(u_a)_{a \in A}$  be nets in  $\mathcal{H}$ , and let  $x$  and  $u$  be points in  $\mathcal{H}$ . Suppose that  $(x_a)_{a \in A}$  is bounded, that  $x_a \rightarrow x$ , and that  $u_a \rightarrow u$ . Then  $\langle x_a \mid u_a \rangle \rightarrow \langle x \mid u \rangle$ .*

*Proof.* We have  $\sup_{a \in A} \|x_a\| < +\infty$ ,  $\langle x_a - x \mid u \rangle \rightarrow 0$ , and  $\|u_a - u\| \rightarrow 0$ . Since, for every  $a \in A$ ,

$$\begin{aligned} |\langle x_a \mid u_a \rangle - \langle x \mid u \rangle| &= |\langle x_a \mid u_a - u \rangle + \langle x_a - x \mid u \rangle| \\ &\leq \left( \sup_{b \in A} \|x_b\| \right) \|u_a - u\| + |\langle x_a - x \mid u \rangle|, \end{aligned} \quad (2.28)$$

we obtain  $\langle x_a \mid u_a \rangle - \langle x \mid u \rangle \rightarrow 0$ . □

## 2.5 Weak Convergence of Sequences

**Lemma 2.37** *Let  $(x_n)_{n \in \mathbb{N}}$  be a bounded sequence in  $\mathcal{H}$ . Then  $(x_n)_{n \in \mathbb{N}}$  possesses a weakly convergent subsequence.*

*Proof.* First, recall from Lemma 2.23 that  $\mathcal{H}^{\text{weak}}$  is a Hausdorff space. Now set  $\rho = \sup_{n \in \mathbb{N}} \|x_n\|$  and  $C = B(0; \rho)$ . Fact 2.27 and Fact 2.30 imply that  $C$  is weakly sequentially compact. Since  $(x_n)_{n \in \mathbb{N}}$  lies in  $C$ , the claim follows from Definition 1.32.  $\square$

**Lemma 2.38** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly if and only if it is bounded and possesses at most one weak sequential cluster point.*

*Proof.* Suppose that  $x_n \rightharpoonup x \in \mathcal{H}$ . Then it follows from Lemma 2.23 and Fact 1.9 that  $x$  is the unique weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ . Moreover, for every  $u \in \mathcal{H}$ ,  $\langle x_n | u \rangle \rightarrow \langle x | u \rangle$  and therefore  $\sup_{n \in \mathbb{N}} |\langle x_n | u \rangle| < +\infty$ . Upon applying Lemma 2.16 to the sequence of continuous linear functionals  $(\langle x_n | \cdot \rangle)_{n \in \mathbb{N}}$ , we obtain the boundedness of  $(\|x_n\|)_{n \in \mathbb{N}}$ . Conversely, suppose that  $(x_n)_{n \in \mathbb{N}}$  is bounded and possesses at most one weak sequential cluster point. Then Lemma 2.37 asserts that it possesses exactly one weak sequential cluster point. Moreover, it follows from Fact 2.27 and Fact 2.30 that  $(x_n)_{n \in \mathbb{N}}$  lies in a weakly sequentially compact set. Therefore, appealing to Lemma 2.23, we apply Lemma 1.34 in  $\mathcal{H}^{\text{weak}}$  to obtain the desired conclusion.  $\square$

**Lemma 2.39** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty subset of  $\mathcal{H}$ . Suppose that, for every  $x \in C$ ,  $(\|x_n - x\|)_{n \in \mathbb{N}}$  converges and that every weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$  belongs to  $C$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .*

*Proof.* By assumption,  $(x_n)_{n \in \mathbb{N}}$  is bounded. Therefore, in view of Lemma 2.38, it is enough to show that  $(x_n)_{n \in \mathbb{N}}$  cannot have two distinct weak sequential cluster points in  $C$ . To this end, let  $x$  and  $y$  be weak sequential cluster points of  $(x_n)_{n \in \mathbb{N}}$  in  $C$ , say  $x_{k_n} \rightharpoonup x$  and  $x_{l_n} \rightharpoonup y$ . Since  $x$  and  $y$  belong to  $C$ , the sequences  $(\|x_n - x\|)_{n \in \mathbb{N}}$  and  $(\|x_n - y\|)_{n \in \mathbb{N}}$  converge. In turn, since

$$(\forall n \in \mathbb{N}) \quad 2 \langle x_n | x - y \rangle = \|x_n - y\|^2 - \|x_n - x\|^2 + \|x\|^2 - \|y\|^2, \quad (2.29)$$

$(\langle x_n | x - y \rangle)_{n \in \mathbb{N}}$  converges as well, say  $\langle x_n | x - y \rangle \rightarrow \ell$ . Passing to the limit along  $(x_{k_n})_{n \in \mathbb{N}}$  and along  $(x_{l_n})_{n \in \mathbb{N}}$  yields, respectively,  $\ell = \langle x | x - y \rangle = \langle y | x - y \rangle$ . Therefore,  $\|x - y\|^2 = 0$  and hence  $x = y$ .  $\square$

**Proposition 2.40** *Let  $(e_i)_{i \in I}$  be a totally ordered family in  $\mathcal{H}$  such that  $\overline{\text{span}}\{e_i\}_{i \in I} = \mathcal{H}$ , let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$ , and let  $x$  be a point in  $\mathcal{H}$ . Then the following are equivalent:*

- (i)  $x_n \rightharpoonup x$ .

(ii)  $(x_n)_{n \in \mathbb{N}}$  is bounded and  $(\forall i \in I) \langle x_n | e_i \rangle \rightarrow \langle x | e_i \rangle$  as  $n \rightarrow +\infty$ .

*Proof.* (i) $\Rightarrow$ (ii): Lemma 2.38. (ii) $\Rightarrow$ (i): Set  $(y_n)_{n \in \mathbb{N}} = (x_n - x)_{n \in \mathbb{N}}$ . Lemma 2.37 asserts that  $(y_n)_{n \in \mathbb{N}}$  possesses a weak sequential cluster point  $y$ , say  $y_{k_n} \rightharpoonup y$ . In view of Lemma 2.38, it suffices to show that  $y = 0$ . For this purpose, fix  $\varepsilon \in \mathbb{R}_{++}$ . Then there exists a finite subset  $J$  of  $I$  such that  $\|y - z\| \sup_{n \in \mathbb{N}} \|y_{k_n}\| \leq \varepsilon$ , where  $z = \sum_{j \in J} \langle y | e_j \rangle e_j$ . Thus, by Cauchy–Schwarz,

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad |\langle y_{k_n} | y \rangle| &\leq |\langle y_{k_n} | y - z \rangle| + |\langle y_{k_n} | z \rangle| \\ &\leq \varepsilon + \sum_{j \in J} |\langle y | e_j \rangle| |\langle y_{k_n} | e_j \rangle|. \end{aligned} \quad (2.30)$$

Hence  $\overline{\lim} |\langle y_{k_n} | y \rangle| \leq \varepsilon$ . Letting  $\varepsilon \downarrow 0$  yields  $\|y\|^2 = \lim \langle y_{k_n} | y \rangle = 0$ .  $\square$

**Lemma 2.41** *Let  $(x_n)_{n \in \mathbb{N}}$  and  $(u_n)_{n \in \mathbb{N}}$  be sequences in  $\mathcal{H}$ , and let  $x$  and  $u$  be points in  $\mathcal{H}$ . Then the following hold:*

- (i)  $[x_n \rightharpoonup x \text{ and } \overline{\lim} \|x_n\| \leq \|x\|] \Leftrightarrow x_n \rightarrow x$ .
- (ii) *Suppose that  $\mathcal{H}$  is finite-dimensional. Then  $x_n \rightarrow x \Leftrightarrow x_n \rightharpoonup x$ .*
- (iii) *Suppose that  $x_n \rightarrow x$  and  $u_n \rightarrow u$ . Then  $\langle x_n | u_n \rangle \rightarrow \langle x | u \rangle$ .*

*Proof.* (i): Suppose that  $x_n \rightharpoonup x$  and that  $\overline{\lim} \|x_n\| \leq \|x\|$ . Then it follows from Lemma 2.35 that  $\|x\| \leq \underline{\lim} \|x_n\| \leq \overline{\lim} \|x_n\| \leq \|x\|$ , hence  $\|x_n\| \rightarrow \|x\|$ . In turn,  $\|x_n - x\|^2 = \|x_n\|^2 - 2\langle x_n | x \rangle + \|x\|^2 \rightarrow 0$ . Conversely, suppose that  $x_n \rightarrow x$ . Then  $\|x_n\| \rightarrow \|x\|$  by continuity of the norm. On the other hand,  $x_n \rightharpoonup x$  since for every  $n \in \mathbb{N}$  and every  $u \in \mathcal{H}$ , the Cauchy–Schwarz inequality yields  $0 \leq |\langle x_n - x | u \rangle| \leq \|x_n - x\| \|u\|$ .

(ii): Set  $\dim \mathcal{H} = m$  and let  $(e_k)_{1 \leq k \leq m}$  be an orthonormal basis of  $\mathcal{H}$ . Now assume that  $x_n \rightarrow x$ . Then  $\|x_n - x\|^2 = \sum_{k=1}^m |\langle x_n - x | e_k \rangle|^2 \rightarrow 0$ .

(iii): Combine Lemma 2.36 and Lemma 2.38.  $\square$

The combination of Lemma 2.35 and Lemma 2.41(i) yields the following characterization of strong convergence.

**Corollary 2.42** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $x$  be in  $\mathcal{H}$ . Then  $x_n \rightarrow x \Leftrightarrow [x_n \rightharpoonup x \text{ and } \|x_n\| \rightarrow \|x\|]$ .*

## 2.6 Differentiability

In this section,  $\mathcal{K}$  is a real Banach space.

**Definition 2.43** Let  $C$  be a subset of  $\mathcal{H}$ , let  $T: C \rightarrow \mathcal{K}$ , and let  $x \in C$  be such that  $(\forall y \in \mathcal{H})(\exists \alpha \in \mathbb{R}_{++}) [x, x + \alpha y] \subset C$ . Then  $T$  is *Gâteaux differentiable* at  $x$  if there exists an operator  $DT(x) \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , called the *Gâteaux derivative* of  $T$  at  $x$ , such that

$$(\forall y \in \mathcal{H}) \quad DT(x)y = \lim_{\alpha \downarrow 0} \frac{T(x + \alpha y) - T(x)}{\alpha}. \quad (2.31)$$

Higher-order Gâteaux derivatives are defined inductively. Thus, the *second Gâteaux derivative* of  $T$  at  $x$  is the operator  $D^2T(x) \in \mathcal{B}(\mathcal{H}, \mathcal{B}(\mathcal{H}, \mathcal{K}))$  that satisfies

$$(\forall y \in \mathcal{H}) \quad D^2T(x)y = \lim_{\alpha \downarrow 0} \frac{DT(x + \alpha y) - DT(x)}{\alpha}. \quad (2.32)$$

The Gâteaux derivative  $DT(x)$  in Definition 2.43 is unique whenever it exists (Exercise 2.9). Moreover, since  $DT(x)$  is linear, for every  $y \in \mathcal{H}$ , we have  $DT(x)y = -DT(x)(-y)$ , and we can therefore replace (2.31) by

$$(\forall y \in \mathcal{H}) \quad DT(x)y = \lim_{0 \neq \alpha \rightarrow 0} \frac{T(x + \alpha y) - T(x)}{\alpha}. \quad (2.33)$$

If the convergence in (2.33) is uniform with respect to  $y$  on bounded sets, then  $x \in \text{int } C$  and we obtain the following notion.

**Remark 2.44** Let  $C$  be a subset of  $\mathcal{H}$ , let  $f: C \rightarrow \mathbb{R}$ , and suppose that  $f$  is Gâteaux differentiable at  $x \in C$ . Then, by Fact 2.17, there exists a unique vector  $\nabla f(x) \in \mathcal{H}$  such that

$$(\forall y \in \mathcal{H}) \quad Df(x)y = \langle y \mid \nabla f(x) \rangle. \quad (2.34)$$

We call  $\nabla f(x)$  the *Gâteaux gradient* of  $f$  at  $x$ . If  $f$  is Gâteaux differentiable on  $C$ , the *gradient operator* is  $\nabla f: C \rightarrow \mathcal{H}: x \mapsto \nabla f(x)$ . Likewise, if  $f$  is twice Gâteaux differentiable at  $x$ , we can identify  $D^2f(x)$  with an operator  $\nabla^2 f(x) \in \mathcal{B}(\mathcal{H})$  in the sense that

$$(\forall y \in \mathcal{H})(\forall z \in \mathcal{H}) \quad (D^2f(x)y)z = \langle z \mid \nabla^2 f(x)y \rangle. \quad (2.35)$$

We call  $\nabla^2 f(x)$  the (Gâteaux) *Hessian* of  $f$  at  $x$ .

**Definition 2.45** Let  $x \in \mathcal{H}$ , let  $C \in \mathcal{V}(x)$ , and let  $T: C \rightarrow \mathcal{K}$ . Then  $T$  is *Fréchet differentiable* at  $x$  if there exists an operator  $DT(x) \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , called the *Fréchet derivative* of  $T$  at  $x$ , such that

$$\lim_{0 \neq \|y\| \rightarrow 0} \frac{\|T(x + y) - Tx - DT(x)y\|}{\|y\|} = 0. \quad (2.36)$$

Higher-order Fréchet derivatives are defined inductively. Thus, the *second Fréchet derivative* of  $T$  at  $x$  is the operator  $D^2T(x) \in \mathcal{B}(\mathcal{H}, \mathcal{B}(\mathcal{H}, \mathcal{K}))$  that satisfies

$$\lim_{0 \neq \|y\| \rightarrow 0} \frac{\|DT(x + y) - DTx - D^2T(x)y\|}{\|y\|} = 0. \quad (2.37)$$

The *Fréchet gradient* of a function  $f: C \rightarrow \mathbb{R}$  at  $x \in C$  is defined as in Remark 2.44. Here are some examples.



**Example 2.46** Let  $L \in \mathcal{B}(\mathcal{H})$ , let  $u \in \mathcal{H}$ , let  $x \in \mathcal{H}$ , and set  $f: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \langle Ly \mid y \rangle - \langle y \mid u \rangle$ . Then  $f$  is twice Fréchet differentiable on  $\mathcal{H}$  with  $\nabla f(x) = (L + L^*)x - u$  and  $\nabla^2 f(x) = L + L^*$ .

*Proof.* Take  $y \in \mathcal{H}$ . Since

$$\begin{aligned} f(x+y) - f(x) &= \langle Lx \mid y \rangle + \langle Ly \mid x \rangle + \langle Ly \mid y \rangle - \langle y \mid u \rangle \\ &= \langle y \mid (L + L^*)x \rangle - \langle y \mid u \rangle + \langle Ly \mid y \rangle, \end{aligned} \quad (2.38)$$

we have

$$|f(x+y) - f(x) - \langle y \mid (L + L^*)x - u \rangle| = |\langle Ly \mid y \rangle| \leq \|L\| \|y\|^2. \quad (2.39)$$

In view of (2.36),  $f$  is Fréchet differentiable at  $x$  with  $\nabla f(x) = (L + L^*)x - u$ . In turn, (2.37) yields  $\nabla^2 f(x) = L + L^*$ .  $\square$

**Example 2.47** Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow \mathbb{R}$  be a symmetric bilinear form such that, for some  $\beta \in \mathbb{R}_+$ ,

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad |F(x, y)| \leq \beta \|x\| \|y\|, \quad (2.40)$$

let  $\ell \in \mathcal{B}(\mathcal{H}, \mathbb{R})$ , let  $x \in \mathcal{H}$ , and set  $f: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto (1/2)F(y, y) - \ell(y)$ . Then  $f$  is Fréchet differentiable on  $\mathcal{H}$  with  $Df(x) = F(x, \cdot) - \ell$ .

*Proof.* Take  $y \in \mathcal{H}$ . Then,

$$\begin{aligned} f(x+y) - f(x) &= \frac{1}{2}F(x+y, x+y) - \ell(x+y) - \frac{1}{2}F(x, x) + \ell(x) \\ &= \frac{1}{2}F(y, y) + F(x, y) - \ell(y). \end{aligned} \quad (2.41)$$

Consequently, (2.40) yields

$$2|f(x+y) - f(x) - (F(x, y) - \ell(y))| = |F(y, y)| \leq \beta \|y\|^2, \quad (2.42)$$

and we infer from (2.36) and (2.40) that  $Df(x)y = F(x, y) - \ell(y)$ .  $\square$

**Example 2.48** Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $r \in \mathcal{K}$ , let  $x \in \mathcal{H}$ , and set  $f: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \|Ly - r\|^2$ . Then  $f$  is twice Fréchet differentiable on  $\mathcal{H}$  with  $\nabla f(x) = 2L^*(Lx - r)$  and  $\nabla^2 f(x) = 2L^*L$ .

*Proof.* Set  $F: \mathcal{H} \times \mathcal{H} \rightarrow \mathbb{R}: (y, z) \mapsto (1/2)\langle L^*Ly \mid z \rangle$ ,  $\ell: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \langle y \mid L^*r \rangle$ , and  $\alpha = (1/2)\|r\|^2$ . Then  $(\forall y \in \mathcal{H}) f(y) = 2(F(y, y) - \ell(y) + \alpha)$ . Hence we derive from Example 2.47 that  $\nabla f(x) = 2L^*(Lx - r)$ , and from (2.37) that  $\nabla^2 f(x) = 2L^*L$ .  $\square$

**Lemma 2.49** Let  $x \in \mathcal{H}$ , let  $C \in \mathcal{V}(x)$ , and let  $T: C \rightarrow \mathcal{K}$ . Suppose that  $T$  is Fréchet differentiable at  $x$ . Then the following hold:

- (i)  $T$  is Gâteaux differentiable at  $x$  and the two derivatives coincide.

(ii)  $T$  is continuous at  $x$ .

*Proof.* Denote the Fréchet derivative of  $T$  at  $x$  by  $L_x$ .

(i): Let  $\alpha \in \mathbb{R}_{++}$  and  $y \in \mathcal{H} \setminus \{0\}$ . Then

$$\left\| \frac{T(x + \alpha y) - Tx}{\alpha} - L_x y \right\| = \|y\| \frac{\|T(x + \alpha y) - Tx - L_x(\alpha y)\|}{\|\alpha y\|} \quad (2.43)$$

converges to 0 as  $\alpha \downarrow 0$ .

(ii): Fix  $\varepsilon \in \mathbb{R}_{++}$ . By (2.36), we can find  $\delta \in ]0, \varepsilon/(\varepsilon + \|L_x\|)]$  such that  $(\forall y \in B(0; \delta)) \|T(x+y) - Tx - L_x y\| \leq \varepsilon \|y\|$ . Thus  $(\forall y \in B(0; \delta)) \|T(x+y) - Tx\| \leq \|T(x+y) - Tx - L_x y\| + \|L_x y\| \leq \varepsilon \|y\| + \|L_x\| \|y\| \leq \delta(\varepsilon + \|L_x\|) \leq \varepsilon$ . It follows that  $T$  is continuous at  $x$ .  $\square$

**Fact 2.50** Let  $T: \mathcal{H} \rightarrow \mathcal{K}$  and let  $x \in \mathcal{H}$ . Suppose that the Gâteaux derivative of  $T$  exists on a neighborhood of  $x$  and that  $DT$  is continuous at  $x$ . Then  $T$  is Fréchet differentiable at  $x$ .

**Fact 2.51** Let  $x \in \mathcal{H}$ , let  $U$  be a neighborhood of  $x$ , let  $\mathcal{G}$  be a real Banach space, let  $T: U \rightarrow \mathcal{G}$ , let  $V$  be a neighborhood of  $Tx$ , and let  $R: V \rightarrow \mathcal{K}$ . Suppose that  $T$  is Fréchet differentiable at  $x$  and that  $R$  is Gâteaux differentiable at  $Tx$ . Then  $R \circ T$  is Gâteaux differentiable at  $x$  and  $D(R \circ T)(x) = (DR(Tx)) \circ DT(x)$ . If  $R$  is Fréchet differentiable at  $x$ , then so is  $R \circ T$ .

**Example 2.52** Let  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \|x\|$ . Then  $f = \sqrt{\|\cdot\|^2}$  and, since Example 2.48 asserts that  $\|\cdot\|^2$  is Fréchet differentiable with gradient operator  $\nabla \|\cdot\|^2 = 2\text{Id}$ , it follows from Fact 2.51 that  $f$  is Fréchet differentiable on  $\mathcal{H} \setminus \{0\}$  with  $(\forall x \in \mathcal{H} \setminus \{0\}) \nabla f(x) = x/\|x\|$ . On the other hand,  $f$  is not Gâteaux differentiable at  $x = 0$ , since although the limit in (2.31) exists, it is not linear with respect to  $y$ :  $(\forall y \in \mathcal{H}) \lim_{\alpha \downarrow 0} (\|0 + \alpha y\| - \|0\|)/\alpha = \|y\|$ .

**Fact 2.53** Let  $x \in \mathcal{H}$ , let  $U$  be a neighborhood of  $x$ , let  $\mathcal{K}$  be a real Banach space, and let  $T: U \rightarrow \mathcal{K}$ . Suppose that  $T$  is twice Fréchet differentiable at  $x$ . Then  $(\forall (y, z) \in \mathcal{H} \times \mathcal{H}) (D^2 T(x)y)z = (D^2 T(x)z)y$ .

**Example 2.54** Let  $x \in \mathcal{H}$ , let  $U$  be a neighborhood of  $x$ , and let  $f: U \rightarrow \mathbb{R}$ . Suppose that  $f$  is twice Fréchet differentiable at  $x$ . Then, in view of Fact 2.53 and (2.35),  $\nabla^2 f(x)$  is self-adjoint.

## Exercises

**Exercise 2.1** Let  $x$  and  $y$  be points in  $\mathcal{H}$ , and let  $\alpha$  and  $\beta$  be real numbers in  $\mathbb{R}_+$ . Show that  $4 \langle \alpha x - \beta y \mid y - x \rangle \leq \alpha \|y\|^2 + \beta \|x\|^2$ .

**Exercise 2.2** Let  $x$  and  $y$  be in  $\mathcal{H}$  and let  $\alpha$  and  $\beta$  be in  $\mathbb{R}$ . Show that

$$\begin{aligned} & \alpha(1-\alpha)\|\beta x + (1-\beta)y\|^2 + \beta(1-\beta)\|\alpha x - (1-\alpha)y\|^2 \\ &= (\alpha + \beta - 2\alpha\beta)(\alpha\beta\|x\|^2 + (1-\alpha)(1-\beta)\|y\|^2). \end{aligned} \quad (2.44)$$

**Exercise 2.3** Suppose that  $\mathcal{H}$  is infinite-dimensional. Show that every weakly compact set has an empty weak interior.

**Exercise 2.4** Provide an unbounded convergent net in  $\mathbb{R}$  (compare with Lemma 2.38).

**Exercise 2.5** Construct a sequence in  $\mathcal{H}$  that converges weakly and possesses a strong sequential cluster point, but that does not converge strongly.

**Exercise 2.6** Let  $C$  be a subset of  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) C \cap B(0; n)$  is weakly sequentially closed. Show that  $C$  is weakly sequentially closed and compare with Lemma 1.39.

**Exercise 2.7** Show that the conclusion of Lemma 2.41(iii) fails if the strong convergence of  $(u_n)_{n \in \mathbb{N}}$  is replaced by weak convergence.

**Exercise 2.8 (Opial's condition)** Let  $(x_n)_{n \in \mathbb{N}}$  be a weakly convergent sequence in  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Show that  $x_n \rightharpoonup x$  if and only if

$$(\forall y \in \mathcal{H}) \quad \underline{\lim} \|x_n - y\|^2 = \|x - y\|^2 + \underline{\lim} \|x_n - x\|^2. \quad (2.45)$$

In particular, if  $x_n \rightharpoonup x$  and  $y \neq x$ , then  $\underline{\lim} \|x_n - y\| > \underline{\lim} \|x_n - x\|$ . This implication is known as *Opial's condition*.

**Exercise 2.9** Show that if the derivative  $DT(x)$  exists in Definition 2.43, it is unique.

**Exercise 2.10** Let  $D$  be a nonempty open interval in  $\mathbb{R}$ , let  $f: D \rightarrow \mathbb{R}$ , and let  $x \in D$ . Show that the notions of Gâteaux and Fréchet differentiability of  $f$  at  $x$  coincide with classical differentiability, and that the Gâteaux and Fréchet derivatives coincide with the classical derivative

$$f'(x) = \lim_{h \neq 0 \rightarrow 0} \frac{f(x+h) - f(x)}{h}. \quad (2.46)$$

**Exercise 2.11** Consider the function

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}: (\xi_1, \xi_2) \mapsto \begin{cases} \frac{\xi_1^2 \xi_2^4}{\xi_1^4 + \xi_2^8}, & \text{if } (\xi_1, \xi_2) \neq (0, 0); \\ 0, & \text{if } (\xi_1, \xi_2) = (0, 0). \end{cases} \quad (2.47)$$

Show that  $f$  is Gâteaux differentiable, but not continuous, at  $(0, 0)$ .

**Exercise 2.12** Consider the function

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}: x = (\xi_1, \xi_2) \mapsto \begin{cases} \frac{\xi_1 \xi_2^4}{\xi_1^2 + \xi_2^4}, & \text{if } (\xi_1, \xi_2) \neq (0, 0); \\ 0, & \text{if } (\xi_1, \xi_2) = (0, 0). \end{cases} \quad (2.48)$$

Show that  $f$  is Fréchet differentiable at  $(0, 0)$  and that  $\nabla f$  is not continuous at  $(0, 0)$ . Conclude that the converse of Fact 2.50 does not hold.

**Exercise 2.13** Consider the function

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}: (\xi_1, \xi_2) \mapsto \begin{cases} \frac{\xi_1 \xi_2^3}{\xi_1^2 + \xi_2^4}, & \text{if } (\xi_1, \xi_2) \neq (0, 0); \\ 0, & \text{if } (\xi_1, \xi_2) = (0, 0). \end{cases} \quad (2.49)$$

Show that, at  $(0, 0)$ ,  $f$  is continuous and Gâteaux differentiable, but not Fréchet differentiable.

**Exercise 2.14** Consider the function

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}: (\xi_1, \xi_2) \mapsto \begin{cases} \frac{\xi_1 \xi_2^2}{\xi_1^2 + \xi_2^2}, & \text{if } (\xi_1, \xi_2) \neq (0, 0); \\ 0, & \text{if } (\xi_1, \xi_2) = (0, 0). \end{cases} \quad (2.50)$$

Show that  $f$  is continuous, and that at  $x = (0, 0)$  the limit on the right-hand side of (2.31) exists, but it is not linear as a function of  $y = (\eta_1, \eta_2)$ . Conclude that  $f$  is not Gâteaux differentiable at  $(0, 0)$ .

# Chapter 3

## Convex Sets

In this chapter we introduce the fundamental notion of the convexity of a set and establish various properties of convex sets. The key result is Theorem 3.14, which asserts that every nonempty closed convex subset  $C$  of  $\mathcal{H}$  is a Chebyshev set, i.e., that every point in  $\mathcal{H}$  possesses a unique best approximation from  $C$ , and which provides a characterization of this best approximation.

### 3.1 Definition and Examples

**Definition 3.1** A subset  $C$  of  $\mathcal{H}$  is *convex* if  $(\forall \alpha \in ]0, 1[) \alpha C + (1 - \alpha)C = C$  or, equivalently, if

$$(\forall x \in C)(\forall y \in C) \quad ]x, y[ \subset C. \quad (3.1)$$

In particular,  $\mathcal{H}$  and  $\emptyset$  are convex.

**Example 3.2** In each of the following cases,  $C$  is a convex subset of  $\mathcal{H}$ .

- (i)  $C$  is a ball.
- (ii)  $C$  is an affine subspace.
- (iii)  $C$  is a half-space.
- (iv)  $C = \bigcap_{i \in I} C_i$ , where  $(C_i)_{i \in I}$  is a family of convex subsets of  $\mathcal{H}$ .

The intersection property (iv) above justifies the following definition.

**Definition 3.3** Let  $C \subset \mathcal{H}$ . The *convex hull* of  $C$  is the intersection of all the convex subsets of  $\mathcal{H}$  containing  $C$ , i.e., the smallest convex subset of  $\mathcal{H}$  containing  $C$ . It is denoted by  $\text{conv } C$ . The *closed convex hull* of  $C$  is the smallest closed convex subset of  $\mathcal{H}$  containing  $C$ . It is denoted by  $\overline{\text{conv}} C$ .

The proof of the following simple fact is left as Exercise 3.1.

**Proposition 3.4** Let  $C \subset \mathcal{H}$  and let  $D$  be the set of all convex combinations of points in  $C$ , i.e.,

$$D = \left\{ \sum_{i \in I} \alpha_i x_i \mid I \text{ finite, } \{x_i\}_{i \in I} \subset C, \{\alpha_i\}_{i \in I} \subset ]0, 1], \sum_{i \in I} \alpha_i = 1 \right\}. \quad (3.2)$$

Then  $D = \text{conv } C$ .

**Proposition 3.5** Let  $\mathcal{K}$  be a real Hilbert space, let  $T: \mathcal{H} \rightarrow \mathcal{K}$  be an affine operator, and let  $C$  and  $D$  be convex subsets of  $\mathcal{H}$  and  $\mathcal{K}$ , respectively. Then  $T(C)$  and  $T^{-1}(D)$  are convex subsets of  $\mathcal{K}$  and  $\mathcal{H}$ , respectively.

*Proof.* It follows from (1.12) that  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) T([x, y]) = ]Tx, Ty[$ . Now take two points in  $T(C)$ , say  $Tx$  and  $Ty$ , where  $x$  and  $y$  are in  $C$ . By convexity,  $]x, y[ \subset C$  and, therefore,  $]Tx, Ty[ = T([x, y]) \subset T(C)$ . Thus,  $T(C)$  is convex. Finally, let  $x$  and  $y$  be two points in  $T^{-1}(D)$ . Then  $Tx$  and  $Ty$  are in  $D$  and, by convexity,  $T([x, y]) = ]Tx, Ty[ \subset D$ . Therefore  $]x, y[ \subset T^{-1}(T([x, y])) \subset T^{-1}(D)$ , which proves the convexity of  $T^{-1}(D)$ .  $\square$

**Proposition 3.6** Let  $(C_i)_{i \in I}$  be a totally ordered finite family of  $m$  convex subsets of  $\mathcal{H}$ . Then the following hold:

- (i)  $\times_{i \in I} C_i$  is convex.
- (ii)  $(\forall (\alpha_i)_{i \in I} \in \mathbb{R}^m) \sum_{i \in I} \alpha_i C_i$  is convex.

*Proof.* (i): Straightforward.

(ii): This is a consequence of (i) and Proposition 3.5 since  $\sum_{i \in I} \alpha_i C_i = L(\times_{i \in I} C_i)$ , where  $L: \mathcal{H}^m \rightarrow \mathcal{H}: (x_i)_{i \in I} \mapsto \sum_{i \in I} \alpha_i x_i$  is linear.  $\square$

## 3.2 Best Approximation Properties

**Definition 3.7** Let  $C$  be a subset of  $\mathcal{H}$ , let  $x \in \mathcal{H}$ , and let  $p \in C$ . Then  $p$  is a *best approximation* to  $x$  from  $C$  (or a *projection* of  $x$  onto  $C$ ) if  $\|x - p\| = d_C(x)$ . If every point in  $\mathcal{H}$  has at least one projection onto  $C$ , then  $C$  is *proximal*. If every point in  $\mathcal{H}$  has exactly one projection onto  $C$ , then  $C$  is a *Chebyshev set*. In this case, the *projector* (or *projection operator*) onto  $C$  is the operator, denoted by  $P_C$ , that maps every point in  $\mathcal{H}$  to its unique projection onto  $C$ .

**Example 3.8** Let  $\{e_i\}_{i \in I}$  be a finite orthonormal set in  $\mathcal{H}$ , let  $V = \text{span}\{e_i\}_{i \in I}$ , and let  $x \in \mathcal{H}$ . Then  $V$  is a Chebyshev set,

$$P_V x = \sum_{i \in I} \langle x | e_i \rangle e_i, \quad \text{and} \quad d_V(x) = \sqrt{\|x\|^2 - \sum_{i \in I} |\langle x | e_i \rangle|^2}. \quad (3.3)$$

*Proof.* For every family  $(\alpha_i)_{i \in I}$  in  $\mathbb{R}$ , we have

$$\begin{aligned} \left\| x - \sum_{i \in I} \alpha_i e_i \right\|^2 &= \|x\|^2 - 2 \left\langle x \mid \sum_{i \in I} \alpha_i e_i \right\rangle + \left\| \sum_{i \in I} \alpha_i e_i \right\|^2 \\ &= \|x\|^2 - 2 \sum_{i \in I} \alpha_i \langle x \mid e_i \rangle + \sum_{i \in I} |\alpha_i|^2 \\ &= \|x\|^2 - \sum_{i \in I} |\langle x \mid e_i \rangle|^2 + \sum_{i \in I} |\alpha_i - \langle x \mid e_i \rangle|^2. \end{aligned} \quad (3.4)$$

Therefore, the function  $(\alpha_i)_{i \in I} \mapsto \|x - \sum_{i \in I} \alpha_i e_i\|^2$  admits  $(\langle x \mid e_i \rangle)_{i \in I}$  as its unique minimizer, and its minimum value is  $\|x\|^2 - \sum_{i \in I} |\langle x \mid e_i \rangle|^2$ .  $\square$

**Remark 3.9** Let  $C$  be a nonempty subset of  $\mathcal{H}$ .

- (i) Since  $\overline{C} = \{x \in \mathcal{H} \mid d_C(x) = 0\}$ , no point in  $\overline{C} \setminus C$  has a projection onto  $C$ . A proximal set (in particular a Chebyshev set) must therefore be closed.
- (ii) If  $C$  is a finite-dimensional linear subspace of  $\mathcal{H}$ , then it is a Chebyshev set and hence closed by (i). This follows from Example 3.8, since  $C$  possesses a finite orthonormal basis by Gram–Schmidt.

**Proposition 3.10** Suppose that  $\mathcal{H}$  is finite-dimensional and let  $C$  be a Chebyshev subset of  $\mathcal{H}$ . Then  $P_C$  is continuous.

*Proof.* Let  $x \in \mathcal{H}$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  such that  $x_n \rightarrow x$ . By Example 1.47,  $d_C$  is continuous and thus

$$\|x_n - P_C x_n\| = d_C(x_n) \rightarrow d_C(x) = \|x - P_C x\|. \quad (3.5)$$

Hence,  $(P_C x_n)_{n \in \mathbb{N}}$  is bounded. Now let  $y$  be a cluster point of  $(P_C x_n)_{n \in \mathbb{N}}$ , say  $P_C x_{k_n} \rightarrow y$ . Then Remark 3.9(i) asserts that  $y \in C$ , and (3.5) implies that  $\|x_{k_n} - P_C x_{k_n}\| \rightarrow \|x - y\| = d_C(x)$ . It follows that  $y = P_C x$  is the only cluster point of the bounded sequence  $(P_C x_n)_{n \in \mathbb{N}}$ . Therefore,  $P_C x_n \rightarrow P_C x$ .  $\square$

In connection with Remark 3.9(i), the next example shows that closedness is not sufficient to guarantee proximality.

**Example 3.11** Suppose that  $\mathcal{H}$  is infinite-dimensional, let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ , and let  $(\alpha_n)_{n \in \mathbb{N}}$  be a sequence in  $]1, +\infty[$  such that  $\alpha_n \downarrow 1$ . Set  $(\forall n \in \mathbb{N}) x_n = \alpha_n e_n$  and  $C = \{x_n\}_{n \in \mathbb{N}}$ . Then, for any two distinct points  $x_n$  and  $x_m$  in  $C$ , we have  $\|x_n - x_m\|^2 = \|x_n\|^2 + \|x_m\|^2 > \|e_n\|^2 + \|e_m\|^2 = 2$ . Therefore, every convergent sequence in  $C$  is eventually constant and  $C$  is thus closed. However, 0 has no projection onto  $C$ , since  $(\forall n \in \mathbb{N}) d_C(0) = 1 < \alpha_n = \|0 - x_n\|$ .

**Proposition 3.12** Let  $C$  be a nonempty weakly closed subset of  $\mathcal{H}$ . Then  $C$  is proximal.

*Proof.* Let  $x \in \mathcal{H}$  and  $z \in C$ , and set  $D = C \cap B(x; \|x - z\|)$  and  $f: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \|x - y\|$ . It is enough to show that  $f$  admits a minimizer over  $D$ , for it will be a projection of  $x$  onto  $C$ . Since  $C$  is weakly closed and  $B(x; \|x - z\|)$  is weakly compact by Fact 2.27, it follows from Lemma 1.12 and Lemma 2.23 that  $D$  is weakly compact and, by construction, nonempty. Hence, since  $f$  is weakly lower semicontinuous by Lemma 2.35, applying Theorem 1.28 in  $\mathcal{H}^{\text{weak}}$  yields the existence of a minimizer of  $f$  over  $D$ .  $\square$

**Corollary 3.13** *Suppose that  $\mathcal{H}$  is finite-dimensional. Let  $C$  be a nonempty subset of  $\mathcal{H}$ . Then  $C$  is proximal if and only if it is closed.*

*Proof.* An immediate consequence of Remark 3.9(i), Proposition 3.12, and Fact 2.26.  $\square$

Every Chebyshev set is a proximal set. However, a proximal set may not be a Chebyshev set: for instance, in  $\mathcal{H} = \mathbb{R}$ , the projections of 0 onto  $C = \{-1, 1\}$  are 1 and  $-1$ . A fundamental result is that nonempty closed convex sets are Chebyshev sets.

**Theorem 3.14** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $C$  is a Chebyshev set and, for every  $x$  and  $p$  in  $\mathcal{H}$ ,*

$$p = P_C x \iff [p \in C \text{ and } (\forall y \in C) \langle y - p \mid x - p \rangle \leq 0]. \quad (3.6)$$

*Proof.* Let  $x \in \mathcal{H}$ . Then it is enough to show that  $x$  admits a unique projection onto  $C$ , and that this projection is characterized by (3.6). By definition of  $d_C$  (see (1.45)), there exists a sequence  $(y_n)_{n \in \mathbb{N}}$  in  $C$  such that  $d_C(x) = \lim \|y_n - x\|$ . Now take  $m$  and  $n$  in  $\mathbb{N}$ . Since  $C$  is convex,  $(y_n + y_m)/2 \in C$  and therefore  $\|x - (y_n + y_m)/2\| \geq d_C(x)$ . It follows from Apollonius's identity (Lemma 2.11(iv)) that

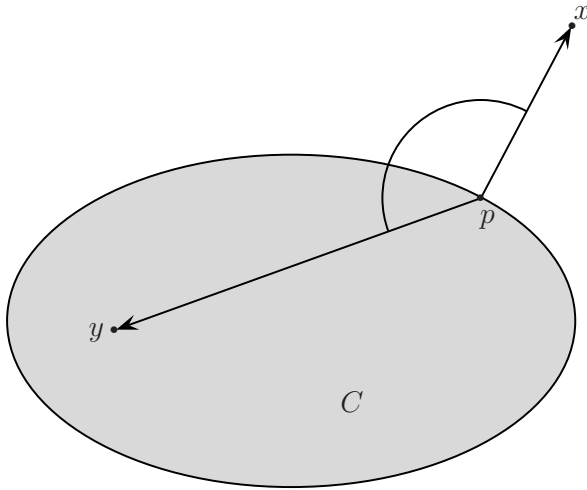
$$\begin{aligned} \|y_n - y_m\|^2 &= 2\|y_n - x\|^2 + 2\|y_m - x\|^2 - 4\|x - (y_n + y_m)/2\|^2 \\ &\leq 2\|y_n - x\|^2 + 2\|y_m - x\|^2 - 4d_C^2(x). \end{aligned} \quad (3.7)$$

Letting  $m$  and  $n$  go to  $+\infty$ , we obtain that  $(y_n)_{n \in \mathbb{N}}$  is a Cauchy sequence. It therefore converges to some point  $p \in C$ , since  $C$  is complete as a closed subset of  $\mathcal{H}$ . The continuity of  $\|\cdot - x\|$  then yields  $\lim \|y_n - x\| = \|p - x\|$ , hence  $d_C(x) = \|p - x\|$ . This shows the existence of  $p$ . Now suppose that  $q \in C$  satisfies  $d_C(x) = \|q - x\|$ . Then  $(p + q)/2 \in C$  and, by Apollonius's identity,  $\|p - q\|^2 = 2\|p - x\|^2 + 2\|q - x\|^2 - 4\|x - (p + q)/2\|^2 = 4d_C^2(x) - 4\|x - (p + q)/2\|^2 \leq 0$ . This implies that  $p = q$  and shows uniqueness. Finally, for every  $y \in C$  and  $\alpha \in [0, 1]$ , set  $y_\alpha = \alpha y + (1 - \alpha)p$ , which belongs to  $C$  by convexity. Lemma 2.12(i) yields

$$\begin{aligned} \|x - p\| = d_C(x) &\iff (\forall y \in C)(\forall \alpha \in [0, 1]) \quad \|x - p\| \leq \|x - y_\alpha\| \\ &\iff (\forall y \in C)(\forall \alpha \in [0, 1]) \quad \|x - p\| \leq \|x - p - \alpha(y - p)\| \\ &\iff (\forall y \in C) \quad \langle y - p \mid x - p \rangle \leq 0, \end{aligned} \quad (3.8)$$



which establishes the characterization.  $\square$



**Fig. 3.1** Projection onto a nonempty closed convex set  $C$  in the Euclidean plane. The characterization (3.6) states that  $p \in C$  is the projection of  $x$  onto  $C$  if and only if the vectors  $x - p$  and  $y - p$  form a right or obtuse angle for every  $y \in C$ .

**Remark 3.15** Theorem 3.14 states that every nonempty closed convex set is a Chebyshev set. Conversely, as seen above, a Chebyshev set must be nonempty and closed. The famous *Chebyshev problem* asks whether every Chebyshev set must indeed be convex. The answer is affirmative if  $\mathcal{H}$  is finite-dimensional (see Corollary 21.13), but remains an open problem otherwise. For a discussion, see [100].

The following example is obtained by checking (3.6) (further examples will be provided in Chapter 28).

**Example 3.16** Let  $C = B(0; 1)$ . Then

$$(\forall x \in \mathcal{H}) \quad P_C x = \frac{1}{\max\{\|x\|, 1\}} x. \quad (3.9)$$

**Proposition 3.17** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , and let  $x$  and  $y$  be in  $\mathcal{H}$ . Then  $P_{y+C} x = y + P_C(x - y)$ .

*Proof.* It is clear that  $y + P_C(x - y) \in y + C$ . Using Theorem 3.14, we obtain

$$\begin{aligned} (\forall z \in C) \quad & \langle (y + z) - (y + P_C(x - y)) \mid x - (y + P_C(x - y)) \rangle \\ &= \langle z - P_C(x - y) \mid (x - y) - P_C(x - y) \rangle \\ &\leq 0, \end{aligned}$$

and we conclude that  $P_{y+C}x = y + P_C(x - y)$ .  $\square$

A decreasing sequence of nonempty closed convex sets can have an empty intersection (consider the sequence  $([n, +\infty])_{n \in \mathbb{N}}$  in  $\mathcal{H} = \mathbb{R}$ ). Next, we show that the boundedness of the sets prevents such a situation.

**Proposition 3.18** *Let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of nonempty bounded closed convex subsets of  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) C_{n+1} \subset C_n$ . Then  $\bigcap_{n \in \mathbb{N}} C_n \neq \emptyset$ .*

*Proof.* Using Theorem 3.14, we define  $(\forall n \in \mathbb{N}) p_n = P_{C_n} 0$ . The assumptions imply that  $(\|p_n\|)_{n \in \mathbb{N}}$  is increasing and bounded, hence convergent. For every  $m$  and  $n$  in  $\mathbb{N}$  such that  $m \leq n$ , since  $(p_n + p_m)/2 \in C_m$ , Lemma 2.11 yields  $\|p_n - p_m\|^2 = 2(\|p_n\|^2 + \|p_m\|^2) - 4\|(p_n + p_m)/2\|^2 \leq 2(\|p_n\|^2 - \|p_m\|^2) \rightarrow 0$  as  $m, n \rightarrow +\infty$ . Hence,  $(p_n)_{n \in \mathbb{N}}$  is a Cauchy sequence and, in turn,  $p_n \rightarrow p$  for some  $p \in \mathcal{H}$ . For every  $n \in \mathbb{N}$ ,  $(p_k)_{k \geq n}$  lies in  $C_n$  and hence  $p \in C_n$  since  $C_n$  is closed. Therefore,  $p \in \bigcap_{n \in \mathbb{N}} C_n$ .  $\square$

**Proposition 3.19** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then  $(\forall \lambda \in \mathbb{R}_+) P_C(P_Cx + \lambda(x - P_Cx)) = P_Cx$ .*

*Proof.* Let  $\lambda \in \mathbb{R}_+$  and  $y \in C$ . We derive from Theorem 3.14 that  $\langle y - P_Cx \mid (P_Cx + \lambda(x - P_Cx)) - P_Cx \rangle = \lambda \langle y - P_Cx \mid x - P_Cx \rangle \leq 0$  and in turn that  $P_C(P_Cx + \lambda(x - P_Cx)) = P_Cx$ .  $\square$

Projectors onto affine subspaces have additional properties.

**Corollary 3.20** *Let  $C$  be a closed affine subspace of  $\mathcal{H}$ . Then the following hold:*

(i) *Let  $x$  and  $p$  be in  $\mathcal{H}$ . Then  $p = P_Cx$  if and only if*

$$p \in C \quad \text{and} \quad (\forall y \in C)(\forall z \in C) \quad \langle y - z \mid x - p \rangle = 0. \quad (3.10)$$

(ii)  *$P_C$  is an affine operator.*

*Proof.* Let  $x \in \mathcal{H}$ .

(i): Let  $y \in C$  and  $z \in C$ . By (1.2),  $2P_Cx - y = 2P_Cx + (1 - 2)y \in C$  and (3.6) therefore yields  $\langle y - P_Cx \mid x - P_Cx \rangle \leq 0$  and  $-\langle y - P_Cx \mid x - P_Cx \rangle = \langle (2P_Cx - y) - P_Cx \mid x - P_Cx \rangle \leq 0$ . Altogether,  $\langle y - P_Cx \mid x - P_Cx \rangle = 0$ . Likewise,  $\langle z - P_Cx \mid x - P_Cx \rangle = 0$ . By subtraction,  $\langle y - z \mid x - P_Cx \rangle = 0$ . Conversely, it is clear that (3.10) implies the right-hand side of (3.6).

(ii): Let  $y \in \mathcal{H}$  and  $\alpha \in \mathbb{R}$ , and set  $z = \alpha x + (1 - \alpha)y$  and  $p = \alpha P_Cx + (1 - \alpha)P_Cy$ . We derive from (i) and (1.2) that  $p \in C$ . Now fix  $u$  and  $v$  in  $C$ . Then we also derive from (i) that  $\langle u - v \mid z - p \rangle = \alpha \langle u - v \mid x - P_Cx \rangle + (1 - \alpha) \langle u - v \mid y - P_Cy \rangle = 0$ . Altogether, it follows from (i) that  $p = P_Cz$ .  $\square$

**Example 3.21** Suppose that  $u$  is a nonzero vector in  $\mathcal{H}$ , let  $\eta \in \mathbb{R}$ , and set  $C = \{x \in \mathcal{H} \mid \langle x \mid u \rangle = \eta\}$ . Then

$$(\forall x \in \mathcal{H}) \quad P_C x = x + \frac{\eta - \langle x | u \rangle}{\|u\|^2} u, \quad (3.11)$$

and therefore  $d_C(x) = |\langle x | u \rangle - \eta|/\|u\|$ .

*Proof.* Check that (3.10) is satisfied to get (3.11).  $\square$

Next, we collect basic facts about projections onto linear subspaces.

**Corollary 3.22** *Let  $V$  be a closed linear subspace of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then the following hold:*

- (i)  $P_V x$  is characterized by  $P_V x \in V$  and  $x - P_V x \perp V$ .
- (ii)  $\|P_V x\|^2 = \langle P_V x | x \rangle$ .
- (iii)  $P_V \in \mathcal{B}(\mathcal{H})$ ,  $\|P_V\| = 1$  if  $V \neq \{0\}$ , and  $\|P_V\| = 0$  if  $V = \{0\}$ .
- (iv)  $V^{\perp\perp} = V$ .
- (v)  $P_{V^\perp} = \text{Id} - P_V$ .
- (vi)  $P_V^* = P_V$ .
- (vii)  $\|x\|^2 = \|P_V x\|^2 + \|P_{V^\perp} x\|^2 = d_V^2(x) + d_{V^\perp}^2(x)$ .

*Proof.* (i): A special case of Corollary 3.20(i).

(ii): We deduce from (i) that  $\langle x - P_V x | P_V x \rangle = 0$ .

(iii): Let  $y \in \mathcal{H}$  and  $\alpha \in \mathbb{R}$ , and set  $z = \alpha x + y$  and  $p = \alpha P_V x + P_V y$ . Then (i) yields  $p \in V$ ,  $x - P_V x \in V^\perp$ , and  $y - P_V y \in V^\perp$ ; hence  $z - p = \alpha(x - P_V x) + (y - P_V y) \in V^\perp$ . Altogether (i) yields  $p = P_V z$ . This shows the linearity of  $P_V$ . The other assertions follow from (2.18), (ii), and Cauchy–Schwarz.

(iv): The inclusion  $V \subset V^{\perp\perp}$  is clear. Conversely, suppose that  $x \in V^{\perp\perp}$ . It follows from (i) that  $\langle P_V x | x - P_V x \rangle = 0$ . Likewise, since  $x \in V^{\perp\perp}$  and  $x - P_V x \in V^\perp$  by (i), we have  $\langle x | x - P_V x \rangle = 0$ . Thus,  $\|x - P_V x\|^2 = \langle x - P_V x | x - P_V x \rangle = 0$ , i.e.,  $x = P_V x \in V$ . We conclude that  $V^{\perp\perp} \subset V$ .

(v): By (i),  $(\text{Id} - P_V)x$  lies in the closed linear subspace  $V^\perp$ . On the other hand,  $P_V x \in V \Rightarrow (\forall v \in V^\perp) \langle x - (\text{Id} - P_V)x | v \rangle = \langle P_V x | v \rangle = 0 \Rightarrow x - (\text{Id} - P_V)x \perp V^\perp$ . Altogether, in view of (i), we conclude that  $P_{V^\perp} x = (\text{Id} - P_V)x$ .

(vi): Take  $y \in \mathcal{H}$ . Then (v) yields  $\langle P_V x | y \rangle = \langle P_V x | P_V y + P_{V^\perp} y \rangle = \langle P_V x | P_V y \rangle = \langle P_V x + P_{V^\perp} x | P_V y \rangle = \langle x | P_V y \rangle$ .

(vii): By (i) and (v),  $\|x\|^2 = \|(x - P_V x) + P_V x\|^2 = \|x - P_V x\|^2 + \|P_V x\|^2 = \|P_{V^\perp} x\|^2 + \|P_V x\|^2 = \|x - P_V x\|^2 + \|x - P_{V^\perp} x\|^2 = d_V^2(x) + d_{V^\perp}^2(x)$ .  $\square$

**Proposition 3.23** *Let  $C$  be a nonempty subset of  $\mathcal{H}$ , let  $V = \overline{\text{span}} C$ , and let*

$$\Pi_C: \mathcal{H} \rightarrow 2^C: x \mapsto \{p \in C \mid \|x - p\| = d_C(x)\} \quad (3.12)$$

*be the set-valued projector onto  $C$ . Then  $\Pi_C = \Pi_C \circ P_V$ . Consequently,  $C$  is a proximal subset of  $\mathcal{H}$  if and only if  $C$  is a proximal subset of  $V$ .*

*Proof.* Let  $x \in \mathcal{H}$  and  $p \in C$ . In view of Corollary 3.22(vii)&(iii),

$$(\forall z \in C) \quad \|x - z\|^2 = \|P_V x - z\|^2 + \|P_{V^\perp} x\|^2. \quad (3.13)$$

Therefore  $p \in \Pi_C x \Leftrightarrow (\forall z \in C) \|x - p\|^2 \leq \|x - z\|^2 \Leftrightarrow (\forall z \in C) \|P_V x - p\|^2 + \|P_{V^\perp} x\|^2 \leq \|P_V x - z\|^2 + \|P_{V^\perp} x\|^2 \Leftrightarrow (\forall z \in C) \|P_V x - p\|^2 \leq \|P_V x - z\|^2 \Leftrightarrow p \in \Pi_C(P_V x)$ .  $\square$

An important application of Corollary 3.22 is the notion of least-squares solutions to linear equations.

**Definition 3.24** Let  $\mathcal{K}$  be a real Hilbert space, let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $y \in \mathcal{K}$ , and let  $x \in \mathcal{H}$ . Then  $x$  is a *least-squares solution* to the equation  $Tz = y$  if

$$\|Tx - y\| = \min_{z \in \mathcal{H}} \|Tz - y\|. \quad (3.14)$$

**Proposition 3.25** Let  $\mathcal{K}$  be a real Hilbert space, let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed, and let  $y \in \mathcal{K}$ . Then the equation  $Tz = y$  has at least one least-squares solution. Moreover, for every  $x \in \mathcal{H}$ , the following are equivalent:

- (i)  $x$  is a least-squares solution.
- (ii)  $Tx = P_{\text{ran } T} y$ .
- (iii)  $T^*Tx = T^*y$  (normal equation).

*Proof.* (i) $\Leftrightarrow$ (ii): Since  $\text{ran } T$  is a closed linear subspace, Theorem 3.14 asserts that  $P_{\text{ran } T} y$  is a well-defined point in  $\text{ran } T$ . Now fix  $x \in \mathcal{H}$ . Then

$$\begin{aligned} (\forall z \in \mathcal{H}) \quad \|Tx - y\| \leq \|Tz - y\| &\Leftrightarrow (\forall r \in \text{ran } T) \quad \|Tx - y\| \leq \|r - y\| \\ &\Leftrightarrow Tx = P_{\text{ran } T} y. \end{aligned} \quad (3.15)$$

Hence, the set of solutions to (3.14) is the nonempty set  $T^{-1}(\{P_{\text{ran } T} y\})$ .

(ii) $\Leftrightarrow$ (iii): We derive from Corollary 3.22(i) that

$$\begin{aligned} Tx = P_{\text{ran } T} y &\Leftrightarrow (\forall r \in \text{ran } T) \quad \langle r \mid Tx - y \rangle = 0 \\ &\Leftrightarrow (\forall z \in \mathcal{H}) \quad \langle Tz \mid Tx - y \rangle = 0 \\ &\Leftrightarrow (\forall z \in \mathcal{H}) \quad \langle z \mid T^*(Tx - y) \rangle = 0 \\ &\Leftrightarrow T^*Tx = T^*y, \end{aligned} \quad (3.16)$$

which completes the proof.  $\square$

The set of least-squares solutions  $\{x \in \mathcal{H} \mid T^*Tx = T^*y\}$  in Proposition 3.25 is a closed affine subspace. Hence, by Theorem 3.14, it possesses a unique minimal norm element, which will be denoted by  $T^\dagger y$ .

**Definition 3.26** Let  $\mathcal{K}$  be a real Hilbert space, let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed, and, for every  $y \in \mathcal{K}$ , set  $C_y = \{x \in \mathcal{H} \mid T^*Tx = T^*y\}$ . The *generalized* (or *Moore–Penrose*) *inverse* of  $T$  is  $T^\dagger: \mathcal{K} \rightarrow \mathcal{H}: y \mapsto P_{C_y} 0$ .

**Example 3.27** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $T^*T$  is invertible. Then  $T^\dagger = (T^*T)^{-1}T^*$ .

**Proposition 3.28** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed. Then the following hold:

- (i)  $(\forall y \in \mathcal{K}) \{x \in \mathcal{H} \mid T^*Tx = T^*y\} \cap (\ker T)^\perp = \{T^\dagger y\}$ .
- (ii)  $P_{\text{ran } T} = TT^\dagger$ .
- (iii)  $P_{\ker T} = \text{Id} - T^*T^\dagger$ .
- (iv)  $T^\dagger \in \mathcal{B}(\mathcal{K}, \mathcal{H})$ .
- (v)  $\text{ran } T^\dagger = \text{ran } T^*$ .
- (vi)  $P_{\text{ran } T^\dagger} = T^\dagger T$ .

*Proof.* (i): Fix  $y \in \mathcal{K}$  and  $z \in C_y$ . Then it follows from Fact 2.18(v) that  $C_y = \{x \in \mathcal{H} \mid T^*Tx = T^*Ty\} = \{x \in \mathcal{H} \mid x - z \in \ker T^*T = \ker T\} = z + \ker T$ . Hence, since  $C_y$  is a closed affine subspace, it follows from (3.6) and (3.10) that

$$\begin{aligned} z = T^\dagger y &\Leftrightarrow z = P_{C_y} 0 \\ &\Leftrightarrow (\forall x \in C_y) \langle x - z \mid 0 - z \rangle = 0 \\ &\Leftrightarrow (\forall x \in z + \ker T) \langle x - z \mid z \rangle = 0 \\ &\Leftrightarrow z \perp \ker T. \end{aligned} \tag{3.17}$$

(ii): Fix  $y \in \mathcal{K}$ . Since  $T^\dagger y$  is a least-squares solution, Proposition 3.25(ii) yields  $T(T^\dagger y) = P_{\text{ran } T} y$ .

(iii): It follows from Corollary 3.22(v), Fact 2.18(iii), Fact 2.19, and (ii) that  $P_{\ker T} = \text{Id} - P_{(\ker T)^\perp} = \text{Id} - \overline{P_{\text{ran } T^*}} = \text{Id} - P_{\text{ran } T^*} = \text{Id} - T^*T^\dagger$ .

(iv): Fix  $y_1$  and  $y_2$  in  $\mathcal{K}$ , fix  $\alpha \in \mathbb{R}$ , and set  $x = \alpha T^\dagger y_1 + T^\dagger y_2$ . To establish the linearity of  $T^\dagger$ , we must show that  $x = T^\dagger(\alpha y_1 + y_2)$ , i.e., by (i), that  $T^*Tx = T^*(\alpha y_1 + y_2)$  and that  $x \perp \ker T$ . Since, by (i),  $T^*TT^\dagger y_1 = T^*y_1$  and  $T^*TT^\dagger y_2 = T^*y_2$ , it follows immediately from the linearity of  $T^*T$  that  $T^*Tx = \alpha T^*TT^\dagger y_1 + T^*TT^\dagger y_2 = \alpha T^*y_1 + T^*y_2$ . On the other hand, since (i) also implies that  $T^\dagger y_1$  and  $T^\dagger y_2$  lie in the linear subspace  $(\ker T)^\perp$ , so does their linear combination  $x$ . This shows the linearity of  $T^\dagger$ . It remains to show that  $T^\dagger$  is bounded. It follows from (i) that  $(\forall y \in \mathcal{K}) T^\dagger y \in (\ker T)^\perp$ . Hence, Fact 2.19 asserts that there exists  $\alpha \in \mathbb{R}_{++}$  such that  $(\forall y \in \mathcal{K}) \|TT^\dagger y\| \geq \alpha \|T^\dagger y\|$ . Consequently, we derive from (ii) and Corollary 3.22(iii) that

$$(\forall y \in \mathcal{K}) \quad \alpha \|T^\dagger y\| \leq \|TT^\dagger y\| = \|P_{\text{ran } T} y\| \leq \|y\|, \tag{3.18}$$

which establishes the boundedness of  $T^\dagger$ .

(v): Fact 2.18(iii) and Fact 2.19 yield  $(\ker T)^\perp = \overline{\text{ran } T^*} = \text{ran } T^*$ . Hence (i) implies that  $\text{ran } T^\dagger \subset \text{ran } T^*$ . Conversely, take  $x \in \text{ran } T^*$  and set  $y = Tx$ . Then  $T^*y = T^*Tx \in \text{ran } T^* = (\ker T)^\perp$  and (i) yields  $x = T^\dagger y \in \text{ran } T^\dagger$ .

(vi): By (ii) and (v),  $TT^\dagger T = P_{\text{ran } T} T = T = T(P_{\ker T} + P_{(\ker T)^\perp}) = TP_{\text{ran } T^*} = TP_{\text{ran } T^\dagger}$ . Hence  $T(T^\dagger T - P_{\text{ran } T^\dagger}) = 0$ , i.e.,  $\text{ran}(T^\dagger T - P_{\text{ran } T^\dagger}) \subset$

$\ker T$ . On the other hand,  $\text{ran}(T^\dagger T - P_{\text{ran } T^\dagger}) \subset \text{ran } T^\dagger = \text{ran } T^* = (\ker T)^\perp$ . Altogether,  $\text{ran}(T^\dagger T - P_{\text{ran } T^\dagger}) \subset \ker T \cap (\ker T)^\perp = \{0\}$ , which implies that  $T^\dagger T = P_{\text{ran } T^\dagger}$ .  $\square$

**Proposition 3.29 (Moore–Desoer–Whalen)** *Let  $\mathcal{K}$  be a real Hilbert space, let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed, and let  $\tilde{T} \in \mathcal{B}(\mathcal{K}, \mathcal{H})$  be such that  $\text{ran } \tilde{T}$  is closed. Then the following are equivalent:*

- (i)  $\tilde{T} = T^\dagger$ .
- (ii)  $T\tilde{T} = P_{\text{ran } T}$  and  $\tilde{T}T = P_{\text{ran } \tilde{T}}$ .
- (iii)  $\tilde{T}|_{(\ker T)^\perp} = \text{Id}$  and  $\tilde{T}|_{(\text{ran } T)^\perp} = 0$ .

*Proof.* (i) $\Rightarrow$ (ii): See Proposition 3.28(ii)&(vi).

(ii) $\Rightarrow$ (iii): Since  $T = P_{\text{ran } T}T = (T\tilde{T})T$ , we have  $T(\text{Id} - \tilde{T}T) = 0$  and thus  $\text{ran}(\text{Id} - \tilde{T}T) \subset \ker T$ . Hence, for every  $x \in (\ker T)^\perp$ ,  $\|x\|^2 \geq \|P_{\text{ran } \tilde{T}}x\|^2 = \|\tilde{T}Tx\|^2 = \|\tilde{T}Tx - x\|^2 + \|x\|^2$  and therefore  $\tilde{T}Tx = x$ . Furthermore,  $\tilde{T} = P_{\text{ran } \tilde{T}}\tilde{T} = (\tilde{T}T)\tilde{T} = \tilde{T}(T\tilde{T}) = \tilde{T}P_{\text{ran } T}$ , which implies that  $\tilde{T}|_{(\text{ran } T)^\perp} = 0$ .

(iii) $\Rightarrow$ (i): Take  $y \in \mathcal{K}$ , and set  $y_1 = P_{\text{ran } T}y$  and  $y_2 = P_{(\text{ran } T)^\perp}y$ . Then there exists  $x_1 \in (\ker T)^\perp$  such that  $y_1 = Tx_1$ . Hence

$$\tilde{T}y = \tilde{T}y_1 + \tilde{T}y_2 = \tilde{T}y_1 = \tilde{T}Tx_1 = x_1. \quad (3.19)$$

It follows that  $T\tilde{T}y = Tx_1 = y_1 = P_{\text{ran } T}y$  and Proposition 3.25 yields  $T^*T\tilde{T}y = T^*y$ . Since (3.19) asserts that  $\tilde{T}y = x_1 \in (\ker T)^\perp$ , we deduce from Proposition 3.28(i) that  $\tilde{T}y = T^\dagger y$ .  $\square$

**Corollary 3.30** *Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed. Then the following hold:*

- (i)  $T^{\dagger\dagger} = T$ .
- (ii)  $T^{\dagger*} = T^{*\dagger}$ .

*Proof.* (i): Proposition 3.28(v) asserts that  $\text{ran } T^* = \text{ran } T^\dagger$  is closed. Hence, it follows from Proposition 3.29 that

$$TT^\dagger = P_{\text{ran } T} \quad \text{and} \quad T^\dagger T = P_{\text{ran } T^\dagger}. \quad (3.20)$$

Combining Proposition 3.29 applied to  $T^\dagger$  with (3.20) yields  $T = T^{\dagger\dagger}$ .

(ii): It follows from (3.20), Corollary 3.22(vi), and Proposition 3.28(v) that  $T^{\dagger*}T^* = P_{\text{ran } T}^* = P_{\text{ran } T} = P_{\text{ran } (T^*)^*} = P_{\text{ran } (T^*)^\dagger}$  and that  $T^*T^{\dagger*} = P_{\text{ran } T^\dagger} = P_{\text{ran } T^*}$ . Therefore, by Proposition 3.29,  $T^{\dagger*} = T^{*\dagger}$ .  $\square$

### 3.3 Topological Properties

Since a Hilbert space is a metric space, the notions of closedness and sequential closedness coincide for the strong topology (see Section 1.12). The

following example illustrates the fact that more care is required for the weak topology.

**Example 3.31** Suppose that  $\mathcal{H}$  is infinite-dimensional, let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ , and set  $C = \{\alpha_n e_n\}_{n \in \mathbb{N}}$ , where  $(\alpha_n)_{n \in \mathbb{N}}$  is a sequence in  $[1, +\infty[$  such that  $\sum_{n \in \mathbb{N}} 1/\alpha_n^2 = +\infty$  and  $\alpha_n \uparrow +\infty$  (e.g.,  $(\forall n \in \mathbb{N}) \alpha_n = \sqrt{n+1}$ ). Then  $C$  is closed and weakly sequentially closed, but not weakly closed. In fact, 0 belongs to the weak closure of  $C$  but not to  $C$ .

*Proof.* Since the distance between two distinct points in  $C$  is at least  $\sqrt{2}$ , every Cauchy sequence in  $C$  is eventually constant and  $C$  is therefore closed. It follows from Lemma 2.38 that every weakly convergent sequence in  $C$  is bounded and hence, since  $\alpha_n \uparrow +\infty$ , eventually constant. Thus,  $C$  is weakly sequentially closed. Now let  $V$  be a weak neighborhood of 0. Then there exist  $\varepsilon \in \mathbb{R}_{++}$  and a finite family  $(u_i)_{i \in I}$  in  $\mathcal{H}$  such that

$$U = \{x \in \mathcal{H} \mid (\forall i \in I) |\langle x \mid u_i \rangle| < \varepsilon\} \subset V. \quad (3.21)$$

Set  $(\forall n \in \mathbb{N}) \zeta_n = \sum_{i \in I} |\langle u_i \mid e_n \rangle|$ . We have  $(\forall i \in I) (\langle u_i \mid e_n \rangle)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N})$ . Consequently, the sequence  $(\zeta_n)_{n \in \mathbb{N}}$  belongs to the vector space  $\ell^2(\mathbb{N})$ . Hence  $\sum_{n \in \mathbb{N}} \zeta_n^2 < +\infty = \sum_{n \in \mathbb{N}} 1/\alpha_n^2$ , and the set  $\mathbb{M} = \{n \in \mathbb{N} \mid \zeta_n < \varepsilon/\alpha_n\}$  therefore contains infinitely many elements. Since  $(\forall i \in I)(\forall n \in \mathbb{M}) |\langle u_i \mid \alpha_n e_n \rangle| = \alpha_n |\langle u_i \mid e_n \rangle| \leq \alpha_n \zeta_n < \varepsilon$ , we deduce that

$$\{\alpha_n e_n\}_{n \in \mathbb{M}} \subset U \subset V. \quad (3.22)$$

Thus, every weak neighborhood of 0 contains elements from  $C$ . Therefore, 0 belongs to the weak closure of  $C$ . Finally, it is clear that  $0 \notin C$ .  $\square$

Next, we show that the distinctions illustrated in Example 3.31 among the various types of closure disappear for convex sets.

**Theorem 3.32** *Let  $C$  be a convex subset of  $\mathcal{H}$ . Then the following are equivalent:*

- (i)  $C$  is weakly sequentially closed.
- (ii)  $C$  is sequentially closed.
- (iii)  $C$  is closed.
- (iv)  $C$  is weakly closed.

*Proof.* Suppose that  $C$  is nonempty (otherwise the conclusion is clear).

- (i)  $\Rightarrow$  (ii): This follows from Corollary 2.42.
- (ii)  $\Leftrightarrow$  (iii): Since  $\mathcal{H}$  is a metric space, see Section 1.12.
- (iii)  $\Rightarrow$  (iv): Take a net  $(x_a)_{a \in A}$  in  $C$  that converges weakly to some point  $x \in \mathcal{H}$ . Then (3.6) yields

$$(\forall a \in A) \quad \langle x_a - P_C x \mid x - P_C x \rangle \leq 0. \quad (3.23)$$

Since  $x_a \rightharpoonup x$ , passing to the limit in (3.23) yields  $\|x - P_C x\|^2 \leq 0$ , hence  $x \in C$ .

(iv) $\Rightarrow$ (i): Clear.  $\square$

**Theorem 3.33** *Let  $C$  be a bounded closed convex subset of  $\mathcal{H}$ . Then  $C$  is weakly compact and weakly sequentially compact.*

*Proof.* Theorem 3.32 asserts that  $C$  is weakly closed. In turn, it follows from Lemma 2.29 that  $C$  is weakly compact. Therefore, by Fact 2.30,  $C$  is weakly sequentially compact.  $\square$

In infinite-dimensional Hilbert spaces, the sum of two closed linear subspaces may fail to be closed.

**Example 3.34** Suppose that  $\mathcal{H}$  is infinite-dimensional and let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ . Set

$$C = \overline{\text{span}}\{e_{2n}\}_{n \in \mathbb{N}} \quad \text{and} \quad D = \overline{\text{span}}\{\cos(\theta_n)e_{2n} + \sin(\theta_n)e_{2n+1}\}_{n \in \mathbb{N}}, \quad (3.24)$$

where  $(\theta_n)_{n \in \mathbb{N}}$  is a sequence in  $]0, \pi/2]$  such that  $\sum_{n \in \mathbb{N}} \sin^2(\theta_n) < +\infty$ . Then  $C \cap D = \{0\}$  and  $C + D$  is a linear subspace of  $\mathcal{H}$  that is not closed.

*Proof.* It follows from (3.24) that the elements of  $C$  and  $D$  are of the form  $\sum_{n \in \mathbb{N}} \gamma_n e_{2n}$  and  $\sum_{n \in \mathbb{N}} \delta_n (\cos(\theta_n)e_{2n} + \sin(\theta_n)e_{2n+1})$ , respectively, where  $(\gamma_n)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N})$  and  $(\delta_n)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N})$ . It is straightforward that  $C \cap D = \{0\}$ . Now let

$$x = \sum_{n \in \mathbb{N}} \sin(\theta_n) e_{2n+1} \quad (3.25)$$

and observe that  $x \in \overline{C + D}$ . Assume that  $x \in C + D$ . Then there exist sequences  $(\gamma_n)_{n \in \mathbb{N}}$  and  $(\delta_n)_{n \in \mathbb{N}}$  in  $\ell^2(\mathbb{N})$  such that

$$(\forall n \in \mathbb{N}) \quad 0 = \gamma_n + \delta_n \cos(\theta_n) \quad \text{and} \quad \sin(\theta_n) = \delta_n \sin(\theta_n). \quad (3.26)$$

Thus  $\delta_n \equiv 1$  and  $\gamma_n = -\cos(\theta_n) \rightarrow -1$ , which is impossible since  $(\gamma_n)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N})$ .  $\square$

**Proposition 3.35** *Let  $C$  be a convex subset of  $\mathcal{H}$ . Then*

$$(\forall x \in \text{int } C)(\forall y \in \overline{C}) \quad [x, y[ \subset \text{int } C. \quad (3.27)$$

*Proof.* Fix  $x \in \text{int } C$  and  $y \in \overline{C}$ . If  $x = y$ , the conclusion is trivial. Now assume that  $x \neq y$  and fix  $z \in [x, y[$ , say  $z = \alpha x + (1 - \alpha)y$ , where  $\alpha \in ]0, 1]$ . Since  $x \in \text{int } C$ , there exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $B(x; \varepsilon(2 - \alpha)/\alpha) \subset C$ . On the other hand, since  $y \in \overline{C}$ , we have  $y \in C + B(0; \varepsilon)$ . Therefore, by convexity,

$$\begin{aligned} B(z; \varepsilon) &= \alpha x + (1 - \alpha)y + B(0; \varepsilon) \\ &\subset \alpha x + (1 - \alpha)(C + B(0; \varepsilon)) + B(0; \varepsilon) \\ &= \alpha B(x; \varepsilon(2 - \alpha)/\alpha) + (1 - \alpha)C \end{aligned}$$



$$\begin{aligned}
&\subset \alpha C + (1 - \alpha)C \\
&= C.
\end{aligned} \tag{3.28}$$

Hence  $z \in \text{int } C$ .  $\square$

**Proposition 3.36** *Let  $C$  be a convex subset of  $\mathcal{H}$ . Then the following hold:*

- (i)  $\overline{C}$  is convex.
- (ii)  $\text{int } C$  is convex.
- (iii) Suppose that  $\text{int } C \neq \emptyset$ . Then  $\text{int } C = \text{int } \overline{C}$  and  $\overline{C} = \overline{\text{int } C}$ .

*Proof.* (i): Take  $x$  and  $y$  in  $\overline{C}$ , and  $\alpha \in ]0, 1[$ . Then there exist sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}}$  in  $C$  such that  $x_n \rightarrow x$  and  $y_n \rightarrow y$ . By convexity,  $C \ni \alpha x_n + (1 - \alpha)y_n \rightarrow \alpha x + (1 - \alpha)y$  and, therefore,  $\alpha x + (1 - \alpha)y \in \overline{C}$ .

(ii): Take  $x$  and  $y$  in  $\text{int } C$ . Then, since  $y \in \overline{C}$ , Proposition 3.35 implies that  $]x, y[ \subset ]x, y[ \subset \text{int } C$ .

(iii): It is clear that  $\text{int } C \subset \text{int } \overline{C}$ . Conversely, let  $y \in \text{int } \overline{C}$ . Then we can find  $\varepsilon \in \mathbb{R}_{++}$  such that  $B(y; \varepsilon) \subset \overline{C}$ . Now take  $x \in \text{int } C$  and  $\alpha \in \mathbb{R}_{++}$  such that  $x \neq y$  and  $y + \alpha(y - x) \in B(y; \varepsilon)$ . Then Proposition 3.35 implies that  $y \in ]x, y + \alpha(y - x)[ \subset \text{int } C$  and we deduce that  $\text{int } \overline{C} \subset \text{int } C$ . Thus,  $\text{int } C = \text{int } \overline{C}$ . It is also clear that  $\overline{\text{int } C} \subset \overline{C}$ . Now take  $x \in \text{int } C$ ,  $y \in \overline{C}$ , and define  $(\forall \alpha \in ]0, 1[)$   $y_\alpha = \alpha x + (1 - \alpha)y$ . Proposition 3.35 implies that  $(y_\alpha)_{\alpha \in ]0, 1[}$  lies in  $]x, y[ \subset \text{int } C$ . Hence  $y = \lim_{\alpha \downarrow 0} y_\alpha \in \overline{\text{int } C}$ . Therefore  $\overline{C} \subset \overline{\text{int } C}$ , and we conclude that  $\overline{C} = \overline{\text{int } C}$ .  $\square$

### 3.4 Separation

**Definition 3.37** Let  $C$  and  $D$  be subsets of  $\mathcal{H}$ . Then  $C$  and  $D$  are *separated* if (see Figure 3.2)

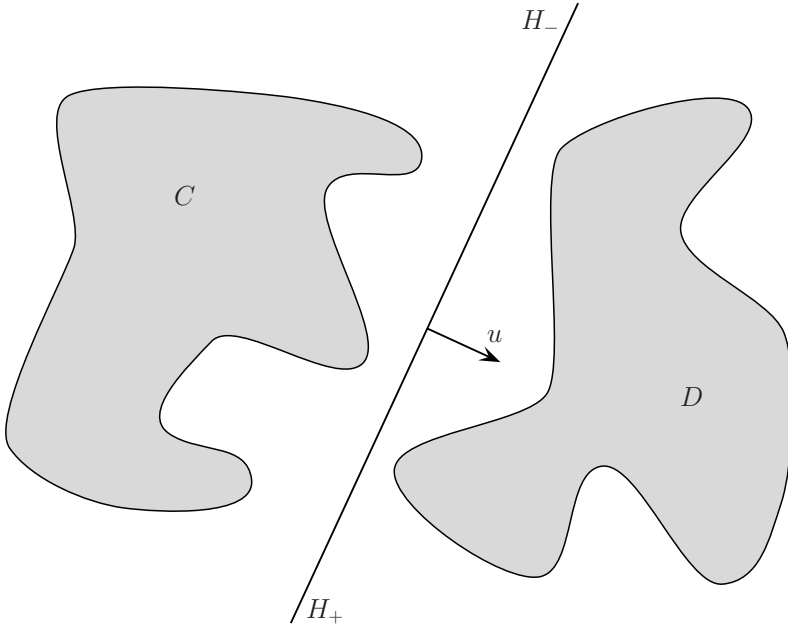
$$(\exists u \in \mathcal{H} \setminus \{0\}) \quad \sup \langle C \mid u \rangle \leq \inf \langle D \mid u \rangle, \tag{3.29}$$

and *strongly separated* if the above inequality is strict. Moreover, a point  $x \in \mathcal{H}$  is separated from  $D$  if the sets  $\{x\}$  and  $D$  are separated; likewise,  $x$  is strongly separated from  $D$  if  $\{x\}$  and  $D$  are strongly separated.

**Theorem 3.38** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H} \setminus C$ . Then  $x$  is strongly separated from  $C$ .*

*Proof.* Set  $u = x - P_C x$  and fix  $y \in C$ . Then  $u \neq 0$  and (3.6) yields  $\langle y - x + u \mid u \rangle \leq 0$ , i.e.,  $\langle y - x \mid u \rangle \leq -\|u\|^2$ . Hence  $\sup \langle C - x \mid u \rangle \leq -\|u\|^2 < 0$ .  $\square$

**Corollary 3.39** *Let  $C$  and  $D$  be nonempty subsets of  $\mathcal{H}$  such that  $C \cap D = \emptyset$  and  $C - D$  is closed and convex. Then  $C$  and  $D$  are strongly separated.*



**Fig. 3.2** The sets  $C$  and  $D$  are separated: by (3.29), there exist  $u \in \mathcal{H} \setminus \{0\}$  and  $\eta \in \mathbb{R}$  such that  $C$  is contained in the half-space  $H_- = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \eta\}$  and  $D$  is contained in the half-space  $H_+ = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \geq \eta\}$ .

*Proof.* Since  $0 \notin C - D$ , Theorem 3.38 asserts that the vector 0 is strongly separated from  $C - D$ . However, it follows from Definition 3.37 that  $C$  and  $D$  are strongly separated if and only if 0 is strongly separated from  $C - D$ .  $\square$

**Corollary 3.40** *Let  $C$  and  $D$  be nonempty closed convex subsets of  $\mathcal{H}$  such that  $C \cap D = \emptyset$  and  $D$  is bounded. Then  $C$  and  $D$  are strongly separated.*

*Proof.* In view of Corollary 3.39, it is enough to show that  $C - D$  is closed and convex. The convexity of  $C - D$  follows from Proposition 3.6(ii). To show that  $C - D$  is closed, take a convergent sequence in  $C - D$ , say  $x_n - y_n \rightarrow z$ , where  $(x_n)_{n \in \mathbb{N}}$  lies in  $C$ ,  $(y_n)_{n \in \mathbb{N}}$  lies in  $D$ , and  $z \in \mathcal{H}$ . Since  $D$  is weakly sequentially compact by Theorem 3.33, there exists a subsequence  $(y_{k_n})_{n \in \mathbb{N}}$  converging weakly to a point  $y \in D$ . Therefore  $x_{k_n} \rightharpoonup z + y$ . However, since  $C$  is weakly sequentially closed by Theorem 3.32, we have  $z + y \in C$  and, in turn,  $z \in C - D$ .  $\square$

We conclude this section by pointing out that separation of nonintersecting closed convex sets may not be achievable.

**Example 3.41** Suppose that  $\mathcal{H}$  is infinite-dimensional. Then there exist two closed affine subspaces that do not intersect and that are not separated.

*Proof.* Let  $C$  and  $D$  be as in Example 3.34 and fix  $z \in \overline{C+D} \setminus (C+D)$ . Define two closed affine subspaces by  $U = C + (C+D)^\perp$  and  $V = z + D$ . Then  $U \cap V = \emptyset$  and, since Corollary 3.22(v) implies that  $\overline{C+D} + (C+D)^\perp = \mathcal{H}$ ,  $U - V = (C+D) - z + (C+D)^\perp$  is dense in  $\mathcal{H}$ . Now suppose that  $u \in \mathcal{H}$  satisfies  $\inf \langle U | u \rangle \geq \sup \langle V | u \rangle$ . Then  $\inf \langle U - V | u \rangle \geq 0$ , and hence  $\inf \langle \mathcal{H} | u \rangle \geq 0$ . This implies that  $u = 0$  and therefore that the separation of  $U$  and  $V$  is impossible.  $\square$

## Exercises

**Exercise 3.1** Prove Proposition 3.4.

**Exercise 3.2** Let  $I$  be a totally ordered finite index set and, for every  $i \in I$ , let  $C_i$  be a subset of a real Hilbert space  $\mathcal{H}_i$ . Show that  $\text{conv}(\times_{i \in I} C_i) = \times_{i \in I} \text{conv}(C_i)$ .

**Exercise 3.3** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$  and let  $\alpha$  and  $\beta$  be two positive real numbers. Show that  $\alpha C + \beta C = (\alpha + \beta)C$  and that this property fails if  $C$  is not convex.

**Exercise 3.4** Let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of convex subsets of  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) C_n \subset C_{n+1}$ , and set  $C = \bigcup_{n \in \mathbb{N}} C_n$ .

- (i) Show that  $C$  is convex.
- (ii) Find an example in which the sets  $(C_n)_{n \in \mathbb{N}}$  are closed and  $C$  is not closed.

**Exercise 3.5** A subset  $C$  of  $\mathcal{H}$  is *midpoint convex* if

$$(\forall x \in C)(\forall y \in C) \quad \frac{x+y}{2} \in C. \quad (3.30)$$

- (i) Suppose that  $\mathcal{H} = \mathbb{R}$  and let  $C$  be the set of rational numbers. Show that  $C$  is midpoint convex but not convex.
- (ii) Suppose that  $C$  is closed and midpoint convex. Show that  $C$  is convex.

**Exercise 3.6** Let  $C$  a subset of  $\mathcal{H}$ . Show that  $\overline{\text{conv } C} = \overline{\text{conv } \overline{C}}$ .

**Exercise 3.7** Consider the setting of Proposition 3.5 and suppose that, in addition,  $T$  is continuous and  $C$  and  $D$  are closed. Show that  $T^{-1}(D)$  is closed and find an example in which  $T(C)$  is not closed.

**Exercise 3.8** Consider the setting of Proposition 3.6 and suppose that, in addition, each  $C_i$  is closed. Show that the set  $\times_{i=1}^m C_i$  in item (i) is closed and find an example in which the set  $\sum_{i=1}^m \alpha_i C_i$  in item (ii) is not closed.

**Exercise 3.9** Let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ , let  $V = \overline{\text{span}}\{e_n\}_{n \in \mathbb{N}}$ , and let  $x \in \mathcal{H}$ . Show that  $P_V x = \sum_{n \in \mathbb{N}} \langle x | e_n \rangle e_n$  and that  $d_V(x) = \sqrt{\|x\|^2 - \sum_{n \in \mathbb{N}} |\langle x | e_n \rangle|^2}$ .

**Exercise 3.10** Let  $V$  be a closed linear subspace of  $\mathcal{H}$ . Show that  $P_V^\dagger = P_V$ .

**Exercise 3.11 (Penrose)** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed. Show the following:  $TT^\dagger = (TT^\dagger)^*$ ,  $T^\dagger T = (T^\dagger T)^*$ ,  $TT^\dagger T = T$ , and  $T^\dagger TT^\dagger = T^\dagger$ .

**Exercise 3.12** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed. Show that  $\ker T^\dagger = \ker T^*$ .

**Exercise 3.13** Let  $\mathcal{K}$  be a real Hilbert space and let  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } T$  is closed. Show that  $T^\dagger = P_{\text{ran } T^*} \circ T^{-1} \circ P_{\text{ran } T}$ , where all operators are understood in the set-valued sense.

**Exercise 3.14** Let  $C$  be a nonempty subset of  $\mathcal{H}$  consisting of vectors of equal norm, let  $x \in \mathcal{H}$  and  $p \in C$ . Prove that  $p$  is a projection of  $x$  onto  $C$  if and only if  $\langle x | p \rangle = \sup \langle x | C \rangle$ .

**Exercise 3.15** Provide a subset  $C$  of  $\mathcal{H}$  that is not weakly closed and such that, for every  $n \in \mathbb{N}$ ,  $C \cap B(0; n)$  is weakly closed. Compare with Lemma 1.39 and with Exercise 2.6.

**Exercise 3.16** Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) C \cap B(0; n)$  is weakly closed. Show that  $C$  is weakly closed. Compare with Exercise 3.15.

**Exercise 3.17** Show that the closed convex hull of a weakly compact subset of  $\mathcal{H}$  is weakly compact. In contrast, provide an example of a compact subset of  $\mathcal{H}$  the convex hull of which is not closed.

**Exercise 3.18** Suppose that  $\mathcal{H}$  is finite-dimensional, let  $x_0 \in \mathcal{H}$ , and let  $\rho \in \mathbb{R}_{++}$ . Show that there exists a finite family  $(y_i)_{i \in I}$  in  $\mathcal{H}$  such that  $B(x_0; \rho) \subset \text{conv}\{y_i\}_{i \in I}$ .

**Exercise 3.19** Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty subset of  $\mathcal{H}$ . Suppose that, for every  $y \in C$ ,  $(\|x_n - y\|)_{n \in \mathbb{N}}$  converges. Show that this property remains true for every  $y \in \overline{\text{conv}} C$ . Use this result to obtain an extension of Lemma 2.39.

**Exercise 3.20 (Rådström's cancellation)** Let  $C$ ,  $D$ , and  $E$  be subsets of  $\mathcal{H}$ . Suppose that  $D$  is nonempty and bounded, that  $E$  is closed and convex, and that  $C + D \subset E + D$ . Show that  $C \subset E$  and that this inclusion fails if  $E$  is not convex.

**Exercise 3.21** Find a nonempty compact subset  $C$  of  $\mathcal{H}$  and a point  $x \in \mathcal{H} \setminus C$  that cannot be strongly separated from  $C$ . Compare with Corollary 3.39.

**Exercise 3.22** Find two nonempty closed convex subsets  $C$  and  $D$  of  $\mathcal{H}$  such that  $C \cap D = \emptyset$ , and  $C$  and  $D$  are not strongly separated. Compare with Corollary 3.40.

# Chapter 4

## Convexity and Nonexpansiveness

Nonexpansive operators are Lipschitz continuous operators with Lipschitz constant 1. They play a central role in applied mathematics, because many problems in nonlinear analysis reduce to finding fixed points of nonexpansive operators. In this chapter, we discuss nonexpansiveness and several variants. The properties of the fixed point sets of nonexpansive operators are investigated, in particular in terms of convexity.

### 4.1 Nonexpansive Operators

**Definition 4.1** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$ . Then  $T$  is

(i) *firmly nonexpansive* if

$$(\forall x \in D)(\forall y \in D) \quad \|Tx - Ty\|^2 + \|(\text{Id} - T)x - (\text{Id} - T)y\|^2 \leq \|x - y\|^2; \quad (4.1)$$

(ii) *nonexpansive* if it is Lipschitz continuous with constant 1, i.e.,

$$(\forall x \in D)(\forall y \in D) \quad \|Tx - Ty\| \leq \|x - y\|; \quad (4.2)$$

(iii) *quasinonexpansive* if

$$(\forall x \in D)(\forall y \in \text{Fix } T) \quad \|Tx - y\| \leq \|x - y\|; \quad (4.3)$$

(iv) and *strictly quasinonexpansive* if

$$(\forall x \in D \setminus \text{Fix } T)(\forall y \in \text{Fix } T) \quad \|Tx - y\| < \|x - y\|. \quad (4.4)$$

It is clear that firm nonexpansiveness implies nonexpansiveness, which itself implies quasinonexpansiveness (as will be seen in Example 4.9, these im-

plications are strict). In addition, firm nonexpansiveness implies strict quasi nonexpansiveness, which implies quasinonexpansiveness.

**Proposition 4.2** *Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$ . Then the following are equivalent:*

- (i)  $T$  is firmly nonexpansive.
- (ii)  $\text{Id} - T$  is firmly nonexpansive.
- (iii)  $2T - \text{Id}$  is nonexpansive.
- (iv)  $(\forall x \in D)(\forall y \in D) \|Tx - Ty\|^2 \leq \langle x - y \mid Tx - Ty \rangle$ .
- (v)  $(\forall x \in D)(\forall y \in D) 0 \leq \langle Tx - Ty \mid (\text{Id} - T)x - (\text{Id} - T)y \rangle$ .
- (vi)  $(\forall x \in D)(\forall y \in D)(\forall \alpha \in [0, 1]) \|Tx - Ty\| \leq \|\alpha(x - y) + (1 - \alpha)(Tx - Ty)\|$ .

*Proof.* (i) $\Leftrightarrow$ (ii): See (4.1).

(i) $\Leftrightarrow$ (iii): Fix  $x$  and  $y$  in  $D$ , and define  $R = 2T - \text{Id}$ ,  $\mu = \|Tx - Ty\|^2 + \|(\text{Id} - T)x - (\text{Id} - T)y\|^2 - \|x - y\|^2$ , and  $\nu = \|Rx - Ry\|^2 - \|x - y\|^2$ . Corollary 2.14 implies that  $\|Rx - Ry\|^2 = \|2(Tx - Ty) + (1 - 2)(x - y)\|^2 = 2\|Tx - Ty\|^2 - \|x - y\|^2 + 2\|(\text{Id} - T)x - (\text{Id} - T)y\|^2$ ; hence  $\nu = 2\mu$ . Thus,  $R$  is nonexpansive  $\Leftrightarrow \nu \leq 0 \Leftrightarrow \mu \leq 0 \Leftrightarrow T$  is firmly nonexpansive.

(i) $\Leftrightarrow$ (iv): Write  $\|(\text{Id} - T)x - (\text{Id} - T)y\|^2 = \|x - y\|^2 + \|Tx - Ty\|^2 - 2\langle x - y \mid Tx - Ty \rangle$  in (4.1).

(iv) $\Leftrightarrow$ (v): Clear.

(v) $\Leftrightarrow$ (vi): Use Lemma 2.12(i). □

**Corollary 4.3** *Let  $T \in \mathcal{B}(\mathcal{H})$ . Then the following are equivalent:*

- (i)  $T$  is firmly nonexpansive.
- (ii)  $\|2T - \text{Id}\| \leq 1$ .
- (iii)  $(\forall x \in \mathcal{H}) \|Tx\|^2 \leq \langle x \mid Tx \rangle$ .
- (iv)  $T^*$  is firmly nonexpansive.
- (v)  $T + T^* - 2T^*T$  is positive.

*Proof.* (i) $\Leftrightarrow$ (ii) $\Leftrightarrow$ (iii): This follows from the equivalences (i) $\Leftrightarrow$ (iii) $\Leftrightarrow$ (iv) in Proposition 4.2.

(i) $\Leftrightarrow$ (iv): Since  $\|2T^* - \text{Id}\| = \|(2T - \text{Id})^*\| = \|2T - \text{Id}\|$ , this follows from the equivalence (i) $\Leftrightarrow$ (ii).

(iii) $\Leftrightarrow$ (v): Indeed, (iii)  $\Leftrightarrow (\forall x \in \mathcal{H}) \langle x \mid (T - T^*T)x \rangle \geq 0 \Leftrightarrow (\forall x \in \mathcal{H}) \langle x \mid (T - T^*T)x + (T - T^*T)^*x \rangle \geq 0 \Leftrightarrow (\forall x \in \mathcal{H}) \langle x \mid (T + T^* - 2T^*T)x \rangle \geq 0 \Leftrightarrow$  (v). □

**Definition 4.4** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , and let  $\beta \in \mathbb{R}_{++}$ . Then  $T$  is  $\beta$ -cocoercive (or  $\beta$ -inverse strongly monotone) if  $\beta T$  is firmly nonexpansive, i.e.,

$$(\forall x \in D)(\forall y \in D) \quad \langle x - y \mid Tx - Ty \rangle \geq \beta \|Tx - Ty\|^2. \quad (4.5)$$

**Proposition 4.5** *Let  $\mathcal{K}$  be a real Hilbert space, let  $\beta \in \mathbb{R}_{++}$ , let  $T: \mathcal{K} \rightarrow \mathcal{K}$  be  $\beta$ -cocoercive, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $L \neq 0$ , and set  $\gamma = \beta/\|L\|^2$ . Then  $L^*TL$  is  $\gamma$ -cocoercive.*

*Proof.* Let  $x$  and  $y$  be in  $\mathcal{H}$ . Then it follows from (4.5) and Fact 2.18(ii) that  $\langle x - y \mid L^*TLx - L^*TLy \rangle = \langle Lx - Ly \mid TLx - TLy \rangle \geq \beta\|TLx - TLy\|^2 \geq \gamma\|L^*TLx - L^*TLy\|^2$ .  $\square$

**Corollary 4.6** *Let  $\mathcal{K}$  be a real Hilbert space, let  $T: \mathcal{K} \rightarrow \mathcal{K}$  be firmly nonexpansive, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\|L\| \leq 1$ . Then  $L^*TL$  is firmly nonexpansive.*

*Proof.* We assume that  $L \neq 0$ , and we set  $\beta = 1$  in Proposition 4.5.  $\square$

**Example 4.7** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be firmly nonexpansive and let  $V$  be a closed linear subspace of  $\mathcal{H}$ . Then it follows from Corollary 3.22(iii)&(vi) and Corollary 4.6 that  $P_VTP_V$  is firmly nonexpansive.

## 4.2 Projectors onto Convex Sets

**Proposition 4.8** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then the projector  $P_C$  is firmly nonexpansive.*

*Proof.* Fix  $x$  and  $y$  in  $\mathcal{H}$ . Theorem 3.14 asserts that  $\langle P_Cy - P_Cx \mid x - P_Cx \rangle \leq 0$  and that  $\langle P_Cx - P_Cy \mid y - P_Cy \rangle \leq 0$ . Adding these two inequalities yields  $\|P_Cx - P_Cy\|^2 \leq \langle x - y \mid P_Cx - P_Cy \rangle$ . The claim therefore follows from Proposition 4.2.  $\square$

**Example 4.9** Suppose that  $\mathcal{H} \neq \{0\}$  and let  $\alpha \in ]0, 1]$ . Let  $T_1$  and  $T_2$  be respectively the *soft* and *hard thresholds* defined for every  $x \in \mathcal{H}$  by

$$T_1x = \begin{cases} (1 - 1/\|x\|)x, & \text{if } \|x\| > 1; \\ 0, & \text{if } \|x\| \leq 1, \end{cases} \quad \text{and} \quad T_2x = \begin{cases} \alpha x, & \text{if } \|x\| > 1; \\ 0, & \text{if } \|x\| \leq 1. \end{cases} \quad (4.6)$$

Then  $T_1$  is firmly nonexpansive. Moreover, for  $\alpha < 1$ ,  $T_2$  is quasinonexpansive but not nonexpansive and, for  $\alpha = 1$ ,  $T_2$  is not quasinonexpansive. Finally, the operator  $T_3$  defined for every  $x \in \mathcal{H}$  by

$$T_3x = \begin{cases} (1 - 2/\|x\|)x, & \text{if } \|x\| > 1; \\ -x, & \text{if } \|x\| \leq 1, \end{cases} \quad (4.7)$$

is nonexpansive but not firmly nonexpansive.

*Proof.* In view of Example 3.16,  $T_1 = \text{Id} - P_{B(0;1)}$ . Hence, it follows from Proposition 4.8 and Proposition 4.2(ii) that  $T_1$  is firmly nonexpansive. Now

suppose that  $\alpha < 1$ . Then 0 is the unique fixed point of  $T_2$  and  $(\forall x \in \mathcal{H}) \|T_2x\| \leq \|x\|$ . Thus,  $T_2$  is quasinonexpansive but not nonexpansive, since it is not continuous. Now suppose that  $\alpha = 1$ , take  $x \in B(0; 1) \setminus \{0\}$ , and set  $y = 2x/\|x\|$ . Then  $y \in \text{Fix } T_2$  but  $\|T_2x - y\| = 2 > 2 - \|x\| = \|x - y\|$ . Thus,  $T_2$  is not quasinonexpansive. Next, we derive from Proposition 4.2(iii) that  $T_3 = 2T_1 - \text{Id}$  is nonexpansive. Finally, take  $x \in \mathcal{H}$  such that  $\|x\| = 1$  and set  $y = -x$ . Then the inequality in (4.1) fails, and therefore  $T$  is not firmly nonexpansive.  $\square$

**Corollary 4.10** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $\text{Id} - P_C$  is firmly nonexpansive and  $2P_C - \text{Id}$  is nonexpansive.*

*Proof.* A consequence of Proposition 4.8 and Proposition 4.2.  $\square$

Proposition 4.8 implies that projectors are continuous. In the affine case, weak continuity also holds.

**Proposition 4.11** *Let  $C$  be a closed affine subspace of  $\mathcal{H}$ . Then the following hold:*

- (i)  $P_C$  is weakly continuous.
- (ii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \|P_Cx - P_Cy\|^2 = \langle x - y \mid P_Cx - P_Cy \rangle$ .

*Proof.* (i): Combine Lemma 2.34 and Corollary 3.20(ii).

(ii): Use Corollary 3.20(i) instead of Theorem 3.14 in the proof of Proposition 4.8.  $\square$

Let us stress that weak continuity of projectors may fail.

**Example 4.12** Suppose that  $\mathcal{H}$  is infinite-dimensional and let  $C = B(0; 1)$ . Then  $P_C$  is not weakly continuous.

*Proof.* Let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence and set  $(\forall n \in \mathbb{N}) x_n = e_1 + e_{2n}$ . Then, as seen in Example 2.25,  $x_n \rightharpoonup e_1$ . However, it follows from Example 3.16 that  $P_Cx_n = x_n/\sqrt{2} \rightharpoonup e_1/\sqrt{2} \neq e_1 = P_Ce_1$ .  $\square$

### 4.3 Fixed Points of Nonexpansive Operators

The projection operator  $P_C$  onto a nonempty closed convex subset  $C$  of  $\mathcal{H}$  is (firmly) nonexpansive by Proposition 4.8 with

$$\text{Fix } P_C = C, \tag{4.8}$$

which is closed and convex. The following results extend this observation.

**Proposition 4.13** *Let  $D$  be a nonempty convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be quasinonexpansive. Then  $\text{Fix } T$  is convex.*



*Proof.* Let  $x$  and  $y$  be in  $\text{Fix } T$ , let  $\alpha \in ]0, 1[$ , and set  $z = \alpha x + (1 - \alpha)y$ . Then  $z \in D$  and, by Corollary 2.14,

$$\begin{aligned}
 \|Tz - z\|^2 &= \|\alpha(Tz - x) + (1 - \alpha)(Tz - y)\|^2 \\
 &= \alpha\|Tz - x\|^2 + (1 - \alpha)\|Tz - y\|^2 - \alpha(1 - \alpha)\|x - y\|^2 \\
 &\leq \alpha\|z - x\|^2 + (1 - \alpha)\|z - y\|^2 - \alpha(1 - \alpha)\|x - y\|^2 \\
 &= \|\alpha(z - x) + (1 - \alpha)(z - y)\|^2 \\
 &= 0.
 \end{aligned} \tag{4.9}$$

Therefore  $z \in \text{Fix } T$ .  $\square$

**Proposition 4.14** *Let  $D$  be a nonempty closed subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be continuous. Then  $\text{Fix } T$  is closed.*

*Proof.* Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{Fix } T$  that converges to a point  $x \in \mathcal{H}$ . Then  $x \in D$  by closedness of  $D$ , while  $Tx_n \rightarrow Tx$  by continuity of  $T$ . On the other hand, since  $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{Fix } T$ ,  $Tx_n \rightarrow x$ . Altogether,  $Tx = x$ .  $\square$

**Corollary 4.15** *Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be nonexpansive. Then  $\text{Fix } T$  is closed and convex.*

**Corollary 4.16** *Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be firmly nonexpansive. Then*

$$\text{Fix } T = \bigcap_{x \in D} \{y \in D \mid \langle y - Tx \mid x - Tx \rangle \leq 0\}. \tag{4.10}$$

*Proof.* Set  $C = \bigcap_{x \in D} \{y \in D \mid \langle y - Tx \mid x - Tx \rangle \leq 0\}$ . For every  $x \in D$  and every  $y \in \text{Fix } T$ , Proposition 4.2(v) yields  $0 \leq \langle Tx - y \mid x - Tx \rangle$ . Hence,  $\text{Fix } T \subset C$ . Conversely, let  $x \in C$ . Then  $x \in \{y \in D \mid \langle y - Tx \mid x - Tx \rangle \leq 0\}$  and therefore  $-\|x - Tx\|^2 = \langle x - Tx \mid x - Tx \rangle \leq 0$ , i.e.,  $x = Tx$ . Thus,  $C \subset \text{Fix } T$ .  $\square$

**Theorem 4.17 (demiclosedness principle)** *Let  $D$  be a nonempty weakly sequentially closed subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$  be a nonexpansive operator, let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $D$ , and let  $x$  and  $u$  be points in  $\mathcal{H}$ . Suppose that  $x_n \rightharpoonup x$  and that  $x_n - Tx_n \rightarrow u$ . Then  $x - Tx = u$ .*

*Proof.* Since  $D \ni x_n \rightharpoonup x$ , we have  $x \in D$ , and  $Tx$  is therefore well defined. For every  $n \in \mathbb{N}$ , it follows from the nonexpansiveness of  $T$  that

$$\begin{aligned}
 \|x - Tx - u\|^2 &= \|x_n - Tx - u\|^2 - \|x_n - x\|^2 - 2\langle x_n - x \mid x - Tx - u \rangle \\
 &= \|x_n - Tx_n - u\|^2 + 2\langle x_n - Tx_n - u \mid Tx_n - Tx \rangle \\
 &\quad + \|Tx_n - Tx\|^2 - \|x_n - x\|^2 - 2\langle x_n - x \mid x - Tx - u \rangle \\
 &\leq \|x_n - Tx_n - u\|^2 + 2\langle x_n - Tx_n - u \mid Tx_n - Tx \rangle
 \end{aligned}$$

$$-2 \langle x_n - x \mid x - Tx - u \rangle. \quad (4.11)$$

However, by assumption,  $x_n - Tx_n - u \rightarrow 0$ ,  $x_n - x \rightarrow 0$ , and hence  $Tx_n - Tx \rightarrow x - Tx - u$ . Taking the limit as  $n \rightarrow +\infty$  in (4.11) and appealing to Lemma 2.41(iii), we obtain  $x - Tx = u$ .  $\square$

**Corollary 4.18** *Let  $D$  be a nonempty closed and convex subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$  be nonexpansive, let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $D$ , and let  $x$  be a point in  $\mathcal{H}$ . Suppose that  $x_n \rightarrow x$  and that  $x_n - Tx_n \rightarrow 0$ . Then  $x \in \text{Fix } T$ .*

*Proof.* Since Theorem 3.32 asserts that  $D$  is weakly sequentially closed, the result follows from Theorem 4.17.  $\square$

The set of fixed points of a nonexpansive operator may be empty (consider a translation by a nonzero vector). The following theorem gives a condition that guarantees the existence of fixed points.

**Theorem 4.19 (Browder–Göhde–Kirk)** *Let  $D$  be a nonempty bounded closed convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow D$  be a nonexpansive operator. Then  $\text{Fix } T \neq \emptyset$ .*

*Proof.* It follows from Theorem 3.32 that  $D$  is weakly sequentially closed, and from Theorem 3.33 that it is weakly sequentially compact. Now fix  $x_0 \in D$  and a sequence  $(\alpha_n)_{n \in \mathbb{N}}$  in  $]0, 1]$  such that  $\alpha_0 = 1$  and  $\alpha_n \downarrow 0$ . For every  $n \in \mathbb{N}$ , the operator  $T_n: D \rightarrow D: x \mapsto \alpha_n x_0 + (1 - \alpha_n)Tx$  is a strict contraction with constant  $1 - \alpha_n$ , and it therefore possesses a fixed point  $x_n \in D$  by Theorem 1.48. Moreover, for every  $n \in \mathbb{N}$ ,  $\|x_n - Tx_n\| = \|T_n x_n - Tx_n\| = \alpha_n \|x_0 - Tx_n\| \leq \alpha_n \text{diam}(D)$ . Hence  $x_n - Tx_n \rightarrow 0$ . On the other hand, since  $(x_n)_{n \in \mathbb{N}}$  lies in  $D$ , by weak sequential compactness we can extract a weakly convergent subsequence, say  $x_{k_n} \rightarrow x \in D$ . Since  $x_{k_n} - Tx_{k_n} \rightarrow 0$ , Corollary 4.18 asserts that  $x \in \text{Fix } T$ .  $\square$

The proof of Theorem 4.19 rests on Lipschitz continuous operators and their unique fixed points. These fixed points determine a curve, which we investigate in more detail in the next result.

**Proposition 4.20 (approximating curve)** *Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow D$  be a nonexpansive operator. Then, for every  $\varepsilon \in ]0, 1[$  and every  $x \in D$ , there exists a unique point  $x_\varepsilon \in D$  such that*

$$x_\varepsilon = \varepsilon x + (1 - \varepsilon)Tx_\varepsilon. \quad (4.12)$$

*Set, for every  $\varepsilon \in ]0, 1[$ ,  $T_\varepsilon: D \rightarrow D: x \mapsto x_\varepsilon$ , and let  $x \in D$ . Then the following hold:*

- (i)  $(\forall \varepsilon \in ]0, 1[) \ T_\varepsilon = \varepsilon \text{Id} + (1 - \varepsilon)TT_\varepsilon = (\text{Id} - (1 - \varepsilon)T)^{-1} \circ \varepsilon \text{Id}$ .
- (ii)  $(\forall \varepsilon \in ]0, 1[) \ T_\varepsilon$  is firmly nonexpansive.
- (iii)  $(\forall \varepsilon \in ]0, 1[) \ \text{Fix } T_\varepsilon = \text{Fix } T$ .
- (iv)  $(\forall \varepsilon \in ]0, 1[) \ \varepsilon(x - Tx_\varepsilon) = x_\varepsilon - Tx_\varepsilon = (1 - \varepsilon)^{-1}\varepsilon(x - x_\varepsilon)$ .

- (v) Suppose that  $\text{Fix } T = \emptyset$ . Then  $\lim_{\varepsilon \downarrow 0} \|x_\varepsilon\| = +\infty$ .
- (vi)  $(\forall \varepsilon \in ]0, 1[)(\forall y \in \text{Fix } T) \|x - x_\varepsilon\|^2 + \|x_\varepsilon - y\|^2 \leq \|x - y\|^2$ .
- (vii) Suppose that  $\text{Fix } T \neq \emptyset$ . Then  $\lim_{\varepsilon \downarrow 0} x_\varepsilon = P_{\text{Fix } T} x$ .
- (viii)  $(\forall \varepsilon \in ]0, 1[)(\forall \delta \in ]0, 1[)$

$$\left(\frac{\varepsilon - \delta}{1 - \varepsilon}\right)^2 \|x_\varepsilon - x\|^2 + \delta(2 - \delta)\|x_\delta - x_\varepsilon\|^2 \leq 2\frac{\varepsilon - \delta}{1 - \varepsilon} \langle x_\varepsilon - x \mid x_\delta - x_\varepsilon \rangle.$$

- (ix)  $(\forall \varepsilon \in ]0, 1[)(\forall \delta \in ]0, \varepsilon[) \|x - x_\varepsilon\|^2 + \|x_\varepsilon - x_\delta\|^2 \leq \|x - x_\delta\|^2$ .
- (x) The function  $]0, 1[ \rightarrow \mathbb{R}_+ : \varepsilon \mapsto \|x - x_\varepsilon\|$  is decreasing.
- (xi) The curve  $]0, 1[ \rightarrow \mathcal{H} : \varepsilon \mapsto x_\varepsilon$  is continuous.
- (xii) If  $x \in \text{Fix } T$ , then  $x_\varepsilon \equiv x$  is constant; otherwise,  $(x_\varepsilon)_{\varepsilon \in ]0, 1[}$  is an injective curve.

*Proof.* Let  $\varepsilon \in ]0, 1[$ . By Theorem 1.48, the operator  $D \rightarrow D : z \mapsto \varepsilon z + (1 - \varepsilon)Tz$  has a unique fixed point. Hence,  $x_\varepsilon$  is unique, and  $T_\varepsilon$  is therefore well defined.

(i): The first identity is clear from (4.12). Furthermore,  $\varepsilon \text{Id} = T_\varepsilon - (1 - \varepsilon)TT_\varepsilon = (\text{Id} - (1 - \varepsilon)T)T_\varepsilon$ , and therefore, since  $\text{Id} - (1 - \varepsilon)T$  is injective, we obtain  $T_\varepsilon = (\text{Id} - (1 - \varepsilon)T)^{-1} \circ \varepsilon \text{Id}$ .

(ii): Let  $y \in D$ . Then

$$\begin{aligned} x - y &= \varepsilon^{-1}((x_\varepsilon - (1 - \varepsilon)Tx_\varepsilon) - (y_\varepsilon - (1 - \varepsilon)Ty_\varepsilon)) \\ &= \varepsilon^{-1}((x_\varepsilon - y_\varepsilon) - (1 - \varepsilon)(Tx_\varepsilon - Ty_\varepsilon)). \end{aligned} \quad (4.13)$$

Using (4.12), (4.13), and Cauchy–Schwarz, we deduce that

$$\begin{aligned} \langle T_\varepsilon x - T_\varepsilon y \mid (\text{Id} - T_\varepsilon)x - (\text{Id} - T_\varepsilon)y \rangle &= \langle x_\varepsilon - y_\varepsilon \mid (1 - \varepsilon)(x - Tx_\varepsilon) - (1 - \varepsilon)(y - Ty_\varepsilon) \rangle \\ &= (1 - \varepsilon) \langle x_\varepsilon - y_\varepsilon \mid (x - y) - (Tx_\varepsilon - Ty_\varepsilon) \rangle \\ &= (1 - \varepsilon)\varepsilon^{-1} \langle x_\varepsilon - y_\varepsilon \mid (x_\varepsilon - y_\varepsilon) - (Tx_\varepsilon - Ty_\varepsilon) \rangle \\ &\geq (\varepsilon^{-1} - 1)(\|x_\varepsilon - y_\varepsilon\|^2 - \|x_\varepsilon - y_\varepsilon\| \|Tx_\varepsilon - Ty_\varepsilon\|) \\ &= (\varepsilon^{-1} - 1)\|x_\varepsilon - y_\varepsilon\|(\|x_\varepsilon - y_\varepsilon\| - \|Tx_\varepsilon - Ty_\varepsilon\|) \\ &\geq 0. \end{aligned} \quad (4.14)$$

Hence, by Proposition 4.2,  $T_\varepsilon$  is firmly nonexpansive.

(iii): Let  $x \in D$ . Suppose first that  $x \in \text{Fix } T$ . Then  $x = \varepsilon x + (1 - \varepsilon)Tx$  and hence  $x = x_\varepsilon$  by uniqueness of  $x_\varepsilon$ . It follows that  $T_\varepsilon x = x$  and therefore that  $x \in \text{Fix } T_\varepsilon$ . Conversely, assume that  $x \in \text{Fix } T_\varepsilon$ . Then  $x = x_\varepsilon = \varepsilon x + (1 - \varepsilon)Tx_\varepsilon = x + (1 - \varepsilon)(Tx - x)$  and thus  $x = Tx$ , i.e.,  $x \in \text{Fix } T$ .

(iv): This follows from (4.12).

(v): Suppose that there exists a sequence  $(\varepsilon_n)_{n \in \mathbb{N}}$  in  $]0, 1[$  such that  $\varepsilon_n \downarrow 0$  and  $(x_{\varepsilon_n})_{n \in \mathbb{N}}$  is bounded. By (iv) and Corollary 4.18, every weak sequential cluster point of  $(x_{\varepsilon_n})_{n \in \mathbb{N}}$  lies in  $\text{Fix } T$ .

(vi): Assume that  $y \in \text{Fix } T$ . By (iii),  $y = T_\varepsilon y = y_\varepsilon$ . Since  $T_\varepsilon$  is firmly nonexpansive by (ii), we have

$$\|x - y\|^2 \geq \|x_\varepsilon - y_\varepsilon\|^2 + \|(x - x_\varepsilon) - (y - y_\varepsilon)\|^2 = \|x_\varepsilon - y\|^2 + \|x - x_\varepsilon\|^2. \quad (4.15)$$

(vii): Let  $(\varepsilon_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1[$  such that  $\varepsilon_n \downarrow 0$ , and set  $(\forall n \in \mathbb{N})$   $z_n = x_{\varepsilon_n}$ . By (vi),  $(z_n)_{n \in \mathbb{N}}$  is bounded. Thus, using (iv), we see that  $z_n - Tz_n \rightarrow 0$ . Let  $z$  be a weak sequential cluster point of  $(z_n)_{n \in \mathbb{N}}$ , say  $\underline{z_{k_n}} \rightharpoonup z$ . Theorem 4.17 implies that  $z \in \text{Fix } T$ . In view of (vi), we obtain  $\lim \|x - z_{k_n}\|^2 \leq \|x - z\|^2$ . Since  $x - z_{k_n} \rightharpoonup x - z$ , Lemma 2.41(i) yields  $x - z_{k_n} \rightarrow x - z$ . Thus  $z_{k_n} \rightarrow z$ . Again, using (vi), we see that  $(\forall n \in \mathbb{N})$   $\|x - z_{k_n}\|^2 + \|z_{k_n} - y\|^2 \leq \|x - y\|^2$ . Taking the limit as  $n \rightarrow +\infty$ , we deduce that  $\|x - z\|^2 + \|z - y\|^2 \leq \|x - y\|^2$ . Hence  $(\forall y \in \text{Fix } T)$   $\langle y - z \mid x - z \rangle \leq 0$ . It now follows from Theorem 3.14 that  $z = P_{\text{Fix } T} x$ . Therefore,  $z_n \rightarrow P_{\text{Fix } T} x$  and hence  $x_\varepsilon \rightarrow P_{\text{Fix } T} x$  as  $\varepsilon \downarrow 0$ .

(viii): Let  $\delta \in ]0, 1[$  and set  $y_\varepsilon = x_\varepsilon - x$  and  $y_\delta = x_\delta - x$ . Since  $y_\delta = y_\varepsilon + x_\delta - x_\varepsilon$ , using (4.12), we obtain

$$\begin{aligned} \|x_\delta - x_\varepsilon\|^2 &\geq \|Tx_\delta - Tx_\varepsilon\|^2 \\ &= \left\| \frac{x_\delta - \delta x}{1 - \delta} - \frac{x_\varepsilon - \varepsilon x}{1 - \varepsilon} \right\|^2 \\ &= \left\| \frac{y_\delta}{1 - \delta} - \frac{y_\varepsilon}{1 - \varepsilon} \right\|^2 \\ &= \frac{1}{(1 - \delta)^2} \left\| \frac{\delta - \varepsilon}{1 - \varepsilon} y_\varepsilon + x_\delta - x_\varepsilon \right\|^2 \\ &= \frac{1}{(1 - \delta)^2} \left( \left( \frac{\delta - \varepsilon}{1 - \varepsilon} \right)^2 \|y_\varepsilon\|^2 + 2 \frac{\delta - \varepsilon}{1 - \varepsilon} \langle y_\varepsilon \mid x_\delta - x_\varepsilon \rangle \right. \\ &\quad \left. + \|x_\delta - x_\varepsilon\|^2 \right). \end{aligned} \quad (4.16)$$

Therefore,

$$\left( \frac{\varepsilon - \delta}{1 - \varepsilon} \right)^2 \|y_\varepsilon\|^2 + \delta(2 - \delta) \|x_\delta - x_\varepsilon\|^2 \leq 2 \frac{\varepsilon - \delta}{1 - \varepsilon} \langle y_\varepsilon \mid x_\delta - x_\varepsilon \rangle, \quad (4.17)$$

which is the desired inequality.

(viii) $\Rightarrow$ (ix): Let  $\delta \in ]0, \varepsilon[$ . Then  $\langle x_\varepsilon - x \mid x_\delta - x_\varepsilon \rangle \geq 0$ . In turn,  $\|x_\delta - x\|^2 = \|x_\delta - x_\varepsilon\|^2 + 2 \langle x_\delta - x_\varepsilon \mid x_\varepsilon - x \rangle + \|x_\varepsilon - x\|^2 \geq \|x_\delta - x_\varepsilon\|^2 + \|x_\varepsilon - x\|^2$ .

(ix) $\Rightarrow$ (x): Clear.

(xi): We derive from (viii) and Cauchy-Schwarz that

$$(\forall \delta \in ]0, \varepsilon[) \quad \|x_\delta - x_\varepsilon\| \leq \frac{2(\varepsilon - \delta)}{\delta(2 - \delta)(1 - \varepsilon)} \|x_\varepsilon - x\|. \quad (4.18)$$

Hence,  $\|x_\delta - x_\varepsilon\| \downarrow 0$  as  $\delta \uparrow \varepsilon$ , and therefore the curve  $]0, 1[ \rightarrow \mathcal{H}: \varepsilon \mapsto x_\varepsilon$  is left-continuous. Likewise, we have

$$(\forall \delta \in ]\varepsilon, 1[) \quad \|x_\delta - x_\varepsilon\| \leq \frac{2(\delta - \varepsilon)}{\delta(2 - \delta)(1 - \varepsilon)} \|x_\varepsilon - x\|, \quad (4.19)$$

so that  $\|x_\delta - x_\varepsilon\| \downarrow 0$  as  $\delta \downarrow \varepsilon$ . Thus, the curve  $]0, 1[ \rightarrow \mathcal{H}: \varepsilon \mapsto x_\varepsilon$  is right-continuous.

(xii): If  $x \in \text{Fix } T$ , then  $x \in \text{Fix } T_\varepsilon$  by (iii) and hence  $x = T_\varepsilon x = x_\varepsilon$ . Now assume that  $x \notin \text{Fix } T$ . If  $\delta \in ]0, \varepsilon[$  and  $x_\varepsilon = x_\delta$ , then (viii) yields  $T_\varepsilon x = x_\varepsilon = x$  and hence  $x \in \text{Fix } T_\varepsilon$ , which is impossible in view of (iii). Hence the curve  $(x_\varepsilon)_{\varepsilon \in ]0, 1[}$  is injective in that case.  $\square$

**Proposition 4.21** *Let  $T_1: \mathcal{H} \rightarrow \mathcal{H}$  and  $T_2: \mathcal{H} \rightarrow \mathcal{H}$  be firmly nonexpansive and set  $T = T_1(2T_2 - \text{Id}) + \text{Id} - T_2$ . Then the following hold:*

- (i)  $2T - \text{Id} = (2T_1 - \text{Id})(2T_2 - \text{Id})$ .
- (ii)  $T$  is firmly nonexpansive.
- (iii)  $\text{Fix } T = \text{Fix}(2T_1 - \text{Id})(2T_2 - \text{Id})$ .
- (iv) Suppose that  $T_1$  is the projector onto a closed affine subspace. Then  $\text{Fix } T = \{x \in \mathcal{H} \mid T_1 x = T_2 x\}$ .

*Proof.* (i): Expand.

(ii): Proposition 4.2 asserts that  $2T_1 - \text{Id}$  and  $2T_2 - \text{Id}$  are nonexpansive. Therefore,  $(2T_1 - \text{Id})(2T_2 - \text{Id})$  is nonexpansive and so is  $2T - \text{Id}$  by (i). In turn,  $T$  is firmly nonexpansive.

(iii): By (i),  $\text{Fix } T = \text{Fix}(2T - \text{Id}) = \text{Fix}(2T_1 - \text{Id})(2T_2 - \text{Id})$ .

(iv): Suppose that  $T_1 = P_C$ , where  $C$  is a closed affine subspace of  $\mathcal{H}$ , and let  $x \in \mathcal{H}$ . It follows from Proposition 4.8 that  $T_1$  is firmly nonexpansive and from Corollary 3.20(ii) that  $x \in \text{Fix } T \Leftrightarrow x = P_C(2T_2 x + (1 - 2)x) + x - T_2 x \Leftrightarrow T_2 x = 2P_C(T_2 x) + (1 - 2)P_C x \in C \Leftrightarrow P_C(T_2 x) = T_2 x = 2P_C(T_2 x) + (1 - 2)P_C x \Leftrightarrow T_2 x = P_C x$ .  $\square$

**Corollary 4.22** *Let  $T_1$  be the projector onto a linear subspace of  $\mathcal{H}$ , let  $T_2: \mathcal{H} \rightarrow \mathcal{H}$  be firmly nonexpansive, and set  $T = T_1 T_2 + (\text{Id} - T_1)(\text{Id} - T_2)$ . Then  $T$  is firmly nonexpansive and  $\text{Fix } T = \{x \in \mathcal{H} \mid T_1 x = T_2 x\}$ .*

*Proof.* Since  $T = T_1(2T_2 - \text{Id}) + \text{Id} - T_2$ , the result follows from Proposition 4.21.  $\square$

## 4.4 Averaged Nonexpansive Operators

**Definition 4.23** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$  be nonexpansive, and let  $\alpha \in ]0, 1[$ . Then  $T$  is *averaged* with constant  $\alpha$ , or  *$\alpha$ -averaged*, if there exists a nonexpansive operator  $R: D \rightarrow \mathcal{H}$  such that  $T = (1 - \alpha)\text{Id} + \alpha R$ .

**Remark 4.24** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$ .

- (i) If  $T$  is averaged, then it is nonexpansive.
- (ii) If  $T$  is nonexpansive, it is not necessarily averaged: consider  $T = -\text{Id}: \mathcal{H} \rightarrow \mathcal{H}$  when  $\mathcal{H} \neq \{0\}$ .
- (iii) It follows from Proposition 4.2 that  $T$  is firmly nonexpansive if and only if it is  $1/2$ -averaged.
- (iv) Let  $\beta \in \mathbb{R}_{++}$ . Then it follows from (iii) that  $T$  is  $\beta$ -cocoercive if and only if  $\beta T$  is  $1/2$ -averaged.

**Proposition 4.25** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$  be nonexpansive, and let  $\alpha \in ]0, 1[$ . Then the following are equivalent:

- (i)  $T$  is  $\alpha$ -averaged.
- (ii)  $(1 - 1/\alpha)\text{Id} + (1/\alpha)T$  is nonexpansive.
- (iii)  $(\forall x \in D)(\forall y \in D) \|Tx - Ty\|^2 \leq \|x - y\|^2 - \frac{1 - \alpha}{\alpha} \|(\text{Id} - T)x - (\text{Id} - T)y\|^2$ .
- (iv)  $(\forall x \in D)(\forall y \in D) \|Tx - Ty\|^2 + (1 - \frac{1 - \alpha}{2\alpha})\|x - y\|^2 \leq 2(1 - \alpha) \langle x - y | Tx - Ty \rangle$ .

*Proof.* Fix  $x$  and  $y$  in  $D$ , set  $R = (1 - \lambda)\text{Id} + \lambda T$ , where  $\lambda = 1/\alpha$ , and note that  $T = (1 - \alpha)\text{Id} + \alpha R$ .

(i)  $\Leftrightarrow$  (ii)  $\Leftrightarrow$  (iii): Corollary 2.14 yields

$$\begin{aligned} \|Rx - Ry\|^2 &= (1 - \lambda)\|x - y\|^2 + \lambda\|Tx - Ty\|^2 \\ &\quad - \lambda(1 - \lambda)\|(\text{Id} - T)x - (\text{Id} - T)y\|^2. \end{aligned} \quad (4.20)$$

In other words,

$$\begin{aligned} \alpha(\|x - y\|^2 - \|Rx - Ry\|^2) &= \|x - y\|^2 - \|Tx - Ty\|^2 \\ &\quad - \frac{1 - \alpha}{\alpha} \|(\text{Id} - T)x - (\text{Id} - T)y\|^2. \end{aligned} \quad (4.21)$$

Now observe that (i)  $\Leftrightarrow$  (ii)  $\Leftrightarrow R$  is nonexpansive  $\Leftrightarrow$  the left-hand side of (4.21) is positive  $\Leftrightarrow$  (iii).

(iii)  $\Leftrightarrow$  (iv): Write  $\|(\text{Id} - T)x - (\text{Id} - T)y\|^2 = \|x - y\|^2 + \|Tx - Ty\|^2 - 2 \langle x - y | Tx - Ty \rangle$  in (iii).  $\square$

**Remark 4.26** It follows from the implication (i)  $\Rightarrow$  (iii) in Proposition 4.25 that averaged operators are strictly quasinonexpansive.

**Remark 4.27** It follows from Proposition 4.25(iii) that if  $T: D \rightarrow \mathcal{H}$  is  $\alpha$ -averaged with  $\alpha \in ]0, 1/2]$ , then  $T$  is firmly nonexpansive.

We now describe some operations that preserve averagedness.

**Proposition 4.28** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , let  $\alpha \in ]0, 1[$ , and let  $\lambda \in ]0, 1/\alpha[$ . Then  $T$  is  $\alpha$ -averaged if and only if  $(1 - \lambda)\text{Id} + \lambda T$  is  $\lambda\alpha$ -averaged.

*Proof.* Set  $R = (1 - \alpha^{-1})\text{Id} + \alpha^{-1}T$ . Then the conclusion follows from the identities  $T = (1 - \alpha)\text{Id} + \alpha R$  and  $(1 - \lambda)\text{Id} + \lambda T = (1 - \lambda\alpha)\text{Id} + \lambda\alpha R$ .  $\square$

**Corollary 4.29** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , and let  $\lambda \in ]0, 2[$ . Then  $T$  is firmly nonexpansive if and only if  $(1 - \lambda)\text{Id} + \lambda T$  is  $\lambda/2$ -averaged.*

*Proof.* Set  $\alpha = 1/2$  in Proposition 4.28 and use Remark 4.24(iii).  $\square$

**Proposition 4.30** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $(T_i)_{i \in I}$  be a finite family of nonexpansive operators from  $D$  to  $\mathcal{H}$ , let  $(\omega_i)_{i \in I}$  be real numbers in  $]0, 1]$  such that  $\sum_{i \in I} \omega_i = 1$ , and let  $(\alpha_i)_{i \in I}$  be real numbers in  $]0, 1]$  such that, for every  $i \in I$ ,  $T_i$  is  $\alpha_i$ -averaged, and set  $\alpha = \max_{i \in I} \alpha_i$ . Then  $\sum_{i \in I} \omega_i T_i$  is  $\alpha$ -averaged.*

*Proof.* Set  $T = \sum_{i \in I} \omega_i T_i$ , and fix  $x$  and  $y$  in  $D$ . The implication (i) $\Rightarrow$ (iii) in Proposition 4.25 yields

$$(\forall i \in I) \quad \|T_i x - T_i y\|^2 + \frac{1 - \alpha_i}{\alpha_i} \|(\text{Id} - T_i)x - (\text{Id} - T_i)y\|^2 \leq \|x - y\|^2. \quad (4.22)$$

Hence, by convexity of  $\|\cdot\|^2$ , since  $(1 - \alpha)/\alpha = \min_{i \in I} (1 - \alpha_i)/\alpha_i$ ,

$$\begin{aligned} & \|Tx - Ty\|^2 + \frac{1 - \alpha}{\alpha} \|(\text{Id} - T)x - (\text{Id} - T)y\|^2 \\ &= \left\| \sum_{i \in I} \omega_i T_i x - \sum_{i \in I} \omega_i T_i y \right\|^2 \\ & \quad + \frac{1 - \alpha}{\alpha} \left\| \sum_{i \in I} \omega_i (\text{Id} - T_i)x - \sum_{i \in I} \omega_i (\text{Id} - T_i)y \right\|^2 \\ &\leq \sum_{i \in I} \omega_i \|T_i x - T_i y\|^2 + \sum_{i \in I} \frac{1 - \alpha_i}{\alpha_i} \omega_i \|(\text{Id} - T_i)x - (\text{Id} - T_i)y\|^2 \\ &\leq \|x - y\|^2. \end{aligned} \quad (4.23)$$

Using the implication (iii) $\Rightarrow$ (i) in Proposition 4.25 completes the proof.  $\square$

**Example 4.31** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $(T_i)_{i \in I}$  be a finite family of firmly nonexpansive operators from  $D$  to  $\mathcal{H}$ , and let  $(\omega_i)_{i \in I}$  be real numbers in  $]0, 1]$  such that  $\sum_{i \in I} \omega_i = 1$ . Then  $\sum_{i \in I} \omega_i T_i$  is firmly nonexpansive.

*Proof.* In view of Remark 4.24(iii), setting  $\alpha_i \equiv 1/2$  in Proposition 4.30 yields the result.  $\square$

Next, we consider compositions of averaged operators.

**Proposition 4.32** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(T_i)_{i \in I}$  be a family of operators from  $D$  to  $D$ , let  $(\alpha_i)_{i \in I}$  be real numbers in  $]0, 1[$  such that, for every  $i \in I$ ,  $T_i$  is  $\alpha_i$ -averaged, and set*

$$T = T_1 \cdots T_m \quad \text{and} \quad \alpha = \frac{m}{m - 1 + \frac{1}{\max_{i \in I} \alpha_i}}. \quad (4.24)$$

*Then  $T$  is  $\alpha$ -averaged.*

*Proof.* Set  $(\forall i \in I) \kappa_i = \alpha_i/(1 - \alpha_i)$ , set  $\kappa = \max_{i \in I} \kappa_i$ , and let  $x$  and  $y$  be in  $D$ . We derive from the convexity of  $\|\cdot\|^2$  and the equivalence (i) $\Leftrightarrow$ (iii) in Proposition 4.25 that

$$\begin{aligned} & \|(\text{Id} - T)x - (\text{Id} - T)y\|^2/m \\ &= \|(x - y) - (T_mx - T_my) + (T_mx - T_my) \\ &\quad - (T_{m-1}T_mx - T_{m-1}T_my) + (T_{m-1}T_mx - T_{m-1}T_my) - \cdots \\ &\quad - (T_2 \cdots T_mx - T_2 \cdots T_my) + (T_2 \cdots T_mx - T_2 \cdots T_my) \\ &\quad - (T_1 \cdots T_mx - T_1 \cdots T_my)\|^2/m \\ &= \|(\text{Id} - T_m)x - (\text{Id} - T_m)y \\ &\quad + (\text{Id} - T_{m-1})T_mx - (\text{Id} - T_{m-1})T_my + \cdots \\ &\quad + (\text{Id} - T_1)T_2 \cdots T_mx - (\text{Id} - T_1)T_2 \cdots T_my\|^2/m \\ &\leq \|(\text{Id} - T_m)x - (\text{Id} - T_m)y\|^2 \\ &\quad + \|(\text{Id} - T_{m-1})T_mx - (\text{Id} - T_{m-1})T_my\|^2 + \cdots \\ &\quad + \|(\text{Id} - T_1)T_2 \cdots T_mx - (\text{Id} - T_1)T_2 \cdots T_my\|^2 \\ &\leq \kappa_m (\|x - y\|^2 - \|T_mx - T_my\|^2) \\ &\quad + \kappa_{m-1} (\|T_mx - T_my\|^2 - \|T_{m-1}T_mx - T_{m-1}T_my\|^2) + \cdots \\ &\quad + \kappa_1 (\|T_2 \cdots T_mx - T_2 \cdots T_my\|^2 - \|T_1 \cdots T_mx - T_1 \cdots T_my\|^2) \\ &\leq \kappa (\|x - y\|^2 - \|Tx - Ty\|^2). \end{aligned} \quad (4.25)$$

Consequently, it follows from Proposition 4.25 and (4.24) that  $T$  is averaged, with constant  $m/(m + 1/\kappa) = \alpha$ .  $\square$

**Proposition 4.33** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $\beta \in \mathbb{R}_{++}$ , let  $T: D \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive, and let  $\gamma \in ]0, 2\beta[$ . Then  $\text{Id} - \gamma T$  is  $\gamma/(2\beta)$ -averaged.*

*Proof.* By Remark 4.24(iv),  $\beta T$  is  $1/2$ -averaged. Hence, there exists a non-expansive operator  $R: D \rightarrow \mathcal{H}$  such that  $T = (\text{Id} + R)/(2\beta)$ . In turn,  $\text{Id} - \gamma T = (1 - \gamma/(2\beta))\text{Id} + (\gamma/(2\beta))(-R)$ .  $\square$



## 4.5 Common Fixed Points

The first result concerns the fixed point set of convex combinations of quasicononexpansive operators and the second that of compositions of strictly quasicononexpansive operators.

**Proposition 4.34** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $(T_i)_{i \in I}$  be a finite family of quasicononexpansive operators from  $D$  to  $\mathcal{H}$  such that  $\bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , let  $(\omega_i)_{i \in I}$  be strictly positive real numbers such that  $\sum_{i \in I} \omega_i = 1$ . Then  $\text{Fix } \sum_{i \in I} \omega_i T_i = \bigcap_{i \in I} \text{Fix } T_i$ .*

*Proof.* Set  $T = \sum_{i \in I} \omega_i T_i$ . It is clear that  $\bigcap_{i \in I} \text{Fix } T_i \subset \text{Fix } T$ . To prove the reverse inclusion, let  $y \in \bigcap_{i \in I} \text{Fix } T_i$ . Then (4.3) yields

$$\begin{aligned} (\forall i \in I)(\forall x \in D) \quad 2 \langle T_i x - x \mid x - y \rangle &= \|T_i x - y\|^2 - \|T_i x - x\|^2 - \|x - y\|^2 \\ &\leq -\|T_i x - x\|^2. \end{aligned} \quad (4.26)$$

Now let  $x \in \text{Fix } T$ . Then we derive from (4.26) that

$$0 = 2 \langle T x - x \mid x - y \rangle = 2 \sum_{i \in I} \omega_i \langle T_i x - x \mid x - y \rangle \leq - \sum_{i \in I} \omega_i \|T_i x - x\|^2. \quad (4.27)$$

Therefore  $\sum_{i \in I} \omega_i \|T_i x - x\|^2 = 0$ , and we conclude that  $x \in \bigcap_{i \in I} \text{Fix } T_i$ .  $\square$

**Proposition 4.35** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , and let  $T_1$  and  $T_2$  be quasicononexpansive operators from  $D$  to  $D$ . Suppose that  $T_1$  or  $T_2$  is strictly quasicononexpansive, and that  $\text{Fix } T_1 \cap \text{Fix } T_2 \neq \emptyset$ . Then the following hold:*

- (i)  $\text{Fix } T_1 T_2 = \text{Fix } T_1 \cap \text{Fix } T_2$ .
- (ii)  $T_1 T_2$  is quasicononexpansive.
- (iii) Suppose that  $T_1$  and  $T_2$  are strictly quasicononexpansive. Then  $T_1 T_2$  is strictly quasicononexpansive.

*Proof.* (i): It is clear that  $\text{Fix } T_1 \cap \text{Fix } T_2 \subset \text{Fix } T_1 T_2$ . Now let  $x \in \text{Fix } T_1 T_2$  and let  $y \in \text{Fix } T_1 \cap \text{Fix } T_2$ . We consider three cases.

- (a)  $T_2 x \in \text{Fix } T_1$ . Then  $T_2 x = T_1 T_2 x = x \in \text{Fix } T_1 \cap \text{Fix } T_2$ .
- (b)  $x \in \text{Fix } T_2$ . Then  $T_1 x = T_1 T_2 x = x \in \text{Fix } T_1 \cap \text{Fix } T_2$ .
- (c)  $T_2 x \notin \text{Fix } T_1$  and  $x \notin \text{Fix } T_2$ . Since  $T_1$  or  $T_2$  is strictly quasicononexpansive, at least one of the inequalities in  $\|x - y\| = \|T_1 T_2 x - y\| \leq \|T_2 x - y\| \leq \|x - y\|$  is strict, which is impossible.

(ii): Let  $x \in D$  and let  $y \in \text{Fix } T_1 T_2 = \text{Fix } T_1 \cap \text{Fix } T_2$ . Then  $\|T_1 T_2 x - y\| \leq \|T_2 x - y\| \leq \|x - y\|$ , and therefore  $T_1 T_2$  is quasicononexpansive.

(iii): Let  $x \in D \setminus \text{Fix } T_1 T_2$  and let  $y \in \text{Fix } T_1 T_2 = \text{Fix } T_1 \cap \text{Fix } T_2$ . If  $x \notin \text{Fix } T_2$ , then  $\|T_1 T_2 x - y\| \leq \|T_2 x - y\| < \|x - y\|$ . Finally, if  $x \in \text{Fix } T_2$ , then  $x \notin \text{Fix } T_1$  (for otherwise  $T_1 x = x \in \text{Fix } T_1 \cap \text{Fix } T_2$ , which is impossible) and hence  $\|T_1 T_2 x - y\| < \|x - y\|$ .  $\square$

**Corollary 4.36** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(T_i)_{i \in I}$  be a family of strictly quasinonexpansive operators from  $D$  to  $D$  such that  $\bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , and set  $T = T_1 \cdots T_m$ . Then  $T$  is strictly quasinonexpansive and  $\text{Fix } T = \bigcap_{i \in I} \text{Fix } T_i$ .*

*Proof.* We proceed by strong induction on  $m$ . The result is clear for  $m = 1$  and for  $m = 2$  by Proposition 4.35. Now suppose that  $m \geq 2$  and that the result holds for up to  $m$  operators. Let  $(T_i)_{1 \leq i \leq m+1}$  be a family of strictly quasinonexpansive operators from  $D$  to  $D$  such that  $\bigcap_{i=1}^{m+1} \text{Fix } T_i \neq \emptyset$ . Set  $R_1 = T_1 \cdots T_m$  and  $R_2 = T_{m+1}$ . Then  $R_2$  is strictly quasinonexpansive with  $\text{Fix } R_2 = \text{Fix } T_{m+1}$ , and, by the induction hypothesis,  $R_1$  is strictly quasinonexpansive with  $\text{Fix } R_1 = \bigcap_{i=1}^m \text{Fix } T_i$ . Therefore, by Proposition 4.35(iii)&(i),  $R_1 R_2 = T_1 \cdots T_{m+1}$  is strictly quasinonexpansive and  $\text{Fix } T_1 \cdots T_{m+1} = \text{Fix } R_1 R_2 = \text{Fix } R_1 \cap \text{Fix } R_2 = \bigcap_{i=1}^{m+1} \text{Fix } T_i$ .  $\square$

**Corollary 4.37** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(T_i)_{i \in I}$  be a family of averaged nonexpansive operators from  $D$  to  $D$  such that  $\bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , and set  $T = T_1 \cdots T_m$ . Then  $\text{Fix } T = \bigcap_{i \in I} \text{Fix } T_i$ .*

*Proof.* In view of Remark 4.26, this follows from Corollary 4.36.  $\square$

## Exercises

**Exercise 4.1** Let  $U$  be a nonempty open interval in  $\mathbb{R}$ , let  $D$  be a closed interval contained in  $U$ , and suppose that  $\tilde{T}: U \rightarrow \mathbb{R}$  is differentiable on  $U$ . Set  $T = \tilde{T}|_D$ . Show the following:

- (i)  $T$  is firmly nonexpansive  $\Leftrightarrow \text{ran } T' \subset [0, 1]$ .
- (ii)  $T$  is nonexpansive  $\Leftrightarrow \text{ran } T' \subset [-1, 1]$ .

**Exercise 4.2** Let  $D$  be a nonempty subset of  $\mathbb{R}$ , and let  $T: D \rightarrow \mathbb{R}$ . Show that  $T$  is firmly nonexpansive if and only if  $T$  is nonexpansive and increasing. Provided that  $\text{ran } T \subset D$ , deduce that if  $T$  is firmly nonexpansive, then so is  $T \circ T$ .

**Exercise 4.3** Suppose that  $\mathcal{H} \neq \{0\}$ . Without using Example 4.9, show that every firmly nonexpansive operator is nonexpansive, but not vice versa. Furthermore, show that every nonexpansive operator is quasinonexpansive, but not vice versa.

**Exercise 4.4** Let  $T \in \mathcal{B}(\mathcal{H})$ . Show that the following are equivalent:

- (i)  $T$  is nonexpansive.
- (ii)  $\|T\| \leq 1$ .
- (iii)  $T$  is quasinonexpansive.

**Exercise 4.5** Let  $T \in \mathcal{B}(\mathcal{H})$  be such that  $\|T\| \leq 1$ .

(i) Suppose that

$$(\exists \alpha \in \mathbb{R}_{++})(\forall x \in \mathcal{H}) \quad \langle Tx \mid x \rangle \geq \alpha \|x\|^2. \quad (4.28)$$

Show that  $\alpha \|T\|^{-2}T$  is firmly nonexpansive.

(ii) Suppose that  $\mathcal{H} = \mathbb{R}^3$  and set  $T: (\xi_1, \xi_2, \xi_3) \mapsto (1/2)(\xi_1 - \xi_2, \xi_1 + \xi_2, 0)$ . Show that  $T$  does not satisfy (4.28), is not self-adjoint, but firmly nonexpansive.

**Exercise 4.6** Let  $T \in \mathcal{B}(\mathcal{H})$  be nonexpansive. Show that  $\text{Fix } T = \text{Fix } T^*$ .

**Exercise 4.7** Let  $\mathcal{K}$  be a real Hilbert space, let  $T: \mathcal{K} \rightarrow \mathcal{K}$  be firmly nonexpansive, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\|L\| \leq 1$ , let  $\bar{x} \in \mathcal{H}$ , and let  $\bar{z} \in \mathcal{K}$ . Show that  $x \mapsto \bar{x} + L^*T(\bar{z} + Lx)$  is firmly nonexpansive.

**Exercise 4.8** As seen in Proposition 4.2, if  $T: \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive, then so is  $\text{Id} - T$ . By way of examples, show that if  $C$  is a nonempty closed convex subset of  $\mathcal{H}$ , then  $\text{Id} - P_C$  may or may not be a projector.

**Exercise 4.9** Let  $C$  and  $D$  be closed linear subspaces of  $\mathcal{H}$  such that  $C \subset D$ . Show that  $P_C = P_D P_C = P_C P_D$ .

**Exercise 4.10** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be firmly nonexpansive, let  $\lambda \in \mathbb{R}$ , and set  $T_\lambda = \lambda T + (1 - \lambda)\text{Id}$ . Show that  $T_\lambda$  is nonexpansive for  $\lambda \in [0, 2]$  and that this interval is the largest possible with this property.

**Exercise 4.11** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $(T_i)_{i \in I}$  be a finite family of operators from  $D$  to  $\mathcal{H}$ , and let  $(\omega_i)_{i \in I}$  be strictly positive real numbers such that  $\sum_{i \in I} \omega_i = 1$ . Set  $T = \sum_{i \in I} \omega_i T_i$ .

(i) Suppose that  $\bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$  and that each  $T_i$  is quasinonexpansive. Show that  $T$  is quasinonexpansive.

(ii) Suppose that each  $T_i$  is nonexpansive. Show that  $T$  is nonexpansive.

**Exercise 4.12** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , and let  $T_1$  and  $T_2$  be firmly nonexpansive operators from  $D$  to  $\mathcal{H}$ . Show that  $T_1 - T_2$  and  $\text{Id} - T_1 - T_2$  are nonexpansive.

**Exercise 4.13** Let  $T$ ,  $T_1$ , and  $T_2$  be operators from  $\mathcal{H}$  to  $\mathcal{H}$ .

(i) Show that if  $T$  is firmly nonexpansive, then  $T^2$  may fail to be firmly nonexpansive even when  $\text{Fix } T \neq \emptyset$ . Compare with Exercise 4.2.

(ii) Show that if  $T_1$  and  $T_2$  are both nonexpansive, then so is  $T_2 T_1$ .

(iii) Show that if  $T$  is quasinonexpansive, then  $T^2$  may fail to be quasinonexpansive even when  $\mathcal{H} = \mathbb{R}$  and  $\text{Fix } T \neq \emptyset$ .

**Exercise 4.14** Provide an example of two closed linear subspaces  $U$  and  $V$  of  $\mathcal{H}$  such that the composition  $T = P_U P_V$  fails to be firmly nonexpansive. Conclude, in particular, that  $T$  is not a projector.

**Exercise 4.15** Let  $D$  be a nonempty compact subset of  $\mathcal{H}$  and suppose that  $T: D \rightarrow \mathcal{H}$  is firmly nonexpansive and that  $\text{Id} - T$  is injective. Show that for every  $\delta \in \mathbb{R}_{++}$ , there exists  $\beta \in [0, 1[$  such that if  $x$  and  $y$  belong to  $D$  and  $\|x - y\| \geq \delta$ , then  $\|Tx - Ty\| \leq \beta\|x - y\|$ . In addition, provide, for  $\mathcal{H} = \mathbb{R}$ , a set  $D$  and an operator  $T$  such that the hypothesis holds and such that for every  $\beta \in [0, 1[$ ,  $T$  is not Lipschitz continuous with constant  $\beta$ .

**Exercise 4.16** Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be nonexpansive. Use Corollary 4.16 to show that  $\text{Fix } T$  is closed and convex.

**Exercise 4.17** Provide a simple proof of Theorem 4.17 for the case when  $\mathcal{H}$  is finite-dimensional.

**Exercise 4.18** Show that each of the following assumptions on  $D$  in Theorem 4.19 is necessary: boundedness, closedness, convexity.

**Exercise 4.19** Use items (ix) and (x) of Proposition 4.20 to prove Theorem 4.19 without using Corollary 4.18.

**Exercise 4.20** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , and let  $(T_i)_{1 \leq i \leq m}$  be a finite family of quasinonexpansive operators from  $D$  to  $D$  such that  $\bigcap_{i=1}^m \text{Fix } T_i \neq \emptyset$  and  $m - 1$  of these operators are strictly quasinonexpansive. Then  $T_1 \cdots T_m$  is quasinonexpansive and  $\text{Fix } T_1 \cdots T_m = \bigcap_{i=1}^m \text{Fix } T_i$ .

# Chapter 5

## Fejér Monotonicity and Fixed Point Iterations

A sequence is Fejér monotone with respect to a set  $C$  if each point in the sequence is not strictly farther from any point in  $C$  than its predecessor. Such sequences possess very attractive properties that greatly simplify the analysis of their asymptotic behavior. In this chapter, we provide the basic theory for Fejér monotone sequences and apply it to obtain in a systematic fashion convergence results for various classical iterations involving nonexpansive operators.

### 5.1 Fejér Monotone Sequences

The following notion is central in the study of various iterative methods, in particular in connection with the construction of fixed points of nonexpansive operators.

**Definition 5.1** Let  $C$  be a nonempty subset of  $\mathcal{H}$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$ . Then  $(x_n)_{n \in \mathbb{N}}$  is *Fejér monotone* with respect to  $C$  if

$$(\forall x \in C)(\forall n \in \mathbb{N}) \quad \|x_{n+1} - x\| \leq \|x_n - x\|. \quad (5.1)$$

**Example 5.2** Let  $(x_n)_{n \in \mathbb{N}}$  be a bounded sequence in  $\mathbb{R}$  that is increasing (respectively decreasing). Then  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $[\sup\{x_n\}_{n \in \mathbb{N}}, +\infty[$  (respectively  $]-\infty, \inf\{x_n\}_{n \in \mathbb{N}}]$ ).

**Example 5.3** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow D$  be a quasinonexpansive—in particular, nonexpansive—operator such that  $\text{Fix } T \neq \emptyset$ , and let  $x_0 \in D$ . Set  $(\forall n \in \mathbb{N}) \ x_{n+1} = Tx_n$ . Then  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .

We start with some basic properties.

**Proposition 5.4** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty subset of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  is bounded.
- (ii) For every  $x \in C$ ,  $(\|x_n - x\|)_{n \in \mathbb{N}}$  converges.
- (iii)  $(d_C(x_n))_{n \in \mathbb{N}}$  is decreasing and converges.

*Proof.* (i): Let  $x \in C$ . Then (5.1) implies that  $(x_n)_{n \in \mathbb{N}}$  lies in  $B(x; \|x_0 - x\|)$ .

(ii): Clear from (5.1).

(iii): Taking the infimum in (5.1) over  $x \in C$  yields  $(\forall n \in \mathbb{N}) d_C(x_{n+1}) \leq d_C(x_n)$ .  $\square$

The next result concerns weak convergence.

**Theorem 5.5** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty subset of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$  and that every weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$  belongs to  $C$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .*

*Proof.* The result follows from Proposition 5.4(ii) and Lemma 2.39.  $\square$

**Example 5.6** Suppose that  $\mathcal{H}$  is infinite-dimensional and let  $(x_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ . Then  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\{0\}$ . As seen in Example 2.25,  $x_n \rightharpoonup 0$  but  $x_n \not\rightarrow 0$ .

While a Fejér monotone sequence with respect to a closed convex set  $C$  may not converge strongly, its “shadow” on  $C$  always does.

**Proposition 5.7** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then the shadow sequence  $(P_C x_n)_{n \in \mathbb{N}}$  converges strongly to a point in  $C$ .*

*Proof.* It follows from (5.1) and (3.6) that, for every  $m$  and  $n$  in  $\mathbb{N}$ ,

$$\begin{aligned}
 \|P_C x_n - P_C x_{n+m}\|^2 &= \|P_C x_n - x_{n+m}\|^2 + \|x_{n+m} - P_C x_{n+m}\|^2 \\
 &\quad + 2 \langle P_C x_n - x_{n+m} \mid x_{n+m} - P_C x_{n+m} \rangle \\
 &\leq \|P_C x_n - x_n\|^2 + d_C^2(x_{n+m}) \\
 &\quad + 2 \langle P_C x_n - P_C x_{n+m} \mid x_{n+m} - P_C x_{n+m} \rangle \\
 &\quad + 2 \langle P_C x_{n+m} - x_{n+m} \mid x_{n+m} - P_C x_{n+m} \rangle \\
 &\leq d_C^2(x_n) - d_C^2(x_{n+m}).
 \end{aligned} \tag{5.2}$$

Consequently, since  $(d_C(x_n))_{n \in \mathbb{N}}$  was seen in Proposition 5.4(iii) to converge,  $(P_C x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence in the complete set  $C$ .  $\square$

**Corollary 5.8** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$ , let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , and let  $x \in C$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$  and that  $x_n \rightharpoonup x$ . Then  $P_C x_n \rightarrow x$ .*

*Proof.* By Proposition 5.7,  $(P_C x_n)_{n \in \mathbb{N}}$  converges strongly to some point  $y \in C$ . Hence, since  $x - P_C x_n \rightarrow x - y$  and  $x_n - P_C x_n \rightarrow x - y$ , it follows from Theorem 3.14 and Lemma 2.41(iii) that  $0 \geq \langle x - P_C x_n \mid x_n - P_C x_n \rangle \rightarrow \|x - y\|^2$ . Thus,  $x = y$ .  $\square$

For sequences that are Fejér monotone with respect to closed affine subspaces, Proposition 5.7 can be strengthened.

**Proposition 5.9** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a closed affine subspace of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then the following hold:*

- (i)  $(\forall n \in \mathbb{N}) P_C x_n = P_C x_0$ .
- (ii) *Suppose that every weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$  belongs to  $C$ . Then  $x_n \rightarrow P_C x_0$ .*

*Proof.* (i): Fix  $n \in \mathbb{N}$ ,  $\alpha \in \mathbb{R}$ , and set  $y_\alpha = \alpha P_C x_0 + (1 - \alpha) P_C x_n$ . Since  $C$  is an affine subspace,  $y_\alpha \in C$ , and it therefore follows from Corollary 3.20(i) and (5.1) that

$$\begin{aligned}
 \alpha^2 \|P_C x_n - P_C x_0\|^2 &= \|P_C x_n - y_\alpha\|^2 \\
 &\leq \|x_n - P_C x_n\|^2 + \|P_C x_n - y_\alpha\|^2 \\
 &= \|x_n - y_\alpha\|^2 \\
 &\leq \|x_0 - y_\alpha\|^2 \\
 &= \|x_0 - P_C x_0\|^2 + \|P_C x_0 - y_\alpha\|^2 \\
 &= d_C^2(x_0) + (1 - \alpha)^2 \|P_C x_n - P_C x_0\|^2. \tag{5.3}
 \end{aligned}$$

Consequently,  $(2\alpha - 1)\|P_C x_n - P_C x_0\|^2 \leq d_C^2(x_0)$  and, letting  $\alpha \rightarrow +\infty$ , we conclude that  $P_C x_n = P_C x_0$ .

- (ii): Combine Theorem 5.5, Corollary 5.8, and (i).  $\square$

We now turn our attention to strong convergence properties.

**Proposition 5.10** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a subset of  $\mathcal{H}$  such that  $\text{int } C \neq \emptyset$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to a point in  $\mathcal{H}$ .*

*Proof.* Take  $x \in \text{int } C$  and  $\rho \in \mathbb{R}_{++}$  such that  $B(x; \rho) \subset C$ . Define a sequence  $(z_n)_{n \in \mathbb{N}}$  in  $B(x; \rho)$  by

$$(\forall n \in \mathbb{N}) \quad z_n = \begin{cases} x, & \text{if } x_{n+1} = x_n; \\ x - \rho \frac{x_{n+1} - x_n}{\|x_{n+1} - x_n\|}, & \text{otherwise.} \end{cases} \tag{5.4}$$

Then (5.1) yields  $(\forall n \in \mathbb{N}) \|x_{n+1} - z_n\|^2 \leq \|x_n - z_n\|^2$  and, after expanding, we obtain

$$(\forall n \in \mathbb{N}) \quad \|x_{n+1} - x\|^2 \leq \|x_n - x\|^2 - 2\rho\|x_{n+1} - x_n\|. \quad (5.5)$$

Thus,  $\sum_{n \in \mathbb{N}} \|x_{n+1} - x_n\| \leq \|x_0 - x\|^2 / (2\rho)$  and  $(x_n)_{n \in \mathbb{N}}$  is therefore a Cauchy sequence.  $\square$

**Theorem 5.11** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then the following are equivalent:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges strongly to a point in  $C$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  possesses a strong sequential cluster point in  $C$ .
- (iii)  $\liminf d_C(x_n) = 0$ .

*Proof.* (i) $\Rightarrow$ (ii): Clear.

(ii) $\Rightarrow$ (iii): Suppose that  $x_{k_n} \rightarrow x \in C$ . Then  $d_C(x_{k_n}) \leq \|x_{k_n} - x\| \rightarrow 0$ .

(iii) $\Rightarrow$ (i): Proposition 5.4(iii) implies that  $d_C(x_n) \rightarrow 0$ . Hence,  $x_n - P_C x_n \rightarrow 0$  and (i) follows from Proposition 5.7.  $\square$

We conclude this section with a linear convergence result.

**Theorem 5.12** *Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Suppose that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$  and that for some  $\kappa \in [0, 1[$ ,*

$$(\forall n \in \mathbb{N}) \quad d_C(x_{n+1}) \leq \kappa d_C(x_n). \quad (5.6)$$

*Then  $(x_n)_{n \in \mathbb{N}}$  converges linearly to a point  $x \in C$ ; more precisely,*

$$(\forall n \in \mathbb{N}) \quad \|x_n - x\| \leq 2\kappa^n d_C(x_0). \quad (5.7)$$

*Proof.* Theorem 5.11 and (5.6) imply that  $(x_n)_{n \in \mathbb{N}}$  converges strongly to some point  $x \in C$ . On the other hand, (5.1) yields

$$\begin{aligned} (\forall n \in \mathbb{N})(\forall m \in \mathbb{N}) \quad \|x_n - x_{n+m}\| &\leq \|x_n - P_C x_n\| + \|x_{n+m} - P_C x_n\| \\ &\leq 2d_C(x_n). \end{aligned} \quad (5.8)$$

Letting  $m \rightarrow +\infty$  in (5.8), we conclude that  $\|x_n - x\| \leq 2d_C(x_n)$ .  $\square$

## 5.2 Krasnosel'skiĭ–Mann Iteration

Given a nonexpansive operator  $T$ , the sequence generated by the Banach–Picard iteration  $x_{n+1} = Tx_n$  of (1.67) may fail to produce a fixed point of  $T$ . A simple illustration of this situation is  $T = -\text{Id}$  and  $x_0 \neq 0$ . In this case, however, it is clear that the *asymptotic regularity* property  $x_n - Tx_n \rightarrow 0$  does not hold. As we shall now see, this property is critical.



**Theorem 5.13** *Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $T: D \rightarrow D$  be a nonexpansive operator such that  $\text{Fix } T \neq \emptyset$ , and let  $x_0 \in D$ . Set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Tx_n \quad (5.9)$$

*and suppose that  $x_n - Tx_n \rightarrow 0$ . Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .
- (ii) Suppose that  $D = -D$  and that  $T$  is odd:  $(\forall x \in D) \quad T(-x) = -Tx$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to a point in  $\text{Fix } T$ .

*Proof.* From Example 5.3,  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .

(i): Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ , say  $x_{k_n} \rightharpoonup x$ . Since  $Tx_{k_n} - x_{k_n} \rightarrow 0$ , Corollary 4.18 asserts that  $x \in \text{Fix } T$ . Appealing to Theorem 5.5, the assertion is proved.

(ii): Since  $D = -D$  is convex,  $0 \in D$  and, since  $T$  is odd,  $0 \in \text{Fix } T$ . Therefore, by Fejér monotonicity,  $(\forall n \in \mathbb{N}) \quad \|x_{n+1}\| \leq \|x_n\|$ . Thus, there exists  $\ell \in \mathbb{R}_+$  such that  $\|x_n\| \downarrow \ell$ . Now let  $m \in \mathbb{N}$ . Then, for every  $n \in \mathbb{N}$ ,

$$\|x_{n+1+m} + x_{n+1}\| = \|Tx_{n+m} - T(-x_n)\| \leq \|x_{n+m} + x_n\|, \quad (5.10)$$

and, by the parallelogram identity,

$$\|x_{n+m} + x_n\|^2 = 2(\|x_{n+m}\|^2 + \|x_n\|^2) - \|x_{n+m} - x_n\|^2. \quad (5.11)$$

However, since  $Tx_n - x_n \rightarrow 0$ , we have  $\lim_n \|x_{n+m} - x_n\| = 0$ . Therefore, since  $\|x_n\| \downarrow \ell$ , (5.10) and (5.11) yield  $\|x_{n+m} + x_n\| \downarrow 2\ell$  as  $n \rightarrow +\infty$ . In turn, we derive from (5.11) that  $\|x_{n+m} - x_n\|^2 \leq 2(\|x_{n+m}\|^2 + \|x_n\|^2) - 4\ell^2 \rightarrow 0$  as  $m, n \rightarrow +\infty$ . Thus,  $(x_n)_{n \in \mathbb{N}}$  is a Cauchy sequence and  $x_n \rightarrow x$  for some  $x \in D$ . Since  $x_{n+1} \rightarrow x$  and  $x_{n+1} = Tx_n \rightarrow Tx$ , we have  $x \in \text{Fix } T$ .  $\square$

We now turn our attention to an alternative iterative method, known as the *Krasnosel'skiĭ–Mann algorithm*.

**Theorem 5.14 (Krasnosel'skiĭ–Mann algorithm)** *Let  $D$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $T: D \rightarrow D$  be a nonexpansive operator such that  $\text{Fix } T \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 1]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(1 - \lambda_n) = +\infty$ , and let  $x_0 \in D$ . Set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n(Tx_n - x_n). \quad (5.12)$$

*Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .
- (ii)  $(Tx_n - x_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (iii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .

*Proof.* Since  $x_0 \in D$  and  $D$  is convex, (5.12) produces a well-defined sequence in  $D$ .

(i): It follows from Corollary 2.14 and the nonexpansiveness of  $T$  that, for every  $y \in \text{Fix } T$  and  $n \in \mathbb{N}$ ,

$$\begin{aligned} \|x_{n+1} - y\|^2 &= \|(1 - \lambda_n)(x_n - y) + \lambda_n(Tx_n - y)\|^2 \\ &= (1 - \lambda_n)\|x_n - y\|^2 + \lambda_n\|Tx_n - Ty\|^2 \\ &\quad - \lambda_n(1 - \lambda_n)\|Tx_n - x_n\|^2 \\ &\leq \|x_n - y\|^2 - \lambda_n(1 - \lambda_n)\|Tx_n - x_n\|^2. \end{aligned} \quad (5.13)$$

Hence,  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .

(ii): We derive from (5.13) that  $\sum_{n \in \mathbb{N}} \lambda_n(1 - \lambda_n)\|Tx_n - x_n\|^2 \leq \|x_0 - y\|^2$ . Since  $\sum_{n \in \mathbb{N}} \lambda_n(1 - \lambda_n) = +\infty$ , we have  $\varliminf \|Tx_n - x_n\| = 0$ . However, for every  $n \in \mathbb{N}$ ,

$$\begin{aligned} \|Tx_{n+1} - x_{n+1}\| &= \|Tx_{n+1} - Tx_n + (1 - \lambda_n)(Tx_n - x_n)\| \\ &\leq \|x_{n+1} - x_n\| + (1 - \lambda_n)\|Tx_n - x_n\| \\ &= \|Tx_n - x_n\|. \end{aligned} \quad (5.14)$$

Consequently,  $(\|Tx_n - x_n\|)_{n \in \mathbb{N}}$  converges and we must have  $Tx_n - x_n \rightarrow 0$ .

(iii): Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ , say  $x_{k_n} \rightharpoonup x$ . Then it follows from Corollary 4.18 that  $x \in \text{Fix } T$ . In view of Theorem 5.5, the proof is complete.  $\square$

**Proposition 5.15** *Let  $\alpha \in ]0, 1[$ , let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be an  $\alpha$ -averaged operator such that  $\text{Fix } T \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 1/\alpha]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(1 - \alpha\lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n(Tx_n - x_n). \quad (5.15)$$

*Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .
- (ii)  $(Tx_n - x_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (iii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .

*Proof.* Set  $R = (1 - 1/\alpha)\text{Id} + (1/\alpha)T$  and  $(\forall n \in \mathbb{N}) \mu_n = \alpha\lambda_n$ . Then  $\text{Fix } R = \text{Fix } T$  and  $R$  is nonexpansive by Proposition 4.25. In addition, we rewrite (5.15) as  $(\forall n \in \mathbb{N}) \ x_{n+1} = x_n + \mu_n(Rx_n - x_n)$ . Since  $(\mu_n)_{n \in \mathbb{N}}$  lies in  $[0, 1]$  and  $\sum_{n \in \mathbb{N}} \mu_n(1 - \mu_n) = +\infty$ , the results follow from Theorem 5.14.  $\square$

**Corollary 5.16** *Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be a firmly nonexpansive operator such that  $\text{Fix } T \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Set  $(\forall n \in \mathbb{N}) \ x_{n+1} = x_n + \lambda_n(Tx_n - x_n)$ . Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$ .
- (ii)  $(Tx_n - x_n)_{n \in \mathbb{N}}$  converges strongly to 0.

(iii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .

*Proof.* In view of Remark 4.24(iii), apply Proposition 5.15 with  $\alpha = 1/2$ .  $\square$

**Example 5.17** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be a firmly nonexpansive operator such that  $\text{Fix } T \neq \emptyset$ , let  $x_0 \in \mathcal{H}$ , and set  $(\forall n \in \mathbb{N}) \ x_{n+1} = Tx_n$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .

The following type of iterative method involves a mix of compositions and convex combinations of nonexpansive operators.

**Corollary 5.18** Let  $(T_i)_{i \in I}$  be a finite family of nonexpansive operators from  $\mathcal{H}$  to  $\mathcal{H}$  such that  $\bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , and let  $(\alpha_i)_{i \in I}$  be real numbers in  $]0, 1[$  such that, for every  $i \in I$ ,  $T_i$  is  $\alpha_i$ -averaged. Let  $p$  be a strictly positive integer, for every  $k \in \{1, \dots, p\}$ , let  $m_k$  be a strictly positive integer and  $\omega_k$  be a strictly positive real number, and suppose that  $i: \{(k, l) \mid k \in \{1, \dots, p\}, l \in \{1, \dots, m_k\}\} \rightarrow I$  is surjective and that  $\sum_{k=1}^p \omega_k = 1$ . For every  $k \in \{1, \dots, p\}$ , set  $I_k = \{i(k, 1), \dots, i(k, m_k)\}$ , and set

$$\alpha = \max_{1 \leq k \leq p} \rho_k, \quad \text{where } (\forall k \in \{1, \dots, p\}) \quad \rho_k = \frac{m_k}{m_k - 1 + \frac{1}{\max_{i \in I_k} \alpha_i}}, \quad (5.16)$$

and let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 1/\alpha]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n (1 - \alpha \lambda_n) = +\infty$ . Furthermore, let  $x_0 \in \mathcal{H}$  and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n \left( \sum_{k=1}^p \omega_k T_{i(k,1)} \cdots T_{i(k,m_k)} x_n - x_n \right). \quad (5.17)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\bigcap_{i \in I} \text{Fix } T_i$ .

*Proof.* Set  $T = \sum_{k=1}^p \omega_k R_k$ , where  $(\forall k \in \{1, \dots, p\}) \ R_k = T_{i(k,1)} \cdots T_{i(k,m_k)}$ . Then (5.17) reduces to (5.15) and, in view of Proposition 5.15, it suffices to show that  $T$  is  $\alpha$ -averaged and that  $\text{Fix } T = \bigcap_{i \in I} \text{Fix } T_i$ . For every  $k \in \{1, \dots, p\}$ , it follows from Proposition 4.32 and (5.16) that  $R_k$  is  $\rho_k$ -averaged and, from Corollary 4.37 that  $\text{Fix } R_k = \bigcap_{i \in I_k} \text{Fix } T_i$ . In turn, we derive from Proposition 4.30 and (5.16) that  $T$  is  $\alpha$ -averaged and, from Proposition 4.34, that  $\text{Fix } T = \bigcap_{k=1}^p \text{Fix } R_k = \bigcap_{k=1}^p \bigcap_{i \in I_k} \text{Fix } T_i = \bigcap_{i \in I} \text{Fix } T_i$ .  $\square$

**Remark 5.19** It follows from Remark 4.24(iii) that Corollary 5.18 is applicable to firmly nonexpansive operators and, a fortiori, to projection operators by Proposition 4.8.

Corollary 5.18 provides an algorithm to solve a *convex feasibility problem*, i.e., to find a point in the intersection of a family of closed convex sets. Here are two more examples.

**Example 5.20 (string-averaged relaxed projections)** Let  $(C_i)_{i \in I}$  be a finite family of closed convex sets such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ . For every  $i \in I$ , let  $\beta_i \in ]0, 2[$  and set  $T_i = (1 - \beta_i)\text{Id} + \beta_i P_{C_i}$ . Let  $p$  be a strictly positive integer; for every  $k \in \{1, \dots, p\}$ , let  $m_k$  be a strictly positive integer and  $\omega_k$  be a strictly positive real number, and suppose that  $i: \{(k, l) \mid k \in \{1, \dots, p\}, l \in \{1, \dots, m_k\}\} \rightarrow I$  is surjective and that  $\sum_{k=1}^p \omega_k = 1$ . Furthermore, let  $x_0 \in \mathcal{H}$  and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = \sum_{k=1}^p \omega_k T_{i(k,1)} \cdots T_{i(k,m_k)} x_n. \quad (5.18)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .

*Proof.* For every  $i \in I$ , set  $\alpha_i = \beta_i/2 \in ]0, 1[$ . Since, for every  $i \in I$ , Proposition 4.8 asserts that  $P_{C_i}$  is firmly nonexpansive, Corollary 4.29 implies that  $T_i$  is  $\alpha_i$ -averaged. Borrowing notation from Corollary 5.18, we note that for every  $k \in \{1, \dots, p\}$ ,  $\max_{i \in I_k} \alpha_i \in ]0, 1[$ , which implies that  $\rho_k \in ]0, 1[$  and thus that  $\alpha \in ]0, 1[$ . Altogether, the result follows from Corollary 5.18 with  $\lambda_n \equiv 1$ .  $\square$

**Example 5.21 (parallel projection algorithm)** Let  $(C_i)_{i \in I}$  be a finite family of closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $(\omega_i)_{i \in I}$  be strictly positive real numbers such that  $\sum_{i \in I} \omega_i = 1$ , and let  $x_0 \in \mathcal{H}$ . Set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n \left( \sum_{i \in I} \omega_i P_i x_n - x_n \right), \quad (5.19)$$

where, for every  $i \in I$ ,  $P_i$  denotes the projector onto  $C_i$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .

*Proof.* This is an application of Corollary 5.16(iii) with  $T = \sum_{i \in I} \omega_i P_i$ . Indeed, since the operators  $(P_i)_{i \in I}$  are firmly nonexpansive by Proposition 4.8, their convex combination  $T$  is also firmly nonexpansive by Example 4.31. Moreover, Proposition 4.34 asserts that  $\text{Fix } T = \bigcap_{i \in I} \text{Fix } P_i = \bigcap_{i \in I} C_i = C$ . Alternatively, apply Corollary 5.18.  $\square$

### 5.3 Iterating Compositions of Averaged Operators

Our first result concerns the asymptotic behavior of iterates of a composition of averaged nonexpansive operators with possibly no common fixed point.

**Theorem 5.22** *Let  $D$  be a nonempty weakly sequentially closed (e.g., closed and convex) subset of  $\mathcal{H}$ , let  $m$  be a strictly positive integer, set  $I =$*

$\{1, \dots, m\}$ , let  $(T_i)_{i \in I}$  be a family of nonexpansive operators from  $D$  to  $D$  such that  $\text{Fix}(T_1 \cdots T_m) \neq \emptyset$ , and let  $(\alpha_i)_{i \in I}$  be real numbers in  $]0, 1[$  such that, for every  $i \in I$ ,  $T_i$  is  $\alpha_i$ -averaged. Let  $x_0 \in D$  and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = T_1 \cdots T_m x_n. \quad (5.20)$$

Then  $x_n - T_1 \cdots T_m x_n \rightarrow 0$ , and there exist points  $y_1 \in \text{Fix } T_1 \cdots T_m$ ,  $y_2 \in \text{Fix } T_2 \cdots T_m T_1$ ,  $\dots$ ,  $y_m \in \text{Fix } T_m T_1 \cdots T_{m-1}$  such that

$$x_n \rightharpoonup y_1 = T_1 y_2, \quad (5.21)$$

$$T_m x_n \rightharpoonup y_m = T_m y_1, \quad (5.22)$$

$$T_{m-1} T_m x_n \rightharpoonup y_{m-1} = T_{m-1} y_m, \quad (5.23)$$

$$\vdots$$

$$T_3 \cdots T_m x_n \rightharpoonup y_3 = T_3 y_4, \quad (5.24)$$

$$T_2 \cdots T_m x_n \rightharpoonup y_2 = T_2 y_3. \quad (5.25)$$

*Proof.* Set  $T = T_1 \cdots T_m$  and  $(\forall i \in I) \beta_i = (1 - \alpha_i)/\alpha_i$ . Now take  $y \in \text{Fix } T$ . The equivalence (i)  $\Leftrightarrow$  (iii) in Proposition 4.25 yields

$$\begin{aligned} \|x_{n+1} - y\|^2 &= \|T x_n - T y\|^2 \\ &\leq \|T_2 \cdots T_m x_n - T_2 \cdots T_m y\|^2 \\ &\quad - \beta_1 \|(\text{Id} - T_1) T_2 \cdots T_m x_n - (\text{Id} - T_1) T_2 \cdots T_m y\|^2 \\ &\leq \|x_n - y\|^2 - \beta_m \|(\text{Id} - T_m) x_n - (\text{Id} - T_m) y\|^2 \\ &\quad - \beta_{m-1} \|(\text{Id} - T_{m-1}) T_m x_n - (\text{Id} - T_{m-1}) T_m y\|^2 - \dots \\ &\quad - \beta_2 \|(\text{Id} - T_2) T_3 \cdots T_m x_n - (\text{Id} - T_2) T_3 \cdots T_m y\|^2 \\ &\quad - \beta_1 \|(\text{Id} - T_1) T_2 \cdots T_m x_n - (\text{Id} - T_1) T_2 \cdots T_m y\|^2. \end{aligned} \quad (5.26)$$

Therefore,  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\text{Fix } T$  and

$$(\text{Id} - T_m) x_n - (\text{Id} - T_m) y \rightarrow 0, \quad (5.27)$$

$$(\text{Id} - T_{m-1}) T_m x_n - (\text{Id} - T_{m-1}) T_m y \rightarrow 0, \quad (5.28)$$

$$\vdots$$

$$(\text{Id} - T_2) T_3 \cdots T_m x_n - (\text{Id} - T_2) T_3 \cdots T_m y \rightarrow 0, \quad (5.29)$$

$$(\text{Id} - T_1) T_2 \cdots T_m x_n - (\text{Id} - T_1) T_2 \cdots T_m y \rightarrow 0. \quad (5.30)$$

Upon adding (5.27)–(5.30), we obtain  $x_n - T x_n \rightarrow 0$ . Hence, since  $T$  is nonexpansive as a composition of nonexpansive operators, it follows from Theorem 5.13(i) that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to some point  $y_1 \in \text{Fix } T$ , which provides (5.21). On the other hand, (5.27) yields  $T_m x_n - x_n \rightarrow T_m y_1 - y_1$ . So altogether  $T_m x_n \rightharpoonup T_m y_1 = y_m$ , and we obtain (5.22). In turn, since (5.28) asserts that  $T_{m-1} T_m x_n - T_m x_n \rightarrow T_{m-1} y_m - y_m$ , we ob-

tain  $T_{m-1}T_mx_n \rightharpoonup T_{m-1}y_m = y_{m-1}$ , hence (5.23). Continuing this process, we arrive at (5.25).  $\square$

As noted in Remark 5.19, results on averaged nonexpansive operators apply in particular to firmly nonexpansive operators and projectors onto convex sets. Thus, by specializing Theorem 5.22 to convex projectors, we obtain the iterative method described in the next corollary, which is known as the POCS (Projections Onto Convex Sets) algorithm in the signal recovery literature.

**Corollary 5.23 (POCS algorithm)** *Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of nonempty closed convex subsets of  $\mathcal{H}$ , let  $(P_i)_{i \in I}$  denote their respective projectors, and let  $x_0 \in \mathcal{H}$ . Suppose that  $\text{Fix}(P_1 \cdots P_m) \neq \emptyset$  and set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = P_1 \cdots P_m x_n. \quad (5.31)$$

*Then there exists  $(y_1, \dots, y_m) \in C_1 \times \cdots \times C_m$  such that  $x_n \rightharpoonup y_1 = P_1 y_2$ ,  $P_m x_n \rightharpoonup y_m = P_m y_1$ ,  $P_{m-1} P_m x_n \rightharpoonup y_{m-1} = P_{m-1} y_m$ ,  $\dots$ ,  $P_3 \cdots P_m x_n \rightharpoonup y_3 = P_3 y_4$ , and  $P_2 \cdots P_m x_n \rightharpoonup y_2 = P_2 y_3$ .*

*Proof.* This follows from Proposition 4.8 and Theorem 5.22.  $\square$

**Remark 5.24** In Corollary 5.23, suppose that, for some  $j \in I$ ,  $C_j$  is bounded. Then  $\text{Fix}(P_1 \cdots P_m) \neq \emptyset$ . Indeed, consider the circular composition of the  $m$  projectors given by  $T = P_j \cdots P_m P_1 \cdots P_{j-1}$ . Then Proposition 4.8 asserts that  $T$  is a nonexpansive operator that maps the nonempty bounded closed convex set  $C_j$  to itself. Hence, it follows from Theorem 4.19 that there exists a point  $x \in C_j$  such that  $Tx = x$ .

The next corollary describes a periodic projection method to solve a convex feasibility problem.

**Corollary 5.25** *Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , let  $(P_i)_{i \in I}$  denote their respective projectors, and let  $x_0 \in \mathcal{H}$ . Set  $(\forall n \in \mathbb{N})$   $x_{n+1} = P_1 \cdots P_m x_n$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .*

*Proof.* Using Corollary 5.23, Proposition 4.8, and Corollary 4.37, we obtain  $x_n \rightharpoonup y_1 \in \text{Fix}(P_1 \cdots P_m) = \bigcap_{i \in I} \text{Fix } P_i = C$ . Alternatively, this is a special case of Example 5.20.  $\square$

**Remark 5.26** If, in Corollary 5.25, all the sets are closed affine subspaces, so is  $C$  and we derive from Proposition 5.9(i) that  $x_n \rightharpoonup_{P_C} x_0$ . Corollary 5.28 is classical, and it states that the convergence is actually strong in this case. In striking contrast, the example constructed in [146] provides a closed hyperplane and a closed convex cone in  $\ell^2(\mathbb{N})$  for which alternating projections converge weakly but not strongly.

The next result will help us obtain a sharper form of Corollary 5.25 for closed affine subspaces.

**Proposition 5.27** *Let  $T \in \mathcal{B}(\mathcal{H})$  be nonexpansive and let  $x_0 \in \mathcal{H}$ . Set  $V = \text{Fix } T$  and  $(\forall n \in \mathbb{N}) \ x_{n+1} = Tx_n$ . Then  $x_n \rightarrow P_V x_0 \Leftrightarrow x_n - x_{n+1} \rightarrow 0$ .*

*Proof.* If  $x_n \rightarrow P_V x_0$ , then  $x_n - x_{n+1} \rightarrow P_V x_0 - P_V x_0 = 0$ . Conversely, suppose that  $x_n - x_{n+1} \rightarrow 0$ . We derive from Theorem 5.13(ii) that there exists  $v \in V$  such that  $x_n \rightarrow v$ . In turn, Proposition 5.9(i) yields  $v = P_V x_0$ .  $\square$

**Corollary 5.28 (von Neumann–Halperin)** *Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of closed affine subspaces of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , let  $(P_i)_{i \in I}$  denote their respective projectors, let  $x_0 \in \mathcal{H}$ , and set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = P_1 \cdots P_m x_n. \quad (5.32)$$

*Then  $x_n \rightarrow P_C x_0$ .*

*Proof.* Set  $T = P_1 \cdots P_m$ . Then  $T$  is nonexpansive, and  $\text{Fix } T = C$  by Corollary 4.37.

We first assume that each set  $C_i$  is a linear subspace. Then  $T$  is odd, and Theorem 5.22 implies that  $x_n - Tx_n \rightarrow 0$ . Thus, by Proposition 5.27,  $x_n \rightarrow P_C x_0$ .

We now turn our attention to the general affine case. Since  $C \neq \emptyset$ , there exists  $y \in C$  such that for every  $i \in I$ ,  $C_i = y + V_i$ , i.e.,  $V_i$  is the closed linear subspace parallel to  $C_i$ , and  $C = y + V$ , where  $V = \bigcap_{i \in I} V_i$ . Proposition 3.17 implies that, for every  $x \in \mathcal{H}$ ,  $P_C x = P_{y+V} x = y + P_V(x - y)$  and  $(\forall i \in I) \ P_i x = P_{y+V_i} x = y + P_{V_i}(x - y)$ . Using these identities repeatedly, we obtain

$$(\forall n \in \mathbb{N}) \quad x_{n+1} - y = (P_{V_1} \cdots P_{V_m})(x_n - y). \quad (5.33)$$

Invoking the already verified linear case, we get  $x_n - y \rightarrow P_V(x_0 - y)$  and conclude that  $x_n \rightarrow y + P_V(x_0 - y) = P_C x_0$ .  $\square$

## Exercises

**Exercise 5.1** Find a nonexpansive operator  $T: \mathcal{H} \rightarrow \mathcal{H}$  that is not firmly nonexpansive and such that, for every  $x_0 \in \mathcal{H}$ , the sequence  $(T^n x_0)_{n \in \mathbb{N}}$  converges weakly but not strongly to a fixed point of  $T$ .

**Exercise 5.2** Construct a non-Cauchy sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathbb{R}$  that is asymptotically regular, i.e.,  $x_n - x_{n+1} \rightarrow 0$ .

**Exercise 5.3** Find an alternative proof of Theorem 5.5 based on Corollary 5.8 in the case when  $C$  is closed and convex.

**Exercise 5.4** Let  $C$  be a nonempty subset of  $\mathcal{H}$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  that is Fejér monotone with respect to  $C$ . Show that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\overline{\text{conv}} C$ .

**Exercise 5.5** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be a nonexpansive operator such that  $\text{Fix } T \neq \emptyset$ , and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  such that

- (i) for every  $x \in \text{Fix } T$ ,  $(\|x_n - x\|)_{n \in \mathbb{N}}$  converges;
- (ii)  $x_n - Tx_n \rightarrow 0$ .

Show that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Fix } T$ .

**Exercise 5.6** Find a nonexpansive operator  $T: \mathcal{H} \rightarrow \mathcal{H}$  that is not firmly nonexpansive and such that, for every  $x_0 \in \mathcal{H}$ , the sequence  $(T^n x_0)_{n \in \mathbb{N}}$  converges weakly but not strongly to a fixed point of  $T$ .

**Exercise 5.7** Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , and let  $(P_i)_{i \in I}$  be their respective projectors. Derive parts (ii) and (iii) from (i) and Theorem 5.5, and also from Corollary 5.18.

- (i) Let  $i \in I$ , let  $x \in C_i$ , and let  $y \in \mathcal{H}$ . Show that  $\|P_i y - x\|^2 \leq \|y - x\|^2 - \|P_i y - y\|^2$ .
- (ii) Set  $x_0 \in \mathcal{H}$  and

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = \frac{1}{m} (P_1 x_n + P_1 P_2 x_n + \dots + P_1 \dots P_m x_n). \quad (5.34)$$

- (a) Let  $x \in C$  and  $n \in \mathbb{N}$ . Show that  $\|x_{n+1} - x\|^2 \leq \|x_n - x\|^2 - (1/m) \sum_{i \in I} \|P_i x_n - x\|^2$ .
- (b) Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ . Show that  $x \in C$ .
- (c) Show that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .

- (iii) Set  $x_0 \in \mathcal{H}$  and

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = \frac{1}{m-1} (P_1 P_2 x_n + P_2 P_3 x_n + \dots + P_{m-1} P_m x_n). \quad (5.35)$$

- (a) Let  $x \in C$  and  $n \in \mathbb{N}$ . Show that  $\|x_{n+1} - x\|^2 \leq \|x_n - x\|^2 - \sum_{i=1}^{m-1} (\|P_{i+1} x_n - x_n\|^2 + \|P_i P_{i+1} x_n - P_{i+1} x_n\|^2) / (m-1)$ .
- (b) Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ . Show that  $x \in C$ .
- (c) Show that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $C$ .



# Chapter 6

## Convex Cones and Generalized Interiors

The notion of a convex cone, which lies between that of a linear subspace and that of a convex set, is the main topic of this chapter. It has been very fruitful in many branches of nonlinear analysis. For instance, closed convex cones provide decompositions analogous to the well-known orthogonal decomposition based on closed linear subspaces. They also arise naturally in convex analysis in the local study of a convex set via the tangent cone and the normal cone operators, and they are central in the analysis of various extensions of the notion of an interior that will be required in later chapters.

### 6.1 Convex Cones

Recall from (1.1) that a subset  $C$  of  $\mathcal{H}$  is a cone if  $C = \mathbb{R}_{++}C$ . Hence,  $\mathcal{H}$  is a cone and the intersection of a family of cones is a cone. The following notions are therefore well defined.

**Definition 6.1** Let  $C$  be a subset of  $\mathcal{H}$ . The *conical hull* of  $C$  is the intersection of all the cones in  $\mathcal{H}$  containing  $C$ , i.e., the smallest cone in  $\mathcal{H}$  containing  $C$ . It is denoted by  $\text{cone } C$ . The *closed conical hull* of  $C$  is the smallest closed cone in  $\mathcal{H}$  containing  $C$ . It is denoted by  $\overline{\text{cone}} C$ .

**Proposition 6.2** Let  $C$  be a subset of  $\mathcal{H}$ . Then the following hold:

- (i)  $\text{cone } C = \mathbb{R}_{++}C$ .
- (ii)  $\overline{\text{cone}} C = \overline{\text{cone}} C$ .
- (iii)  $\text{cone}(\text{conv } C) = \text{conv}(\text{cone } C)$  is the smallest convex cone containing  $C$ .
- (iv)  $\overline{\text{cone}}(\text{conv } C) = \overline{\text{conv}}(\text{cone } C)$  is the smallest closed convex cone containing  $C$ .

*Proof.* Since the results are clear when  $C = \emptyset$ , we assume that  $C \neq \emptyset$ .

(i): Let  $D = \mathbb{R}_{++}C$ . Then  $D$  is a cone and  $C \subset D$ . Therefore  $\text{cone } C \subset \text{cone } D = D$ . Conversely, take  $y \in D$ , say  $y = \lambda x$ , where  $\lambda \in \mathbb{R}_{++}$  and  $x \in C$ . Then  $x \in \text{cone } C$  and therefore  $y = \lambda x \in \text{cone } C$ . Thus,  $D \subset \text{cone } C$ .

(ii): Since  $\overline{\text{cone } C}$  is a closed cone and  $C \subset \overline{\text{cone } C}$ , we have  $\overline{\text{cone } C} \subset \text{cone}(\overline{\text{cone } C}) = \overline{\text{cone } C}$ . Conversely, since the closure of a cone is a cone, we have  $\overline{\text{cone } C} \subset \overline{\text{cone } C}$ .

(iii): Take  $x \in \text{cone}(\text{conv } C)$ . Proposition 3.4 and (i) imply the existence of  $\lambda \in \mathbb{R}_{++}$ , of a finite family  $(\alpha_i)_{i \in I}$  in  $\mathbb{R}_{++}$ , and of a family  $(x_i)_{i \in I}$  in  $C$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $x = \lambda \sum_{i \in I} \alpha_i x_i$ . Thus,  $x = \sum_{i \in I} \alpha_i (\lambda x_i) \in \text{conv}(\text{cone } C)$ . Therefore,  $\text{cone}(\text{conv } C) \subset \text{conv}(\text{cone } C)$ . Conversely, take  $x \in \text{conv}(\text{cone } C)$ . Proposition 3.4 and (i) imply the existence of finite families  $(\alpha_i)_{i \in I}$  in  $\mathbb{R}_{++}$ ,  $(\lambda_i)_{i \in I}$  in  $\mathbb{R}_{++}$ , and  $(x_i)_{i \in I}$  in  $C$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $x = \sum_{i \in I} \alpha_i \lambda_i x_i$ . Set  $(\forall i \in I) \beta_i = \alpha_i \lambda_i$ , and  $\lambda = \sum_{i \in I} \beta_i$ . Then  $\sum_{i \in I} \beta_i \lambda^{-1} x_i \in \text{conv } C$  and hence  $x = \lambda \sum_{i \in I} \beta_i \lambda^{-1} x_i \in \text{cone}(\text{conv } C)$ . Now let  $K$  be the smallest convex cone containing  $C$ . Since  $K$  is a convex cone and  $C \subset K$ , we have  $\text{conv}(\text{cone } C) \subset \text{conv}(\text{cone } K) = K$ . On the other hand,  $\text{conv}(\text{cone } C) = \text{cone}(\text{conv } C)$  is also a convex cone containing  $C$  and, therefore,  $K \subset \text{conv}(\text{cone } C)$ .

(iv): It follows from (iii) that  $\overline{\text{cone}}(\text{conv } C) = \overline{\text{conv}}(\text{cone } C)$ . Denote this set by  $D$  and denote the smallest closed convex cone containing  $C$  by  $K$ . Then  $D$  is closed and contains  $C$ . Exercise 3.6 and (ii) imply that  $D$  is a convex cone. Thus  $K \subset D$ . On the other hand, (iii) yields  $\text{cone}(\text{conv } C) = \text{conv}(\text{cone } C) \subset K$ . Taking closures, we deduce that  $D \subset K$ . Altogether,  $K = D$ .  $\square$

Convex cones are of particular importance due to their ubiquity in convex analysis. We record two simple propositions, the proofs of which we leave as Exercise 6.2 and Exercise 6.3.

**Proposition 6.3** *Let  $C$  be a subset of  $\mathcal{H}$ . Then the following hold:*

- (i) *Suppose that  $C$  is a cone. Then  $C$  is convex if and only if  $C + C \subset C$ .*
- (ii) *Suppose that  $C$  is convex and that  $0 \in C$ . Then  $C$  is a cone if and only if  $C + C \subset C$ .*

**Proposition 6.4** *Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . Then the following hold:*

- (i)  $\text{span } C = \text{cone } C - \text{cone } C = \text{cone } C + \text{cone}(-C)$ .
- (ii) *Suppose that  $C = -C$ . Then  $\text{span } C = \text{cone } C$ .*

Next, we introduce two important properties of convex cones.

**Definition 6.5** Let  $K$  be a convex cone in  $\mathcal{H}$ . Then  $K$  is *pointed* if  $K \cap (-K) \subset \{0\}$ , and  $K$  is *solid* if  $\text{int } K \neq \emptyset$ .

Note that  $\{0\}$  is the only pointed linear subspace and  $\mathcal{H}$  is the only solid linear subspace. The next examples illustrate the fact that various important cones are pointed or solid.

**Example 6.6** Suppose that  $u \in \mathcal{H} \setminus \{0\}$  and set  $K = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq 0\}$ . Then  $K$  is a solid convex cone, and it is not pointed if  $\dim \mathcal{H} > 1$ .

*Proof.* It is straightforward to check that  $K$  is a convex cone and that  $\{u\}^\perp \subset K$ . Moreover, since Cauchy–Schwarz implies that  $B(-u; \|u\|) \subset K$ ,  $K$  is solid. Finally, take  $x \in \{u\}^\perp$  such that  $x \neq 0$ . Then  $\{0\} \neq \text{span}\{x\} \subset K \cap (-K)$ . Thus,  $K$  is not pointed.  $\square$

**Example 6.7** Let  $I$  be a totally ordered set, suppose that  $\mathcal{H} = \ell^2(I)$ , and let  $(e_i)_{i \in I}$  be the standard unit vectors (see (2.8)). Then

$$\ell_+^2(I) = \{(\xi_i)_{i \in I} \in \ell^2(I) \mid (\forall i \in I) \ \xi_i \geq 0\} = \overline{\text{cone}} \text{conv}\{e_i\}_{i \in I} \quad (6.1)$$

is a nonempty closed convex pointed cone, and so is  $\ell_-^2(I) = -\ell_+^2(I)$ . Furthermore,  $\ell_+^2(I)$  is solid if and only if  $I$  is finite. In particular, the positive orthant  $\mathbb{R}_+^N$  is a closed convex cone in  $\mathbb{R}^N$  that is pointed and solid.

*Proof.* It is clear that  $\ell_+^2(I) = \bigcap_{i \in I} \{x \in \ell^2(I) \mid \langle x \mid e_i \rangle \geq 0\}$  is a nonempty pointed closed convex cone. Hence, since  $\{e_i\}_{i \in I} \subset \ell_+^2(I) \subset \overline{\text{cone}} \text{conv}\{e_i\}_{i \in I}$ , we obtain (6.1). If  $I$  is finite, then  $B((1)_{i \in I}; 1) \subset \ell_+^2(I)$ , and hence  $\ell_+^2(I)$  is solid. Now assume that  $I$  is infinite and that  $\ell_+^2(I)$  is solid. Then there exist  $x = (\xi_i)_{i \in I} \in \ell_+^2(I)$  and  $\varepsilon \in \mathbb{R}_{++}$  such that  $B(x; 2\varepsilon) \subset \ell_+^2(I)$ . Since  $I$  is infinite, there exists  $j \in I$  such that  $\xi_j \leq \varepsilon$ . On the one hand,  $(\eta_i)_{i \in I} = x - 2\varepsilon e_j \in B(x; 2\varepsilon) \subset \ell_+^2(I)$ . On the other hand,  $\eta_j = \xi_j - 2\varepsilon \leq -\varepsilon$ , which implies that  $(\eta_i)_{i \in I} \notin \ell_+^2(I)$ . We therefore arrive at a contradiction.  $\square$

**Proposition 6.8** Let  $\{x_i\}_{i \in I}$  be a nonempty finite subset of  $\mathcal{H}$  and set

$$K = \sum_{i \in I} \mathbb{R}_+ x_i. \quad (6.2)$$

Then  $K$  is the smallest closed convex cone containing  $\{x_i\}_{i \in I} \cup \{0\}$ .

*Proof.* We claim that

$$\text{cone}(\text{conv}(\{x_i\}_{i \in I} \cup \{0\})) = K. \quad (6.3)$$

Set  $C = \text{cone}(\text{conv}(\{x_i\}_{i \in I} \cup \{0\}))$  and let  $x \in C$ . Then there exist  $\lambda \in \mathbb{R}_{++}$  and a family  $(\alpha_i)_{i \in I}$  in  $\mathbb{R}_+$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $x = \lambda \sum_{i \in I} \alpha_i x_i$ . Hence  $x = \sum_{i \in I} (\lambda \alpha_i) x_i \in \sum_{i \in I} \mathbb{R}_+ x_i$  and thus  $C \subset K$ . Conversely, let  $x \in K$ . Then there exists a family  $(\alpha_i)_{i \in I}$  in  $\mathbb{R}_+$  such that  $x = \sum_{i \in I} \alpha_i x_i$ . If  $\alpha_i \equiv 0$ , then  $x = 0 \in C$ ; otherwise, set  $\lambda = \sum_{i \in I} \alpha_i$  and observe that  $x = \lambda \sum_{i \in I} (\alpha_i / \lambda) x_i \in C$ . Therefore,  $K \subset C$  and (6.3) follows. Using Proposition 6.2(iii), we deduce that  $K$  is the smallest convex cone containing  $\{x_i\}_{i \in I} \cup \{0\}$ .

In view of (6.3) and Proposition 6.2(iv), it remains to verify that  $K$  is closed. To do so, we consider two alternatives.

(a)  $\{x_i\}_{i \in I}$  is linearly independent: Set  $V = \text{span}\{x_i\}_{i \in I}$ . Then  $\{x_i\}_{i \in I}$  is a basis of  $V$  and  $K \subset V$ . Now let  $z \in \overline{K}$  and let  $(z_n)_{n \in \mathbb{N}}$  be a sequence in  $K$  such that  $z_n \rightarrow z$ . Then  $z \in V$  and hence there exists  $\{\alpha_i\}_{i \in I} \subset \mathbb{R}$  such that  $z = \sum_{i \in I} \alpha_i x_i$ . However, for every  $n \in \mathbb{N}$ , there exists  $\{\alpha_{n,i}\}_{i \in I} \subset \mathbb{R}_+$  such that  $z_n = \sum_{i \in I} \alpha_{n,i} x_i$ . Since  $(z_n)_{n \in \mathbb{N}}$  converges to  $z$  in  $V$  and  $\{x_i\}_{i \in I}$  is a basis of  $V$ , we have  $(\forall i \in I) 0 \leq \alpha_{n,i} \rightarrow \alpha_i$ . Thus  $\min_{i \in I} \alpha_i \geq 0$  and therefore  $z \in K$ .

(b)  $\{x_i\}_{i \in I}$  is linearly dependent: Then there exists  $\{\beta_i\}_{i \in I} \subset \mathbb{R}$  such that

$$\sum_{i \in I} \beta_i x_i = 0 \quad \text{and} \quad J = \{i \in I \mid \beta_i < 0\} \neq \emptyset. \quad (6.4)$$

Fix  $z \in K$ , say  $z = \sum_{i \in I} \alpha_i x_i$ , where  $\{\alpha_i\}_{i \in I} \subset \mathbb{R}_+$ , and set  $(\forall i \in I) \delta_i = \alpha_i - \gamma \beta_i$ , where  $\gamma = \max_{i \in J} \{\alpha_i / \beta_i\}$ . Then  $\gamma \leq 0$ ,  $\{\delta_i\}_{i \in I} \subset \mathbb{R}_+$ , and  $z = \sum_{i \in I} \delta_i x_i$ . Moreover, if  $j \in J$  satisfies  $\alpha_j / \beta_j = \gamma$ , then  $\delta_j = 0$  and therefore  $z = \sum_{i \in I \setminus \{j\}} \delta_i x_i$ . Thus, we obtain the decomposition

$$K = \bigcup_{j \in I} K_j, \quad \text{where} \quad (\forall j \in I) \quad K_j = \sum_{i \in I \setminus \{j\}} \mathbb{R}_+ x_i. \quad (6.5)$$

If the families  $(\{x_i\}_{i \in I \setminus \{j\}})_{j \in I}$  are linearly independent, it follows from (a) that the sets  $(K_j)_{j \in I}$  are closed and that  $K$  is therefore closed. Otherwise, for every  $j \in I$  for which  $\{x_i\}_{i \in I \setminus \{j\}}$  is linearly dependent, we reapply the decomposition procedure to  $K_j$  recursively until it can be expressed as a union of cones of the form  $\sum_{i \in I \setminus I'} \mathbb{R}_+ x_i$ , where  $\{x_i\}_{i \in I \setminus I'}$  is linearly independent. We thus obtain a decomposition of  $K$  as a finite union of closed sets.  $\square$

## 6.2 Generalized Interiors

The interior of a subset  $C$  of  $\mathcal{H}$  can be expressed as

$$\text{int } C = \{x \in C \mid (\exists \rho \in \mathbb{R}_{++}) \quad B(0; \rho) \subset C - x\}. \quad (6.6)$$

This formulation suggests several weaker notions of interiority.

**Definition 6.9** Let  $C$  be a convex subset of  $\mathcal{H}$ . The *core* of  $C$  is

$$\text{core } C = \{x \in C \mid \text{cone}(C - x) = \mathcal{H}\}; \quad (6.7)$$

the *strong relative interior* of  $C$  is

$$\text{sri } C = \{x \in C \mid \text{cone}(C - x) = \overline{\text{span}}(C - x)\}; \quad (6.8)$$

the *relative interior* of  $C$  is

$$\text{ri } C = \{x \in C \mid \text{cone}(C - x) = \text{span}(C - x)\}; \quad (6.9)$$

and the *quasirelative interior* of  $C$  is

$$\text{qri } C = \{x \in C \mid \overline{\text{cone}}(C - x) = \overline{\text{span}}(C - x)\}. \quad (6.10)$$

In addition, we use the notation  $\overline{\text{ri}} C = \overline{\text{ri}} C$  and  $\overline{\text{qri}} C = \overline{\text{qri}} C$ .

**Example 6.10** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$  such that  $C = -C$ . Then the following hold:

- (i)  $0 \in \text{ri } C$ .
- (ii) Suppose that  $\text{span } C$  is closed. Then  $0 \in \text{sri } C$ .

*Proof.* By Proposition 6.4(ii),  $\text{cone } C = \text{span } C$ . □

For every convex subset  $C$  of  $\mathcal{H}$ , since  $\text{cone } C \subset \text{span } C \subset \overline{\text{span}} C$ , we have

$$\text{int } C \subset \text{core } C \subset \text{sri } C \subset \text{ri } C \subset \text{qri } C \subset C. \quad (6.11)$$

As we now illustrate, each of the inclusions in (6.11) can be strict.

**Example 6.11** The following examples show that the reverse inclusions in (6.11) fail.

- (i) Example 8.33(iii) will provide a convex set  $C$  such that  $\text{int } C = \emptyset$  and  $0 \in \text{core } C$ . In contrast, Proposition 6.12 and Fact 6.13 provide common instances when  $\text{int } C = \text{core } C$ .
- (ii) Let  $C$  be a proper closed linear subspace of  $\mathcal{H}$ . Then  $\text{core } C = \emptyset$  and  $\text{sri } C = C$ .
- (iii) Suppose that  $\mathcal{H}$  is infinite-dimensional and separable, let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal basis of  $\mathcal{H}$ , and set

$$C = \left\{ \sum_{n \in \mathbb{N}} \xi_n e_n \mid (\forall n \in \mathbb{N}) \ |\xi_n| \leq \frac{1}{4^n} \right\}. \quad (6.12)$$

Then  $C$  is closed, convex, and  $C = -C$ . Hence, by Proposition 6.4(ii),  $\text{span } C = \text{cone } C$ . Since  $\{4^{-n} e_n\}_{n \in \mathbb{N}} \subset C$ , we see that  $\overline{\text{span}} C = \mathcal{H}$ . Now set  $x = \sum_{n \in \mathbb{N}} 2^{-n} e_n$ . Then  $x \in \overline{\text{span}} C$  and, if we had  $x \in \text{cone } C$ , then there would exist  $\beta \in \mathbb{R}_{++}$  such that  $(\forall n \in \mathbb{N}) \ 2^{-n} \leq \beta 4^{-n}$ , which is impossible. Hence  $x \notin \text{cone } C$  and thus

$$\text{cone } C = \text{span } C \neq \overline{\text{span}} C = \mathcal{H}. \quad (6.13)$$

Therefore,  $0 \in (\text{ri } C) \setminus (\text{sri } C)$ .

- (iv) Suppose that  $\mathcal{H}$  is infinite-dimensional and separable, let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal basis of  $\mathcal{H}$ , and set

$$C = \left\{ \sum_{n \in \mathbb{N}} \xi_n e_n \mid (\forall n \in \mathbb{N}) \quad -\frac{1}{2^n} \leq \xi_n \leq \frac{1}{4^n} \right\}. \quad (6.14)$$

Then  $C$  is closed and convex and, since  $\{e_n, -e_n\}_{n \in \mathbb{N}} \subset \text{cone } C$ , we have  $\overline{\text{cone } C} = \overline{\text{span } C} = \mathcal{H}$ . Moreover, arguing as in (iii), we note that  $x = -\sum_{n \in \mathbb{N}} 2^{-n} e_n \in C \subset \text{cone } C$ , while  $-x \notin \text{cone } C$ . Finally,  $\sum_{n \in \mathbb{N}} 2^{-n/2} e_n \in (\overline{\text{span } C}) \setminus (\text{span } C)$ . Altogether,

$$\text{cone } C \neq \text{span } C \neq \overline{\text{span } C} = \overline{\text{cone } C} = \mathcal{H} \quad (6.15)$$

and, therefore,  $0 \in (\text{qri } C) \setminus (\text{ri } C)$ .

- (v) Suppose that  $\mathcal{H} = \ell^2(\mathbb{R})$ , let  $(e_\rho)_{\rho \in \mathbb{R}}$  denote the standard unit vectors, and set  $C = \ell^2_+(\mathbb{R})$ , i.e.,

$$C = \left\{ \sum_{\rho \in \mathbb{R}} \xi_\rho e_\rho \in \ell^2(\mathbb{R}) \mid (\forall \rho \in \mathbb{R}) \quad \xi_\rho \geq 0 \right\}. \quad (6.16)$$

Then  $C$  is closed and convex. Fix  $x = \sum_{\rho \in \mathbb{R}} \xi_\rho e_\rho \in C$ . Since  $\{\rho \in \mathbb{R} \mid \xi_\rho \neq 0\}$  is countable, there exists  $\gamma \in \mathbb{R}$  such that  $\xi_\gamma = 0$ . Note that the  $\gamma$ -coordinate of every vector in  $\text{cone}(C - x)$  is positive and the same is true for  $\overline{\text{cone}}(C - x)$ . On the other hand, since  $x + e_\gamma \in C$ , we have  $e_\gamma \in (C - x)$  and therefore  $-e_\gamma \in \text{span}(C - x) \subset \overline{\text{span}}(C - x)$ . Altogether,  $-e_\gamma \in (\overline{\text{span}}(C - x)) \setminus (\overline{\text{cone}}(C - x))$  and consequently  $C \setminus (\text{qri } C) = C$ .

**Proposition 6.12** *Let  $C$  be a convex subset of  $\mathcal{H}$ , and suppose that one of the following holds:*

- (i)  $\text{int } C \neq \emptyset$ .
- (ii)  $C$  is closed.
- (iii)  $\mathcal{H}$  is finite-dimensional.

*Then  $\text{int } C = \text{core } C$ .*

*Proof.* Let  $x \in \text{core } C$ . It suffices to show that  $x \in \text{int } C$ . After subtracting  $x$  from  $C$  and replacing  $C$  by  $(-C) \cap C$ , we assume that  $x = 0$  and that  $C = -C$ . Thus, it is enough to assume that  $0 \in \text{core } C$  and to show that  $0 \in \text{int } C$ .

(i): Take  $y \in \text{int } C$ . Since  $C = -C$ ,  $-y \in \text{int } C$ , and Proposition 3.36(ii) yields  $0 \in [-y, y] \subset \text{int } C$ .

(ii): Since  $\bigcup_{n \in \mathbb{N}} nC = \mathcal{H}$ , Lemma 1.43(i) yields  $\text{int } C \neq \emptyset$ . Now apply (i).

(iii): Let  $(e_i)_{i \in I}$  be an orthonormal basis of  $\mathcal{H}$ . There exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $D = \text{conv}\{-\varepsilon e_i, +\varepsilon e_i\}_{i \in I} \subset C$ . Since  $B(0; \varepsilon/\sqrt{\dim \mathcal{H}}) \subset D$ , the proof is complete.  $\square$

The following results refine Proposition 6.12(ii) and provide further information on generalized interiors.

**Fact 6.13** *Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$ . Then  $\text{int}(C - D) = \text{core}(C - D)$ .*

*Proof.* This is a consequence of [233, Corollary 13.2].  $\square$

**Fact 6.14** *Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ .*

- (i) *Suppose that  $\mathcal{H}$  is finite-dimensional. Then  $\text{ri } C$  is the interior of  $C$  relative to  $\text{aff } C$  and  $\text{ri } C \neq \emptyset$ . Moreover,  $\overline{\text{ri } C} = \overline{C}$ ,  $\text{ri } \overline{C} = \text{ri } C$ , and  $\text{sri } C = \text{ri } C = \text{qri } C$ .*
- (ii) *Suppose that  $\mathcal{H}$  is separable and that  $C$  is closed. Then  $C = \overline{\text{qri } C}$  and, in particular,  $\text{qri } C \neq \emptyset$ .*
- (iii) *Suppose that  $\text{int } C \neq \emptyset$ . Then  $\text{int } C = \text{core } C = \text{sri } C = \text{ri } C = \text{qri } C$ .*
- (iv) *Let  $\mathcal{K}$  be a finite-dimensional real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and suppose that  $\text{qri } C \neq \emptyset$ . Then  $\text{ri } L(C) = L(\text{qri } C)$ .*
- (v) *Suppose that  $\mathcal{H}$  is finite-dimensional and let  $D$  be a convex subset of  $\mathcal{H}$  such that  $(\text{ri } C) \cap (\text{ri } D) \neq \emptyset$ . Then  $\text{ri}(C \cap D) = (\text{ri } C) \cap (\text{ri } D)$ .*

*Proof.* (i): It follows from (6.7) and (6.9) that  $\text{ri } C$  is the core relative to  $\text{aff } C$ . However, as seen in Proposition 6.12(iii), since  $\mathcal{H}$  is finite-dimensional, the notions of core and interior coincide. Furthermore,  $\text{ri } C \neq \emptyset$  by [219, Theorem 6.2]. Next, since  $\text{ri } C$  is nonempty and coincides with the interior of  $C$  relative to  $\text{aff } C$ , the identities  $\overline{\text{ri } C} = \overline{C}$  and  $\text{ri } \overline{C} = \text{ri } C$  follow from Proposition 3.36(iii). We also observe that  $\text{sri } C = \text{ri } C$  since finite-dimensional linear subspaces are closed. Now assume that  $x \in \text{qri } C$ . Then  $\overline{\text{cone}}(C - x) = \overline{\text{span}}(C - x) = \text{span}(C - x)$ , and hence  $\text{ri}(\text{cone}(C - x)) = \text{ri}(\overline{\text{cone}}(C - x)) = \text{ri}(\text{span}(C - x)) = \text{span}(C - x)$ . It follows that  $\text{cone}(C - x) = \text{span}(C - x)$ , i.e.,  $x \in \text{ri } C$ . Altogether,  $\text{sri } C = \text{ri } C = \text{qri } C$ .

(ii): See [44, Proposition 2.12 and Theorem 2.19] or [263, Lemma 2.7].

(iii): [44, Corollary 2.14] implies that  $\text{qri } C = \text{int } C$  (see also [263, Lemma 2.8] when  $C$  is closed). The identities thus follow from (6.11).

(iv): See [44, Proposition 2.10].

(v): [219, Theorem 6.5].  $\square$

**Corollary 6.15** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $\mathcal{K}$  be a finite-dimensional real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $C$  and  $D$  be nonempty convex subsets of  $\mathcal{H}$ . Then the following hold:*

- (i)  $\text{ri } L(C) = L(\text{ri } C)$ .
- (ii)  $\text{ri}(C - D) = (\text{ri } C) - (\text{ri } D)$ .

*Proof.* (i): It follows from Fact 6.14(i) that  $\text{ri } C = \text{qri } C \neq \emptyset$ . Hence, Fact 6.14(iv) yields  $\text{ri } L(C) = L(\text{ri } C)$ .

(ii): Set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, y) \mapsto x - y$ . It follows from (i) that  $\text{ri}(C - D) = \text{ri } L(C \times D) = L(\text{ri}(C \times D)) = L((\text{ri } C) \times (\text{ri } D)) = (\text{ri } C) - (\text{ri } D)$ .  $\square$

**Proposition 6.16** *Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $\text{int } C \neq \emptyset$  and  $0 \in C$ . Then  $\text{int}(\text{cone } C) = \text{cone}(\text{int } C)$ .*

*Proof.* It is clear that  $\text{cone}(\text{int } C) \subset \text{cone } C$  and that  $\text{cone}(\text{int } C)$  is open, since it is a union of open sets. Hence

$$\text{cone}(\text{int } C) \subset \text{int}(\text{cone } C). \quad (6.17)$$

To prove the reverse inclusion, take  $x \in \text{int}(\text{cone } C)$ . We must show that

$$x \in \text{cone}(\text{int } C). \quad (6.18)$$

Since  $x \in \text{int}(\text{cone } C)$ , there exist  $\varepsilon_0 \in \mathbb{R}_{++}$ ,  $\gamma \in \mathbb{R}_{++}$ , and  $x_1 \in C$  such that  $B(x; \varepsilon_0) \subset \text{cone } C$  and  $x = \gamma x_1 \in \gamma C$ . If  $x_1 \in \text{int } C$ , then (6.18) holds. We therefore assume that  $x_1 \in C \setminus (\text{int } C)$ . Fix  $y_1 \in \text{int } C$  and set  $y = \gamma y_1 \in \gamma C$ . Since  $x_1 \neq y_1$ , we have  $x \neq y$ . Now set  $\varepsilon = \varepsilon_0 / \|x - y\|$ . Then  $x + \varepsilon(x - y) \in B(x; \varepsilon_0) \subset \text{cone } C$ , and hence there exists  $\rho \in \mathbb{R}_{++}$  such that  $x + \varepsilon(x - y) \in \rho C$ . Set  $\mu = \max\{\gamma, \rho\} > 0$ . Because  $C$  is convex and  $0 \in C$ , we have  $(\gamma C) \cup (\rho C) = \mu C$ . On the other hand, the inclusions  $x \in \gamma C \subset \mu C$ ,  $y \in \gamma C \subset \mu C$ , and  $x + \varepsilon(x - y) \in \rho C \subset \mu C$  yield

$$x/\mu \in C, \quad y/\mu \in C, \quad \text{and} \quad x/\mu + \varepsilon(x/\mu - y/\mu) \in C. \quad (6.19)$$

We claim that

$$y/\mu \in \text{int } C. \quad (6.20)$$

If  $y_1 = 0$ , then  $y = \gamma y_1 = 0$  and thus  $y/\mu = 0 = y_1 \in \text{int } C$ ; otherwise,  $y_1 \neq 0$  and Proposition 3.35 yields  $y/\mu = (\gamma/\mu)y_1 \in ]0, y_1] \subset \text{int } C$ . Hence (6.20) holds. Now set  $\lambda = 1/(1 + \varepsilon) \in ]0, 1[$ . Then  $\lambda\varepsilon = 1 - \lambda$  and, since  $x/\mu + \varepsilon(x/\mu - y/\mu) \in C$  by (6.19) and  $y/\mu \in \text{int } C$  by (6.20), it follows from Proposition 3.35 that  $x/\mu = \lambda(x/\mu + \varepsilon(x/\mu - y/\mu)) + (1 - \lambda)(y/\mu) \in \text{int } C$ . Thus  $x \in \mu \text{int } C$  and (6.18) follows.  $\square$

**Proposition 6.17** *Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $\text{int } C \neq \emptyset$  and  $0 \in C$ . Then the following are equivalent:*

- (i)  $0 \in \text{int } C$ .
- (ii)  $\text{cone}(\text{int } C) = \mathcal{H}$ .
- (iii)  $\text{cone } C = \mathcal{H}$ .
- (iv)  $\overline{\text{cone } C} = \mathcal{H}$ .

*Proof.* (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (iv): Clear.

(iv) $\Rightarrow$ (ii): Since Proposition 6.2(iii) asserts that  $\text{cone } C$  is convex, Proposition 3.36(iii) and Proposition 6.16 imply that

$$\mathcal{H} = \text{int } \mathcal{H} = \text{int}(\overline{\text{cone } C}) = \text{int}(\text{cone } C) = \text{cone}(\text{int } C). \quad (6.21)$$

(ii) $\Rightarrow$ (i): We have  $0 \in \text{cone}(\text{int } C)$  and thus  $0 \in \lambda \text{int } C$ , for some  $\lambda \in \mathbb{R}_{++}$ . We conclude that  $0 \in \text{int } C$ .  $\square$



The next example illustrates the fact that items (i) and (iv) in Proposition 6.17 are no longer equivalent when the assumption on the interior is dropped.

**Example 6.18** Suppose that  $\mathcal{H}$  is infinite-dimensional, let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal sequence in  $\mathcal{H}$ , and set  $S = \overline{\text{conv}} \{\pm 2^{-n} e_n\}_{n \in \mathbb{N}}$  and  $C = S + S^\perp$ . Then  $C$  is closed and convex,  $0 \in C$ ,  $\text{int } C = \emptyset$ , and  $\overline{\text{cone}} C = \mathcal{H}$ .

*Proof.* Since  $\text{span}\{\pm e_n\}_{n \in \mathbb{N}} + S^\perp \subset \text{cone } C$ , we have  $\overline{\text{cone}} C = \mathcal{H}$ . Furthermore,  $0 \in [-e_0, e_0] \subset S \subset C$ . Now assume that  $0 \in \text{int } C$  and let  $m \in \mathbb{N}$ . Then there exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $B(0; \varepsilon) \subset C$  and  $\varepsilon e_m \in \overline{\text{conv}} \{\pm 2^{-n} e_n\}_{n \in \mathbb{N}}$ . Hence  $\varepsilon = \langle \varepsilon e_m \mid e_m \rangle \in \langle \overline{\text{conv}} \{\pm 2^{-n} e_n\}_{n \in \mathbb{N}} \mid e_m \rangle$ . Since  $\langle \cdot \mid e_m \rangle$  is continuous and linear, it follows that

$$\varepsilon \in \overline{\text{conv}} \langle \{\pm 2^{-n} e_n\}_{n \in \mathbb{N}} \mid e_m \rangle = [-2^{-m}, 2^{-m}]. \quad (6.22)$$

Thus,  $\varepsilon \leq 2^{-m}$ , which is impossible since  $m$  is arbitrary. Therefore,  $0 \notin \text{int } C$  and hence  $\text{int } C = \emptyset$  by Proposition 6.17.  $\square$

The property that the origin lies in the strong relative interior of composite sets will be central in several places in this book. The next proposition provides sufficient conditions under which it is satisfied.

**Proposition 6.19** *Let  $C$  be a convex subset of  $\mathcal{H}$ , let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $D$  be a convex subset of  $\mathcal{K}$ . Suppose that one of the following holds:*

- (i)  $D - L(C)$  is a closed linear subspace.
- (ii)  $C$  and  $D$  are linear subspaces and one of the following holds:
  - (a)  $D + L(C)$  is closed.
  - (b)  $D$  is closed, and  $L(C)$  is finite-dimensional or finite-codimensional.
  - (c)  $D$  is finite-dimensional or finite-codimensional, and  $L(C)$  is closed.
- (iii)  $D$  is a cone and  $D - \text{cone } L(C)$  is a closed linear subspace.
- (iv)  $D = L(C)$  and  $\text{span } D$  is closed.
- (v)  $0 \in \text{core}(D - L(C))$ .
- (vi)  $0 \in \text{int}(D - L(C))$ .
- (vii)  $D \cap \text{int } L(C) \neq \emptyset$  or  $L(C) \cap \text{int } D \neq \emptyset$ .
- (viii)  $\mathcal{K}$  is finite-dimensional and  $(\text{ri } D) \cap (\text{ri } L(C)) \neq \emptyset$ .
- (ix)  $\mathcal{K}$  is finite-dimensional and  $(\text{ri } D) \cap L(\text{qri } C) \neq \emptyset$ .
- (x)  $\mathcal{H}$  and  $\mathcal{K}$  are finite-dimensional and  $(\text{ri } D) \cap L(\text{ri } C) \neq \emptyset$ .

Then  $0 \in \text{sri}(D - L(C))$ .

*Proof.* (i): We have  $D - L(C) \subset \text{cone}(D - L(C)) \subset \text{span}(D - L(C)) \subset \overline{\text{span}}(D - L(C))$ . Hence, since the assumption implies that  $D - L(C) = \overline{\text{span}}(D - L(C))$ , we obtain  $\text{cone}(D - L(C)) = \overline{\text{span}}(D - L(C))$ , and (6.8) yields  $0 \in \text{sri}(D - L(C))$ .

- (ii)(a) $\Rightarrow$ (i): Since  $D$  and  $C$  are linear subspaces, so is  $D - L(C) = D + L(C)$ .  
(ii)(b) $\Rightarrow$ (ii)(a) and (ii)(c) $\Rightarrow$ (ii)(a): Fact 2.21.  
(iii): Since  $\overline{\text{span}}(D - L(C)) \subset \overline{\text{span}}(D - \text{cone } L(C)) = D - \text{cone } L(C) = \text{cone}(D - L(C)) \subset \overline{\text{span}}(D - L(C))$ , we have  $\text{cone}(D - L(C)) = \overline{\text{span}}(D - L(C))$ , and (6.8) yields  $0 \in \text{sri}(D - L(C))$ .  
(iv): Since  $D - L(C) = D - D = -(D - D) = -(D - L(C))$ , Proposition 6.4(ii) yields  $\text{cone}(D - L(C)) = \text{cone}(D - D) = \text{span}(D - D) = \text{span } D = \overline{\text{span}} D = \overline{\text{span}}(D - D) = \overline{\text{span}}(D - L(C))$ .  
(v)&(vi): See (6.11).  
(vii) $\Rightarrow$ (vi): Suppose that  $y \in D \cap \text{int } L(C)$ , say  $B(y; \rho) \subset L(C)$  for some  $\rho \in \mathbb{R}_{++}$ . Then  $B(0; \rho) = y - B(y; \rho) \subset D - L(C)$  and therefore  $0 \in \text{int}(D - L(C))$ . The second condition is handled analogously.  
(viii): By Fact 6.14(i),  $\text{ri}(D - L(C)) = \text{sri}(D - L(C))$ . On the other hand, we derive from Corollary 6.15(ii) that  $(\text{ri } D) \cap (\text{ri } L(C)) \neq \emptyset \Leftrightarrow 0 \in (\text{ri } D) - (\text{ri } L(C)) = \text{ri}(D - L(C))$ .  
(ix) $\Rightarrow$ (viii): Fact 6.14(iv).  
(x) $\Rightarrow$ (ix): Fact 6.14(i). □

**Proposition 6.20** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , and let  $(C_i)_{i \in I}$  be convex subsets of  $\mathcal{H}$  such that one of the following holds:*

- (i) *For every  $i \in \{2, \dots, m\}$ ,  $C_i - \bigcap_{j=1}^{i-1} C_j$  is a closed linear subspace.*
- (ii) *The sets  $(C_i)_{i \in I}$  are linear subspaces and, for every  $i \in \{2, \dots, m\}$ ,  $C_i + \bigcap_{j=1}^{i-1} C_j$  is closed.*
- (iii)  *$C_m \cap \bigcap_{i=1}^{m-1} \text{int } C_i \neq \emptyset$ .*
- (iv)  *$\mathcal{H}$  is finite-dimensional and  $\bigcap_{i \in I} \text{ri } C_i \neq \emptyset$ .*

Then

$$0 \in \bigcap_{i=2}^m \text{sri} \left( C_i - \bigcap_{j=1}^{i-1} C_j \right). \quad (6.23)$$

*Proof.* We apply several items from Proposition 6.19 with  $\mathcal{K} = \mathcal{H}$  and  $L = \text{Id}$ .

- (i) $\Rightarrow$ (6.23): Proposition 6.19(i).
- (ii) $\Rightarrow$ (6.23): Proposition 6.19(ii)(a).
- (iii) $\Rightarrow$ (6.23): Proposition 6.19(vii).
- (iv) $\Rightarrow$ (6.23): Proposition 6.19(viii) and Fact 6.14(v). □

## 6.3 Polar and Dual Cones

**Definition 6.21** Let  $C$  be a subset of  $\mathcal{H}$ . The *polar cone* of  $C$  is

$$C^\ominus = \{u \in \mathcal{H} \mid \sup \langle C \mid u \rangle \leq 0\}, \quad (6.24)$$

and the *dual cone* of  $C$  is  $C^\oplus = -C^\ominus$ . If  $C$  is a nonempty convex cone, then  $C$  is *self-dual* if  $C = C^\oplus$ .

The next two results are immediate consequences of Definition 6.21 and (2.2).

**Proposition 6.22** *Let  $C$  be a linear subspace of  $\mathcal{H}$ . Then  $C^\ominus = C^\perp$ .*

**Proposition 6.23** *Let  $C$  be a subset of  $\mathcal{H}$ . Then the following hold:*

- (i) *Let  $D \subset C$ . Then  $C^\ominus \subset D^\ominus$  and  $C^\oplus \subset D^\oplus$ .*
- (ii)  *$C^\ominus$  and  $C^\oplus$  are nonempty closed convex cones.*
- (iii)  *$C^\ominus = (\text{cone } C)^\ominus = (\text{conv } C)^\ominus = \overline{C}^\ominus$ .*
- (iv)  *$C^\ominus \cap C^\oplus = C^\perp$ .*
- (v) *Suppose that  $\overline{\text{cone}} C = -\overline{\text{cone}} C$ . Then  $C^\ominus = C^\oplus = C^\perp$ .*

**Example 6.24** Let  $I$  be a totally ordered set. Then  $\ell_+^2(I)$  is self-dual. In particular, the positive orthant  $\mathbb{R}_+^N$  in  $\mathbb{R}^N$  is self-dual.

*Proof.* Let  $(e_i)_{i \in I}$  be the standard unit vectors of  $\ell_+^2(I)$ . By Example 6.7,  $\ell_+^2(I) = \overline{\text{cone}} \text{conv}\{e_i\}_{i \in I}$ . Hence, using Proposition 6.23(iii),  $(\ell_+^2(I))^\oplus = \{e_i\}_{i \in I}^\oplus = \{x \in \ell^2(I) \mid (\forall i \in I) \langle x \mid e_i \rangle \geq 0\} = \ell_+^2(I)$ .  $\square$

**Example 6.25 (Fejér)** The convex cone of positive semidefinite symmetric matrices in  $\mathbb{S}^N$  is self-dual.

*Proof.* See, e.g., [145, Corollary 7.5.4].  $\square$

**Proposition 6.26** *Let  $K_1$  and  $K_2$  be nonempty cones in  $\mathcal{H}$ . Then*

$$(K_1 + K_2)^\ominus = K_1^\ominus \cap K_2^\ominus. \quad (6.25)$$

*Consequently, if  $K_1$  and  $K_2$  are linear subspaces, then  $(K_1 + K_2)^\perp = K_1^\perp \cap K_2^\perp$ .*

*Proof.* Fix  $x_1 \in K_1$  and  $x_2 \in K_2$ . First, let  $u \in (K_1 + K_2)^\ominus$ . Then, for every  $\lambda_1 \in \mathbb{R}_{++}$  and  $\lambda_2 \in \mathbb{R}_{++}$ ,  $\lambda_1 x_1 + \lambda_2 x_2 \in \lambda_1 K_1 + \lambda_2 K_2 = K_1 + K_2$  and therefore  $\langle \lambda_1 x_1 + \lambda_2 x_2 \mid u \rangle \leq 0$ . Setting  $\lambda_1 = 1$  and letting  $\lambda_2 \downarrow 0$  yields  $u \in K_1^\ominus$ . Likewise, setting  $\lambda_2 = 1$  and letting  $\lambda_1 \downarrow 0$  yields  $u \in K_2^\ominus$ . Thus,  $(K_1 + K_2)^\ominus \subset K_1^\ominus \cap K_2^\ominus$ . Conversely, let  $u \in K_1^\ominus \cap K_2^\ominus$ . Then  $\langle x_1 \mid u \rangle \leq 0$  and  $\langle x_2 \mid u \rangle \leq 0$ ; hence  $\langle x_1 + x_2 \mid u \rangle \leq 0$ . Thus  $u \in (K_1 + K_2)^\ominus$  and therefore  $K_1^\ominus \cap K_2^\ominus \subset (K_1 + K_2)^\ominus$ . Finally, the assertion concerning linear subspaces follows from Proposition 6.22.  $\square$

As just illustrated, the relationship between a cone and its polar cone is, in many respects, similar to that between a linear subspace and its orthogonal complement. The next three results partially generalize Corollary 3.22.

**Proposition 6.27** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ , let  $x \in \mathcal{H}$ , and let  $p \in \mathcal{H}$ . Then  $p = P_K x \Leftrightarrow [p \in K, x - p \perp p, \text{ and } x - p \in K^\ominus]$ .*

*Proof.* Theorem 3.14 asserts that  $p = P_K x$  if and only if

$$p \in K \quad \text{and} \quad (\forall y \in K) \quad \langle y - p \mid x - p \rangle \leq 0. \quad (6.26)$$

Suppose that (6.26) holds. Then  $0 \in K$ ,  $2p \in K$ , and therefore  $\langle -p \mid x - p \rangle \leq 0$  and  $\langle 2p - p \mid x - p \rangle \leq 0$ . Hence,  $\langle p \mid x - p \rangle = 0$ . In turn,  $(\forall y \in K)$   $\langle y \mid x - p \rangle = \langle y - p \mid x - p \rangle + \langle p \mid x - p \rangle \leq 0$ . Thus,  $x - p \in K^\ominus$ . Conversely,  $(\forall y \in K)$   $[\langle p \mid x - p \rangle = 0 \text{ and } \langle y \mid x - p \rangle \leq 0] \Rightarrow \langle y - p \mid x - p \rangle \leq 0$ .  $\square$

**Example 6.28** Let  $I$  be a totally ordered set, suppose that  $\mathcal{H} = \ell^2(I)$ , set  $K = \ell^2_+(I)$ , and let  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ . Then  $P_K x = (\max\{\xi_i, 0\})_{i \in I}$ .

*Proof.* This is a direct application of Proposition 6.27 where, by Example 6.24,  $K^\ominus = -K^\oplus = -K$ .  $\square$

**Theorem 6.29 (Moreau)** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then the following hold:*

- (i)  $x = P_K x + P_{K^\ominus} x$ .
- (ii)  $P_K x \perp P_{K^\ominus} x$ .
- (iii)  $\|x\|^2 = d_K^2(x) + d_{K^\ominus}^2(x)$ .

*Proof.* (i): Set  $q = x - P_K x$ . By Proposition 6.27,  $q \in K^\ominus$ ,  $x - q = P_K x \perp x - P_K x = q$ , and  $x - q = P_K x \in K \subset K^{\ominus\ominus}$ . Appealing once more to Proposition 6.27, we conclude that  $q = P_{K^\ominus} x$ .

(ii): It follows from Proposition 6.27 and (i) that  $P_K x \perp x - P_K x = P_{K^\ominus} x$ .

(iii): Using (i) and (ii), we obtain  $\|x\|^2 = \|P_{K^\ominus} x + P_K x\|^2 = \|P_{K^\ominus} x\|^2 + \|P_K x\|^2 = \|x - P_K x\|^2 + \|x - P_{K^\ominus} x\|^2 = d_K^2(x) + d_{K^\ominus}^2(x)$ .  $\square$

There is only one way to split a vector into the sum of a vector in a closed linear subspace  $V$  and a vector in  $V^\ominus = V^\perp$ . For general convex cones, this is no longer true (for instance, in  $\mathcal{H} = \mathbb{R}$ ,  $0 = x - x$  for every  $x \in K = \mathbb{R}_+ = -\mathbb{R}_-$ ). However, the decomposition provided by Theorem 6.29(i) is unique in several respects.

**Corollary 6.30** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Suppose that  $y \in K \setminus \{P_K x\}$  and  $z \in K^\ominus \setminus \{P_{K^\ominus} x\}$  satisfy  $x = y + z$ . Then the following hold:*

- (i)  $\|P_K x\| < \|y\|$  and  $\|P_{K^\ominus} x\| < \|z\|$ .
- (ii)  $\langle y \mid z \rangle < \langle P_K x \mid P_{K^\ominus} x \rangle = 0$ .
- (iii)  $\|P_K x - P_{K^\ominus} x\| < \|y - z\|$ .

*Proof.* (i): We deduce from Theorem 3.14 that  $\|P_K x\| = \|x - P_{K^\ominus} x\| < \|x - z\| = \|y\|$  and that  $\|P_{K^\ominus} x\| = \|x - P_K x\| < \|x - y\| = \|z\|$ .

(ii): Since  $y \in K$  and  $z \in K^\ominus$ , we have  $\langle y \mid z \rangle \leq 0$ . However, if  $\langle y \mid z \rangle = 0$ , then (i) and Theorem 6.29 yield  $\|x\|^2 = \|y + z\|^2 = \|y\|^2 + \|z\|^2 > \|P_K x\|^2 + \|P_{K^\ominus} x\|^2 = \|x\|^2$ , which is impossible.

(iii): By (i) and (ii),  $\|y - z\|^2 = \|y\|^2 - 2\langle y | z \rangle + \|z\|^2 > \|P_K x\|^2 - 2\langle P_K x | P_{K^\ominus} x \rangle + \|P_{K^\ominus} x\|^2 = \|P_K x - P_{K^\ominus} x\|^2$ .  $\square$

The following fact will be used repeatedly.

**Proposition 6.31** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Suppose that  $\langle x | x - P_K x \rangle \leq 0$ . Then  $x \in K$ .*

*Proof.* By Proposition 6.27,  $\langle P_K x | x - P_K x \rangle = 0$ . Therefore,  $\|x - P_K x\|^2 = \langle x | x - P_K x \rangle - \langle P_K x | x - P_K x \rangle \leq 0$ . Thus,  $x = P_K x \in K$ .  $\square$

**Proposition 6.32** *Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . Then*

$$C^{\ominus\ominus} = \overline{\text{cone } C}. \quad (6.27)$$

*Proof.* Set  $K = \overline{\text{cone } C}$ . Since  $C \subset C^{\ominus\ominus}$ , Proposition 6.23(ii) yields  $K \subset \overline{\text{cone } C^{\ominus\ominus}} = C^{\ominus\ominus}$ . Conversely, let  $x \in C^{\ominus\ominus}$ . By Proposition 6.23(iii) and Proposition 6.27,  $x \in C^{\ominus\ominus} = K^{\ominus\ominus}$  and  $x - P_K x \in K^\ominus$ . Therefore  $\langle x | x - P_K x \rangle \leq 0$  and, by Proposition 6.31,  $x \in K$ . Thus  $C^{\ominus\ominus} \subset K$ .  $\square$

**Corollary 6.33** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ . Then*

$$K^{\ominus\ominus} = K. \quad (6.28)$$

**Proposition 6.34** *Let  $K_1$  and  $K_2$  be two nonempty convex cones in  $\mathcal{H}$ . Then*

$$(\overline{K_1} \cap \overline{K_2})^\ominus = \overline{K_1^\ominus + K_2^\ominus}. \quad (6.29)$$

*Proof.* By Proposition 6.26 and Proposition 6.32,  $(K_1^\ominus + K_2^\ominus)^\ominus = \overline{K_1} \cap \overline{K_2}$ . Taking polars yields the result.  $\square$

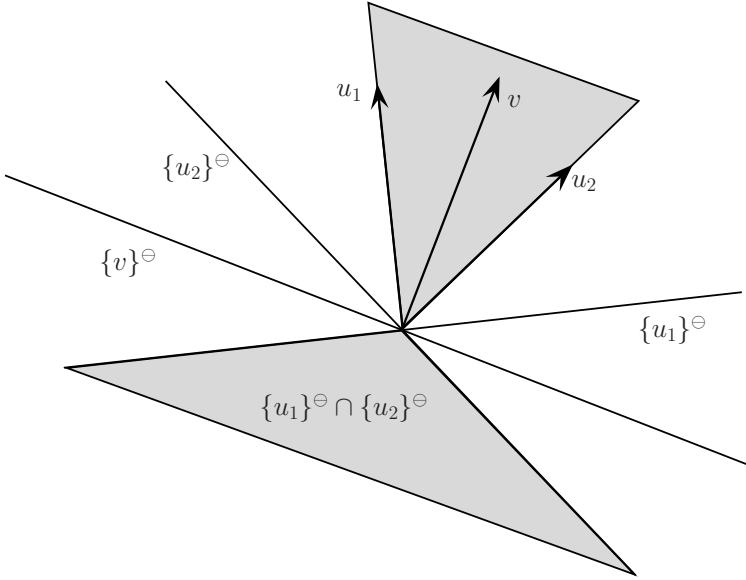
**Theorem 6.35 (Farkas)** *Let  $v \in \mathcal{H}$  and let  $(u_i)_{i \in I}$  be a finite family in  $\mathcal{H}$ . Then  $v \in \sum_{i \in I} \mathbb{R}_+ u_i \Leftrightarrow \bigcap_{i \in I} \{u_i\}^\ominus \subset \{v\}^\ominus$ .*

*Proof.* Set  $K = \sum_{i \in I} \mathbb{R}_+ u_i$ . If  $v \in K$ , then clearly  $\bigcap_{i \in I} \{u_i\}^\ominus \subset \{v\}^\ominus$ . Conversely, assume that  $\bigcap_{i \in I} \{u_i\}^\ominus \subset \{v\}^\ominus$ . Proposition 6.8 and Proposition 6.27 yield  $v - P_K v \in K^\ominus \subset \bigcap_{i \in I} \{u_i\}^\ominus \subset \{v\}^\ominus$ , hence  $\langle v | v - P_K v \rangle \leq 0$ . In turn, Proposition 6.31 yields  $v \in K$ .  $\square$

The next result is a powerful generalization of Fact 2.18(iii)&(iv), which correspond to the case  $K = \{0\}$ .

**Proposition 6.36** *Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $K$  be a nonempty closed convex cone in  $\mathcal{K}$ . Then the following hold:*

- (i)  $(L^{-1}(K))^\ominus = \overline{L^*(K^\ominus)}$ .
- (ii)  $(L^*)^{-1}(K) = (L(K^\ominus))^\ominus$ .



**Fig. 6.1** Farkas's lemma:  $v$  lies in the cone generated by  $u_1$  and  $u_2$  if and only if the half-space  $\{v\}^\ominus$  contains the cone  $\{u_1\}^\ominus \cap \{u_2\}^\ominus$ .

*Proof.* (i): Set  $C = L^{-1}(K)$  and  $D = \overline{L^*(K^\ominus)}$ , and let  $v \in K^\ominus$ . Then  $\sup \langle C \mid L^*v \rangle = \sup \langle L(C) \mid v \rangle \leq \sup \langle K \mid v \rangle = 0$  and hence  $L^*v \in C^\ominus$ . Therefore  $L^*(K^\ominus) \subset C^\ominus$  and thus  $D \subset C^\ominus$ . Conversely, take  $u \in C^\ominus$  and set  $p = P_D u$ , which is well defined since  $D$  is a nonempty closed convex cone. By Proposition 6.27,  $u - p \in D^\ominus$ . Therefore  $\sup \langle K^\ominus \mid L(u - p) \rangle = \sup \langle L^*(K^\ominus) \mid u - p \rangle \leq 0$  and, using Corollary 6.33, we obtain  $L(u - p) \in K^{\ominus\ominus} = K$ . Thus,  $u - p \in C$  and, in turn,  $\langle u \mid u - p \rangle \leq 0$ . Thus, Proposition 6.31 yields  $u \in D$ .

(ii): Since  $L^* \in \mathcal{B}(\mathcal{K}, \mathcal{H})$ ,  $(L^*)^{-1}(K)$  is a nonempty closed convex cone. Hence, Corollary 6.33, (i), and Proposition 6.23(iii) yield  $(L^*)^{-1}(K) = ((L^*)^{-1}(K))^{\ominus\ominus} = \overline{L^{**}(K^\ominus)}^\ominus = (L(K^\ominus))^\ominus$ .  $\square$

## 6.4 Tangent and Normal Cones

**Definition 6.37** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . The *tangent cone* to  $C$  at  $x$  is

$$T_C x = \begin{cases} \overline{\text{cone}}(C - x) = \overline{\bigcup_{\lambda \in \mathbb{R}_{++}} \lambda(C - x)}, & \text{if } x \in C; \\ \emptyset, & \text{otherwise,} \end{cases} \quad (6.30)$$

and the *normal cone* to  $C$  at  $x$  is

$$N_C x = \begin{cases} (C - x)^\ominus = \{u \in \mathcal{H} \mid \sup \langle C - x \mid u \rangle \leq 0\}, & \text{if } x \in C; \\ \emptyset, & \text{otherwise.} \end{cases} \quad (6.31)$$

**Example 6.38** Let  $C = B(0; 1)$  and let  $x \in C$ . Then

$$T_C x = \begin{cases} \{y \in \mathcal{H} \mid \langle y \mid x \rangle \leq 0\}, & \text{if } \|x\| = 1; \\ \mathcal{H}, & \text{if } \|x\| < 1, \end{cases} \quad (6.32)$$

and

$$N_C x = \begin{cases} \mathbb{R}_+ x, & \text{if } \|x\| = 1; \\ \{0\}, & \text{if } \|x\| < 1. \end{cases} \quad (6.33)$$

**Example 6.39** Let  $K$  be a nonempty convex cone in  $\mathcal{H}$  and let  $x \in K$ . Then

$$T_K x = \overline{K + \mathbb{R}x} \quad \text{and} \quad N_K x = K^\ominus \cap \{x\}^\perp. \quad (6.34)$$

*Proof.* Proposition 6.3(i) yields  $K + \mathbb{R}_+ x = K$ , which implies that  $K + \mathbb{R}x = K - \mathbb{R}_{++}x = \bigcup_{\lambda \in \mathbb{R}_{++}} K - \lambda x = \bigcup_{\lambda \in \mathbb{R}_{++}} \lambda(K - x)$ . Taking closures, we obtain  $T_K x = \overline{K + \mathbb{R}x}$ . Let us now show that  $N_K x = K^\ominus \cap \{x\}^\perp$ . If  $u \in K^\ominus \cap \{x\}^\perp$ , then  $(\forall y \in K) \langle y - x \mid u \rangle = \langle y \mid u \rangle \leq 0$  and thus  $u \in N_K x$ . Conversely, take  $u \in N_K x$ . Then  $(\forall y \in K) \langle y - x \mid u \rangle \leq 0$ . Since  $\{x/2, 2x\} \subset K$ , it follows that  $\langle x \mid u \rangle = 0$  and hence that  $u \in K^\ominus$ .  $\square$

**Example 6.40** Let  $I$  be a totally ordered set, suppose that  $\mathcal{H} = \ell^2(I)$ , set  $K = \ell_+^2(I)$ , and let  $x = (\xi_i)_{i \in I} \in K$ . Then

$$N_K x = \{(\nu_i)_{i \in I} \in \mathcal{H} \mid (\forall i \in I) \nu_i \leq 0 = \xi_i \nu_i\}. \quad (6.35)$$

*Proof.* Example 6.24 and Example 6.39 imply that  $N_K x = \ell_-^2(I) \cap \{x\}^\perp$ . Let  $u = (\nu_i)_{i \in I} \in \ell_-^2(I)$ . Since for every  $i \in I$ ,  $\nu_i \leq 0 \leq \xi_i$ , we see that  $\sum_{i \in I} \xi_i \nu_i = \langle x \mid u \rangle = 0$  if and only if  $(\forall i \in I) \xi_i \nu_i = 0$ .  $\square$

**Example 6.41** Let  $N$  be a strictly positive integer, set  $I = \{1, \dots, N\}$ , let  $x = (\xi_i)_{i \in I} \in \mathbb{R}^N$ , and let  $y = (\eta_i)_{i \in I} \in \mathbb{R}^N$ . Then the following hold:

- (i) Suppose that  $x \in \mathbb{R}_+^N$ . Then  $y \in N_{\mathbb{R}_+^N} x \Leftrightarrow (\forall i \in I) \begin{cases} \eta_i \leq 0, & \text{if } \xi_i = 0; \\ \eta_i = 0, & \text{if } \xi_i > 0. \end{cases}$
- (ii) Suppose that  $x \in \mathbb{R}_-^N$ . Then  $y \in N_{\mathbb{R}_-^N} x \Leftrightarrow (\forall i \in I) \begin{cases} \eta_i \geq 0, & \text{if } \xi_i = 0; \\ \eta_i = 0, & \text{if } \xi_i < 0. \end{cases}$

**Example 6.42** Let  $C$  be an affine subspace of  $\mathcal{H}$ , let  $V = C - C$  be its parallel linear subspace, and let  $x \in \mathcal{H}$ . Then

$$T_C x = \begin{cases} \overline{V}, & \text{if } x \in C; \\ \emptyset, & \text{otherwise,} \end{cases} \quad \text{and} \quad N_C x = \begin{cases} V^\perp, & \text{if } x \in C; \\ \emptyset, & \text{otherwise.} \end{cases} \quad (6.36)$$

**Proposition 6.43** *Let  $C$  be a nonempty convex subset of  $\mathcal{H}$  and let  $x \in C$ . Then the following hold:*

- (i)  $T_C^\ominus x = N_C x$  and  $N_C^\ominus x = T_C x$ .
- (ii)  $x \in \text{core } C \Rightarrow T_C x = \mathcal{H} \Leftrightarrow N_C x = \{0\}$ .

*Proof.* (i): Since  $(C - x) \subset T_C x$ , Proposition 6.23(i) yields  $T_C^\ominus x \subset N_C x$ . Now let  $u \in N_C x$ . Then  $(\forall \lambda \in \mathbb{R}_{++}) \sup \langle \lambda(C - x) \mid u \rangle \leq 0$ . Hence,  $\sup \langle \text{cone}(C - x) \mid u \rangle \leq 0$  and therefore  $\sup \langle T_C x \mid u \rangle \leq 0$ , i.e.,  $u \in T_C^\ominus x$ . Altogether,  $T_C^\ominus x = N_C x$ . Furthermore, since  $T_C x$  is a nonempty closed convex cone, Corollary 6.33 yields  $T_C x = T_C^{\ominus\ominus} x = N_C^\ominus x$ .

(ii): If  $x \in \text{core } C$ , then  $\text{cone}(C - x) = \mathcal{H}$ . In turn,  $T_C x = \overline{\text{cone}}(C - x) = \mathcal{H}$ , and by (i),  $N_C x = T_C^\ominus x = \mathcal{H}^\ominus = \{0\}$ . Finally, it follows from (i) that  $N_C x = \{0\} \Rightarrow T_C x = \{0\}^\ominus = \mathcal{H}$ .  $\square$

Interior points of convex sets can be characterized via tangent and normal cones.

**Corollary 6.44** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $C$  be a nonempty convex subset of  $\mathcal{H}$ , and let  $x \in C$ . Then  $x \in \text{int } C \Leftrightarrow T_C x = \mathcal{H} \Leftrightarrow N_C x = \{0\}$ .*

*Proof.* Suppose that  $N_C x = \{0\}$ . Set  $U = \text{aff } C$  and  $V = U - U = U - x$ . Then  $C - x \subset U - x = V$ , and it follows from Proposition 6.23(i) that  $V^\perp = V^\ominus \subset (C - x)^\ominus = N_C x$ . Since  $N_C x = \{0\}$ , we obtain  $V^\perp = 0$  and thus  $V = \mathcal{H}$ . Hence  $\text{aff } C = \mathcal{H}$  and therefore  $\text{int } C = \text{ri } C \neq \emptyset$  by Fact 6.14(i). We have shown that

$$N_C x = \{0\} \quad \Rightarrow \quad \text{int } C \neq \emptyset. \quad (6.37)$$

The result now follows from Proposition 6.43(ii).  $\square$

**Proposition 6.45** *Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $\text{int } C \neq \emptyset$  and let  $x \in C$ . Then  $x \in \text{int } C \Leftrightarrow T_C x = \mathcal{H} \Leftrightarrow N_C x = \{0\}$ .*

*Proof.* Set  $D = C - x$ . Then  $0 \in D$  and  $\text{int } D = \text{int } C - x \neq \emptyset$ . In view of Proposition 6.17,  $0 \in \text{int } D \Leftrightarrow \overline{\text{cone}} D = \mathcal{H}$ , i.e.,  $x \in \text{int } C \Leftrightarrow T_C x = \mathcal{H}$ . The last equivalence is from Proposition 6.43.  $\square$

We conclude this section with a characterization of projections onto closed convex sets.

**Proposition 6.46** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , and let  $x$  and  $p$  be points in  $\mathcal{H}$ . Then  $p = P_C x \Leftrightarrow x - p \in N_C p$ .*

*Proof.* This follows at once from (3.6) and (6.31).  $\square$



## 6.5 Recession and Barrier Cones

**Definition 6.47** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . The *recession cone* of  $C$  is

$$\text{rec } C = \{x \in \mathcal{H} \mid x + C \subset C\}, \quad (6.38)$$

and the *barrier cone* of  $C$  is

$$\text{bar } C = \{u \in \mathcal{H} \mid \sup \langle C \mid u \rangle < +\infty\}. \quad (6.39)$$

**Proposition 6.48** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . Then the following hold:

- (i)  $\text{rec } C$  is a convex cone and  $0 \in \text{rec } C$ .
- (ii)  $\text{bar } C$  is a convex cone and  $C^\ominus \subset \text{bar } C$ .
- (iii) Suppose that  $C$  is bounded. Then  $\text{bar } C = \mathcal{H}$ .
- (iv) Suppose that  $C$  is a cone. Then  $\text{bar } C = C^\ominus$ .
- (v) Suppose that  $C$  is closed. Then  $(\text{bar } C)^\ominus = \text{rec } C$ .

*Proof.* (i): It is readily verified that  $0 \in \text{rec } C$ , that  $\text{rec } C + \text{rec } C \subset \text{rec } C$ , and that  $\text{rec } C$  is convex. Hence, the result follows from Proposition 6.3(ii).

(ii): Clear from (6.24) and (6.39).

(iii): By Cauchy–Schwarz,  $(\forall u \in \mathcal{H}) \sup \langle C \mid u \rangle \leq \|u\| \sup \|C\| < +\infty$ .

(iv): Take  $u \in \text{bar } C$ . Since  $C$  is a cone,  $\sup \langle C \mid u \rangle$  cannot be strictly positive, and hence  $u \in C^\ominus$ . Thus  $\text{bar } C \subset C^\ominus$ , while  $C^\ominus \subset \text{bar } C$  by (ii).

(v): Take  $x \in \text{rec } C$ . Then, for every  $u \in \text{bar } C$ ,  $\langle x \mid u \rangle + \sup \langle C \mid u \rangle = \sup \langle x + C \mid u \rangle \leq \sup \langle C \mid u \rangle < +\infty$ , which implies that  $\langle x \mid u \rangle \leq 0$  and hence that  $x \in (\text{bar } C)^\ominus$ . Thus,  $\text{rec } C \subset (\text{bar } C)^\ominus$ . Conversely, take  $x \in (\text{bar } C)^\ominus$  and  $y \in C$ , and set  $p = P_C(x + y)$ . By Proposition 6.46 and (ii),  $x + y - p \in N_C(p) = (C - p)^\ominus \subset \text{bar } (C - p) = \text{bar } C$ . Hence, since  $x \in (\text{bar } C)^\ominus$ , we obtain  $\langle x + y - p \mid x \rangle \leq 0$ , and (3.6) yields  $\|x + y - p\|^2 = \langle x + y - p \mid x \rangle + \langle x + y - p \mid y - p \rangle \leq 0$ . Hence  $x + y - p \in C$  and  $x \in \text{rec } C$ . Thus,  $(\text{bar } C)^\ominus \subset \text{rec } C$ .  $\square$

**Corollary 6.49** Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ . Then  $\text{rec } K = K$ .

*Proof.* It follows from Proposition 6.48(v), Proposition 6.48(iv), and Corollary 6.33 that  $\text{rec } K = (\text{bar } K)^\ominus = K^{\ominus\ominus} = K$ .  $\square$

The next result makes it clear why the recession cone of  $C$  is sometimes denoted by  $0^+C$ .

**Proposition 6.50** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then the following are equivalent:

- (i)  $x \in \text{rec } C$ .
- (ii) There exist sequences  $(x_n)_{n \in \mathbb{N}}$  in  $C$  and  $(\alpha_n)_{n \in \mathbb{N}}$  in  $]0, 1]$  such that  $\alpha_n \rightarrow 0$  and  $\alpha_n x_n \rightarrow x$ .

- (iii) *There exist sequences  $(x_n)_{n \in \mathbb{N}}$  in  $C$  and  $(\alpha_n)_{n \in \mathbb{N}}$  in  $]0, 1]$  such that  $\alpha_n \rightarrow 0$  and  $\alpha_n x_n \rightarrow x$ .*

*Proof.* Take  $y \in C$ .

(i) $\Rightarrow$ (ii): Proposition 6.48(i) yields  $(\forall n \in \mathbb{N}) (n+1)x \in \text{rec } C$ . Now define  $(\forall n \in \mathbb{N}) x_n = (n+1)x + y \in C$  and  $\alpha_n = 1/(n+1)$ . Then  $\alpha_n \rightarrow 0$  and  $\alpha_n x_n \rightarrow x$ .

(ii) $\Rightarrow$ (iii): Clear.

(iii) $\Rightarrow$ (i): The sequence  $(\alpha_n x_n + (1 - \alpha_n)y)_{n \in \mathbb{N}}$  lies in  $C$ , which is weakly sequentially closed by Theorem 3.32, and hence its weak limit  $x + y$  belongs to  $C$ . It follows that  $x \in \text{rec } C$ .  $\square$

**Corollary 6.51** *Suppose that  $\mathcal{H}$  is finite-dimensional. Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $C$  is bounded if and only if  $\text{rec } C = \{0\}$ .*

*Proof.* If  $C$  is bounded, then clearly  $\text{rec } C = \{0\}$ . Now assume that  $C$  is unbounded. Then there exist a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $C$  and a vector  $y \in \mathcal{H}$  such that  $\|x_n\| \rightarrow +\infty$  and  $x_n/\|x_n\| \rightarrow y$ . Hence  $\|y\| = 1$ , and thus Proposition 6.50 implies that  $y \in (\text{rec } C) \setminus \{0\}$ .  $\square$

**Corollary 6.52** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  such that  $0 \notin C$ . Then  $(\text{cone } C) \cup (\text{rec } C) = \overline{\text{cone } C}$ .*

*Proof.* It is clear that  $\text{cone } C \subset \overline{\text{cone } C}$ . Now take  $x \in \text{rec } C$ . By Proposition 6.50,  $x$  is the weak limit of a sequence  $(\alpha_n x_n)_{n \in \mathbb{N}}$ , where  $(\alpha_n)_{n \in \mathbb{N}}$  lies in  $]0, 1]$ ,  $\alpha_n \rightarrow 0$ , and  $(x_n)_{n \in \mathbb{N}}$  lies in  $C$ . Since  $(\alpha_n x_n)_{n \in \mathbb{N}}$  lies in  $\text{cone } C$ ,  $x$  belongs to the weak closure of  $\text{cone } C$ , which is  $\overline{\text{cone } C}$  by Theorem 3.32 and Proposition 6.2. Thus,

$$(\text{cone } C) \cup (\text{rec } C) \subset \overline{\text{cone } C}. \quad (6.40)$$

Conversely, take  $x \in \overline{\text{cone } C}$ . By Proposition 6.2(ii), there exist sequences  $(\alpha_n)_{n \in \mathbb{N}}$  in  $\mathbb{R}_{++}$  and  $(x_n)_{n \in \mathbb{N}}$  in  $C$  such that  $\alpha_n x_n \rightarrow x$ . After passing to subsequences if necessary, we assume that  $\alpha_n \rightarrow \alpha \in [0, +\infty]$ . If  $\alpha = +\infty$ , then, since  $\|\alpha_n x_n\| \rightarrow \|x\|$ , it follows that  $x_n \rightarrow 0$  and, in turn, that  $0 \in C$ , which violates our hypothesis. Hence  $\alpha \in \mathbb{R}_+$ . If  $\alpha = 0$ , then  $x \in \text{rec } C$  by Proposition 6.50. Otherwise,  $\alpha \in \mathbb{R}_{++}$ , in which case  $x_n = (\alpha_n x_n)/\alpha_n \rightarrow x/\alpha \in C$ , and hence  $x \in \alpha C \subset \text{cone } C$ .  $\square$

## Exercises

**Exercise 6.1** Find a convex subset  $C$  of  $\mathbb{R}^2$  such that  $\overline{\text{cone } C} \neq \bigcup_{\lambda \in \mathbb{R}_+} \lambda C$ .

**Exercise 6.2** Prove Proposition 6.3.

**Exercise 6.3** Prove Proposition 6.4.

**Exercise 6.4** Let  $\alpha \in \mathbb{R}_{++}$  and set  $K_\alpha = \{(x, \xi) \in \mathcal{H} \times \mathbb{R} \mid \|x\| \leq \alpha\xi\}$ . Show that  $K_\alpha^\oplus = K_{1/\alpha}$  and conclude that  $K_1$  is self-dual.

**Exercise 6.5** Let  $C$  be a convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$  be such that  $\text{cone}(C - x)$  is a linear subspace. Show that  $x \in C$ .

**Exercise 6.6** Let  $C$  be a cone in  $\mathcal{H}$ , let  $\mathcal{K}$  be a real Hilbert space, let  $D$  be a cone in  $\mathcal{K}$ , and let  $L: \mathcal{H} \rightarrow \mathcal{K}$  be positively homogeneous. Show that  $L(C)$  is a cone in  $\mathcal{K}$  and that  $L^{-1}(D)$  is a cone in  $\mathcal{H}$ .

**Exercise 6.7** Let  $C$  and  $D$  be nonempty convex subsets of  $\mathcal{H}$ . Show that the implication  $C \subset D \Rightarrow \text{sri } C \subset \text{sri } D$  is false.

**Exercise 6.8** Suppose that  $\mathcal{H} = \ell^2(\mathbb{N})$ , and set  $C = \text{lev}_{\leq 1} f$ , where  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi_k)_{k \in \mathbb{N}} \mapsto \sum_{k \in \mathbb{N}} |\xi_k|$ . Show that  $0 \in \text{ri } C \setminus \text{sri } C$ .

**Exercise 6.9** In connection with Proposition 6.8, find an infinite set  $C$  in  $\mathbb{R}^2$  such that the convex cone  $\sum_{c \in C} \mathbb{R}_+ c$  is not closed.

**Exercise 6.10** Show that the conclusion of Proposition 6.16 fails if the assumption that  $\text{int } C \neq \emptyset$  is omitted.

**Exercise 6.11** Let  $C$  be a subset of  $\mathcal{H}$ . Show that  $-(C^\ominus) = (-C)^\ominus$ , which justifies writing simply  $-C^\ominus$  for these sets, and that  $C^{\ominus\ominus} = C^{\oplus\oplus}$ .

**Exercise 6.12** Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ . Then  $K$  is *acute* if  $K \subset K^\oplus$ , and  $K$  is *obtuse* if  $K^\oplus \subset K$  (hence, a cone is self-dual if and only if it is both acute and obtuse). Prove that  $K$  is obtuse if and only if  $K^\oplus$  is acute.

**Exercise 6.13** Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ . Prove that  $K$  is a linear subspace if and only if  $K \cap K^\oplus = \{0\}$ .

**Exercise 6.14** Let  $K$  be a nonempty closed convex solid cone in  $\mathcal{H}$ . Show that  $K^\ominus$  is pointed.

**Exercise 6.15** Let  $K$  be a nonempty closed convex pointed cone in  $\mathcal{H}$ . Show the following:

- (i) If  $\mathcal{H}$  is finite-dimensional, then  $K^\ominus$  is solid.
- (ii) In general,  $K^\ominus$  fails to be solid.

**Exercise 6.16** Let  $N$  be a strictly positive integer, set  $I = \{1, \dots, N\}$ , and set

$$K = \{(\xi_i)_{i \in I} \in \mathbb{R}^N \mid \xi_1 \geq \xi_2 \geq \dots \geq \xi_N \geq 0\}. \quad (6.41)$$

Show that  $K$  is a nonempty pointed closed convex cone. Use Proposition 6.36 to show that

$$K^\ominus = \{(\zeta_i)_{i \in I} \in \mathbb{R}^N \mid \zeta_1 \leq 0, \zeta_1 + \zeta_2 \leq 0, \dots, \zeta_1 + \dots + \zeta_N \leq 0\}. \quad (6.42)$$

Furthermore, use Exercise 6.15(i) to show that  $K$  is solid.

**Exercise 6.17** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ , and let  $x \in C$ . Show that  $x \in \text{qri } C \Leftrightarrow N_C x$  is a linear subspace.

**Exercise 6.18** Let  $C$  be a convex subset of  $\mathcal{H}$ . Show that  $\text{span } C$  is closed if and only if  $0 \in \text{sri}(C - C)$ . Furthermore, provide an example in which  $\text{span } C$  is not closed even though  $C$  is.

**Exercise 6.19** Using Definition 6.37 directly, provide a proof of Example 6.38 and of Example 6.42.

**Exercise 6.20** Find a closed convex subset  $C$  and a point  $x \in C$  such that  $T_C x = \mathcal{H}$  but  $\text{int } C = \emptyset$ . Compare with Proposition 6.45.

**Exercise 6.21** Suppose that  $\mathcal{H} = \mathbb{R}^2$  and set  $C = \{(\xi_1, \xi_2) \in \mathcal{H} \mid \xi_1^2 \leq \xi_2\}$ . Determine  $\text{bar } C$  and observe that  $\text{bar } C$  is not closed.

**Exercise 6.22** Let  $v \in \mathcal{H}$  and let  $U$  be a nonempty finite subset of  $\mathcal{H}$ . Using Farkas's lemma (Theorem 6.35), show that  $v \in \text{span } U$  if and only if  $\bigcap_{u \in U} \{u\}^\perp \subset \{v\}^\perp$ .

**Exercise 6.23** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Show that  $\text{ran}(\text{Id} - P_C)$  is a cone and that  $\overline{\text{ran}}(\text{Id} - P_C) \subset (\text{rec } C)^\ominus$ .

**Exercise 6.24** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $u \in \mathcal{H}$ . Show that  $u \in (\text{rec } C)^\ominus$  if and only if

$$(\forall x \in \mathcal{H}) \quad \lim_{\lambda \rightarrow +\infty} \frac{P_C(x + \lambda u)}{\lambda} = 0. \quad (6.43)$$

**Exercise 6.25** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Show that  $\overline{\text{ran}}(\text{Id} - P_C) = (\text{rec } C)^\ominus$ .

**Exercise 6.26** Suppose that  $\mathcal{H}$  is infinite-dimensional and separable, and let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal basis of  $\mathcal{H}$ . Set

$$C = \{x \in \mathcal{H} \mid (\forall n \in \mathbb{N}) \quad |\langle x \mid e_n \rangle| \leq n\}. \quad (6.44)$$

Show that  $C$  is an unbounded closed convex set, and that  $\text{rec } C = \{0\}$ . Compare with Corollary 6.51.

**Exercise 6.27** Show that the conclusion of Corollary 6.52 fails if the assumption  $0 \notin C$  is replaced by  $0 \in C$ .

# Chapter 7

## Support Functions and Polar Sets

In this chapter, we develop basic results concerning support points, including the Bishop–Phelps theorem and the representation of a nonempty closed convex set as the intersection of the closed half-spaces containing it. Polar sets are also studied.

### 7.1 Support Points

**Definition 7.1** Let  $C$  be a nonempty subset of  $\mathcal{H}$ , let  $x \in C$ , and suppose that  $u \in \mathcal{H} \setminus \{0\}$ . If

$$\sup \langle C \mid u \rangle \leq \langle x \mid u \rangle, \quad (7.1)$$

then  $\{y \in \mathcal{H} \mid \langle y \mid u \rangle = \langle x \mid u \rangle\}$  is a *supporting hyperplane* of  $C$  at  $x$ , and  $x$  is a *support point* of  $C$  with *normal vector*  $u$ . The set of support points of  $C$  is denoted by  $\text{spts } C$  and the closure of  $\text{spts } C$  by  $\overline{\text{spts } C}$ .

**Proposition 7.2** Let  $C$  and  $D$  be nonempty subsets of  $\mathcal{H}$  such that  $C \subset D$ . Then  $C \cap \text{spts } D \subset \text{spts } C = C \cap \overline{\text{spts } C}$ .

*Proof.* Let  $x \in C \cap \text{spts } D$ . Then  $x \in C$  and there exists  $u \in \mathcal{H} \setminus \{0\}$  such that  $\sup \langle C \mid u \rangle \leq \sup \langle D \mid u \rangle \leq \langle x \mid u \rangle$ . Hence,  $x \in \text{spts } C$ . This verifies the announced inclusion, and the equality is clear from the definition.  $\square$

**Proposition 7.3** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . Then

$$\text{spts } C = \{x \in C \mid N_C x \setminus \{0\} \neq \emptyset\} = N_C^{-1}(\mathcal{H} \setminus \{0\}). \quad (7.2)$$

*Proof.* Let  $x \in C$ . Then  $x \in \text{spts } C \Leftrightarrow (\exists u \in \mathcal{H} \setminus \{0\}) \sup \langle C - x \mid u \rangle \leq 0 \Leftrightarrow (\exists u \in \mathcal{H} \setminus \{0\}) u \in (C - x)^\ominus = N_C x$ .  $\square$

**Theorem 7.4 (Bishop–Phelps)** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $\text{spts } C = P_C(\mathcal{H} \setminus \{0\})$  and  $\overline{\text{spts } C} = \text{bdry } C$ .

*Proof.* We assume that  $C \neq \mathcal{H}$ . Fix  $\varepsilon \in \mathbb{R}_{++}$  and let  $x$  be a support point of  $C$  with normal vector  $u$ . Then  $\sup \langle C - x \mid (x + \varepsilon u) - x \rangle \leq 0$  and Theorem 3.14 implies that  $x = P_C(x + \varepsilon u)$ . Since  $u \neq 0$ , we note that  $x + \varepsilon u \notin C$ . Hence  $\text{spts } C \subset P_C(\mathcal{H} \setminus C)$  and  $x \in \text{bdry } C$ . Thus  $\overline{\text{spts } C} \subset \text{bdry } C$ . Next, assume that  $P_C y = x$ , for some  $y \in \mathcal{H} \setminus C$ . Proposition 6.46 asserts that  $0 \neq y - x \in N_C x$ ; hence  $x \in \text{spts } C$  by Proposition 7.3. Thus,

$$\text{spts } C = P_C(\mathcal{H} \setminus C). \quad (7.3)$$

Now take  $z \in \text{bdry } C$ . Then there exists  $y \in \mathcal{H} \setminus C$  such that  $\|z - y\| \leq \varepsilon$ . Set  $p = P_C y$ . Then  $p \in \text{spts } C$  and Proposition 4.8 yields  $\|p - z\| = \|P_C y - P_C z\| \leq \|y - z\| \leq \varepsilon$ . Therefore,  $z \in \overline{\text{spts } C}$  and hence  $\text{bdry } C \subset \overline{\text{spts } C}$ .  $\square$

**Proposition 7.5** *Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $\text{int } C \neq \emptyset$ . Then  $\text{bdry } C \subset \overline{\text{spts } C}$  and  $C \cap \text{bdry } C \subset \text{spts } C$ .*

*Proof.* If  $C = \mathcal{H}$ , the result is clear. We therefore assume otherwise. Set  $D = \overline{C}$  and let  $x \in \text{bdry } C \subset D$ . Then Proposition 3.36(iii) yields  $x \in D \setminus (\text{int } D)$ , and Proposition 6.45 guarantees the existence of a vector  $u \in N_D x \setminus \{0\}$ . Hence, it follows from Proposition 7.3 that  $x \in \text{spts } D = \text{spts } \overline{C}$ . Therefore,  $\text{bdry } C \subset \text{spts } \overline{C}$  and, furthermore, Proposition 7.2 yields  $C \cap \text{bdry } C \subset C \cap \text{spts } \overline{C} = \text{spts } C$ .  $\square$

**Corollary 7.6** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and suppose that one of the following holds:*

- (i)  $\text{int } C \neq \emptyset$ .
- (ii)  $C$  is a closed affine subspace.
- (iii)  $\mathcal{H}$  is finite-dimensional.

*Then  $\text{spts } C = \text{bdry } C$ .*

*Proof.* (i): Combine Theorem 7.4 and Proposition 7.5.

(ii): Assume that  $C \neq \mathcal{H}$ , let  $V = C - C$  be the closed linear subspace parallel to  $C$ , let  $x \in \text{bdry } C$ , and let  $u \in V^\perp \setminus \{0\}$ . Then  $C = x + V$ ,  $P_C(x + u) = P_{x+V}(x + u) = x + P_V u = x$  by Proposition 3.17, and Theorem 7.4 therefore yields  $x \in \text{spts } C$ .

(iii): In view of (i), assume that  $\text{int } C = \emptyset$ . Let  $D = \text{aff } C$  and let  $x \in \text{bdry } C$ . Then  $D \neq \mathcal{H}$ , since Proposition 6.12 asserts that  $\text{core } C = \text{int } C = \emptyset$ . On the other hand, since  $\mathcal{H}$  is finite-dimensional,  $D$  is closed. Thus,  $D$  is a proper closed affine subspace of  $\mathcal{H}$  and therefore  $x \in \text{bdry } D$ . Altogether, (ii) and Proposition 7.2 imply that  $x \in C \cap \text{spts } D \subset \text{spts } C$ .  $\square$

A boundary point of a closed convex set  $C$  need not be a support point of  $C$ , even if  $C$  is a closed convex cone.

**Example 7.7** Suppose that  $\mathcal{H} = \ell^2(\mathbb{N})$  and set  $C = \ell_+^2(\mathbb{N})$ , which is a nonempty closed convex cone with empty interior (see Example 6.7). For every  $x = (\xi_k)_{k \in \mathbb{N}} \in \mathcal{H}$ , we have  $P_C x = (\max\{\xi_k, 0\})_{k \in \mathbb{N}}$ . Consequently,

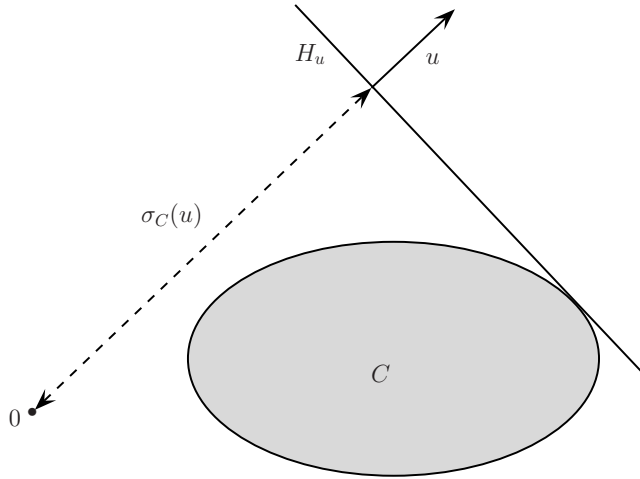
by Theorem 7.4,  $x = (\xi_k)_{k \in \mathbb{N}} \in C$  is a support point of  $C$  if and only if  $\{k \in \mathbb{N} \mid \xi_k = 0\} \neq \emptyset$ . Therefore,  $(1/2^k)_{k \in \mathbb{N}} \in \text{bdry } C \setminus \text{spts } C$ .

## 7.2 Support Functions

**Definition 7.8** Let  $C \subset \mathcal{H}$ . The *support function* of  $C$  is

$$\sigma_C : \mathcal{H} \rightarrow [-\infty, +\infty] : u \mapsto \sup \langle C \mid u \rangle. \quad (7.4)$$

Let  $C$  be a nonempty subset of  $\mathcal{H}$ , and suppose that  $u \in \mathcal{H} \setminus \{0\}$ . If  $\sigma_C(u) < +\infty$ , then (7.4) implies that  $\{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \sigma_C(u)\}$  is the smallest closed half-space with outer normal  $u$  that contains  $C$  (see Figure 7.1). Now suppose that, for some  $x \in C$ , we have  $\sigma_C(u) = \langle x \mid u \rangle$ . Then  $x$  is a support point of  $C$  and  $\{x \in \mathcal{H} \mid \langle x \mid u \rangle = \sigma_C(u)\}$  is a supporting hyperplane of  $C$  at  $x$ .



**Fig. 7.1** Support function of  $C$  evaluated at a vector  $u$  such that  $\|u\| = 1$ . The half-space  $H_u = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \sigma_C(u)\}$  contains  $C$ , and it follows from Example 3.21 that the distance from 0 to the hyperplane  $\{x \in \mathcal{H} \mid \langle x \mid u \rangle = \sigma_C(u)\}$  is  $\sigma_C(u)$ .

**Proposition 7.9** Let  $C \subset \mathcal{H}$  and set

$$(\forall u \in \mathcal{H}) \quad H_u = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \sigma_C(u)\}. \quad (7.5)$$

Then  $\overline{\text{conv}} C = \bigcap_{u \in \mathcal{H}} H_u$ .

*Proof.* The result is clear when  $C = \emptyset$ . So assume that  $C \neq \emptyset$  and set  $D = \bigcap_{u \in \mathcal{H}} H_u$ . Since  $(H_u)_{u \in \mathcal{H}}$  is a family of closed convex sets each of which contains  $C$ , we deduce that  $D$  is closed and convex and that  $\overline{\text{conv}} C \subset D$ . Conversely, take  $x \in D$  and set  $p = P_{\overline{\text{conv}} C} x$ . By (3.6),  $\sigma_{\overline{\text{conv}} C}(x - p) \leq \langle p \mid x - p \rangle$ . On the other hand, since  $x \in D \subset H_{x-p}$ , we have  $\langle x \mid x - p \rangle \leq \sigma_C(x - p)$ . Thus,  $\|x - p\|^2 = \langle x \mid x - p \rangle - \langle p \mid x - p \rangle \leq \sigma_C(x - p) - \sigma_{\overline{\text{conv}} C}(x - p) \leq 0$ , and we conclude that  $x = p \in \overline{\text{conv}} C$ .  $\square$

**Corollary 7.10** *Every closed convex subset of  $\mathcal{H}$  is the intersection of all the closed half-spaces of which it is a subset.*

**Proposition 7.11** *Let  $C \subset \mathcal{H}$ . Then  $\sigma_C = \sigma_{\overline{\text{conv}} C}$ .*

*Proof.* Assume that  $C \neq \emptyset$  and fix  $u \in \mathcal{H}$ . Since  $C \subset \overline{\text{conv}} C$ , we have  $\sigma_C(u) \leq \sigma_{\overline{\text{conv}} C}(u)$ . Conversely, let  $x \in \text{conv} C$ ; by Proposition 3.4, we suppose that  $x = \sum_{i \in I} \alpha_i x_i$  for some finite families  $(x_i)_{i \in I}$  in  $C$  and  $(\alpha_i)_{i \in I}$  in  $]0, 1]$  such that  $\sum_{i \in I} \alpha_i = 1$ . Then  $\langle x \mid u \rangle = \sum_{i \in I} \alpha_i \langle x_i \mid u \rangle \leq \sum_{i \in I} \alpha_i \sigma_C(u) = \sigma_C(u)$ . Therefore  $\sigma_{\text{conv} C}(u) = \sup_{z \in \text{conv} C} \langle z \mid u \rangle \leq \sigma_C(u)$ . Finally, let  $x \in \overline{C}$ , say  $x = \lim x_n$  for some sequence  $(x_n)_{n \in \mathbb{N}}$  in  $C$ . Then  $\langle x \mid u \rangle = \lim \langle x_n \mid u \rangle \leq \sigma_C(u)$ . Hence,  $\sigma_{\overline{C}}(u) = \sup_{z \in \overline{C}} \langle z \mid u \rangle \leq \sigma_C(u)$ . Altogether,  $\sigma_{\overline{\text{conv}} C}(u) \leq \sigma_C(u)$ .  $\square$

## 7.3 Polar Sets

**Definition 7.12** Let  $C \subset \mathcal{H}$ . The *polar set* of  $C$  is

$$C^\odot = \text{lev}_{\leq 1} \sigma_C = \{u \in \mathcal{H} \mid (\forall x \in C) \langle x \mid u \rangle \leq 1\}. \quad (7.6)$$

**Example 7.13** Let  $K$  be a cone in  $\mathcal{H}$ . Then  $K^\odot = K^\ominus$  (Exercise 7.10). In particular, if  $K$  is a linear subspace, then Proposition 6.22 yields  $K^\odot = K^\ominus = K^\perp$ .

**Proposition 7.14** *Let  $C$  and  $D$  be subsets of  $\mathcal{H}$ . Then the following hold:*

- (i)  $C^\perp \subset C^\ominus \subset C^\odot$ .
- (ii)  $0 \in C^\odot$ , and  $C^\odot$  is closed and convex.
- (iii)  $C \cup \{0\} \subset C^{\odot\odot}$ .
- (iv) Suppose that  $C \subset D$ . Then  $D^\odot \subset C^\odot$  and  $C^{\odot\odot} \subset D^{\odot\odot}$ .
- (v)  $C^{\odot\odot\odot} = C^\odot$ .
- (vi)  $(\overline{\text{conv}} C)^\odot = C^\odot$ .

*Proof.* (i)–(iv): Immediate consequences of (2.2), (6.24), and (7.6).

(v): By (iii),  $C \subset C^{\odot\odot}$  and  $C^\odot \subset (C^\odot)^{\odot\odot} = C^{\odot\odot\odot}$ . Hence, using (iv), we obtain  $C^{\odot\odot\odot} = (C^{\odot\odot})^\odot \subset C^\odot \subset C^{\odot\odot\odot}$ .

(vi): By Proposition 7.11,  $\sigma_C = \sigma_{\overline{\text{conv}} C}$ .  $\square$



**Remark 7.15** In view of Example 7.13 and Proposition 7.14, we obtain the following properties for an arbitrary cone  $K$  in  $\mathcal{H}$ :  $K \subset K^{\ominus\ominus}$ ,  $K^{\ominus\ominus\ominus} = K^{\ominus}$ , and  $(\overline{\text{conv}} K)^{\ominus} = K^{\ominus}$ . In particular, we retrieve the following well-known facts for a linear subspace  $V$  of  $\mathcal{H}$ :  $V \subset V^{\perp\perp}$ ,  $V^{\perp\perp\perp} = V^{\perp}$ , and  $\overline{V}^{\perp} = V^{\perp}$ .

**Theorem 7.16** *Let  $C \subset \mathcal{H}$ . Then  $C^{\odot\odot} = \overline{\text{conv}}(C \cup \{0\})$ .*

*Proof.* By Proposition 7.14(ii)&(iii),  $\overline{\text{conv}}(C \cup \{0\}) \subset \overline{\text{conv}} C^{\odot\odot} = C^{\odot\odot}$ . Conversely, suppose that  $x \in \text{lev}_{\leq 1} \sigma_{C^{\odot\odot}} \setminus \overline{\text{conv}}(C \cup \{0\})$ . It follows from Theorem 3.38 that there exists  $u \in \mathcal{H} \setminus \{0\}$  such that

$$\langle x | u \rangle > \sigma_{\overline{\text{conv}}(C \cup \{0\})}(u) \geq \max\{\sigma_C(u), 0\}. \quad (7.7)$$

After scaling  $u$  if necessary, we assume that  $\langle x | u \rangle > 1 \geq \sigma_C(u)$ . Hence,  $u \in C^{\odot}$  and therefore  $1 < \langle u | x \rangle \leq \sigma_{C^{\odot}}(x) \leq 1$ , which is impossible.  $\square$

**Corollary 7.17** *Let  $C \subset \mathcal{H}$ . Then the following hold:*

- (i)  $C$  is closed, convex, and contains the origin if and only if  $C^{\odot\odot} = C$ .
- (ii)  $C$  is a nonempty closed convex cone if and only if  $C^{\ominus\ominus} = C$ .
- (iii)  $C$  is a closed linear subspace if and only if  $C^{\perp\perp} = C$ .

*Proof.* (i): Combine Proposition 7.14(ii) and Theorem 7.16.

(ii): Combine (i), Proposition 6.23(ii), and Example 7.13.

(iii): Combine (ii) and Proposition 6.22.  $\square$

## Exercises

**Exercise 7.1** Let  $C \subset \mathcal{H}$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $\rho \in \mathbb{R} \setminus \{0\}$ .

- (i) Show that  $\sigma_C \circ \rho \text{Id} = \sigma_{\rho C}$ .
- (ii) Show that  $\sigma_C \circ \gamma \text{Id} = \gamma \sigma_C$ .
- (iii) Provide an example in which  $\sigma_C \circ \rho \text{Id} \neq \rho \sigma_C$ .

**Exercise 7.2** Let  $C$  and  $D$  be subsets of  $\mathcal{H}$ . Show that  $\sigma_{C+D} = \sigma_C + \sigma_D$ .

**Exercise 7.3** Suppose that  $\mathcal{H} = \mathbb{R}$  and let  $-\infty < \alpha < \beta < +\infty$ . Determine the support function  $\sigma_C$  in each of the following cases:

- (i)  $C = \{\alpha\}$ .
- (ii)  $C = [\alpha, \beta]$ .
- (iii)  $C = ]-\infty, \alpha]$ .
- (iv)  $C = [\beta, +\infty[$ .
- (v)  $C = \mathbb{R}$ .

**Exercise 7.4** Let  $C \subset \mathcal{H}$  and let  $\gamma \in \mathbb{R}_{++}$ . Show that  $(\gamma C)^{\odot} = \gamma^{-1} C^{\odot}$ .

**Exercise 7.5** Let  $C$  and  $D$  be subsets of  $\mathcal{H}$ . Show that  $(C \cup D)^\circ = C^\circ \cap D^\circ$ .

**Exercise 7.6** Provide two subsets  $C$  and  $D$  of  $\mathbb{R}$  such that  $(C \cap D)^\circ \neq C^\circ \cup D^\circ$ .

**Exercise 7.7** Set  $C = B(0; 1)$ . Show that  $C^\circ = C$ .

**Exercise 7.8** Let  $C \subset \mathcal{H}$  be such that  $C^\circ = C$ . Show that  $C = B(0; 1)$ .

**Exercise 7.9** Suppose that  $\mathcal{H} = \mathbb{R}^N$ , let  $p \in [1, +\infty]$ , and set  $C_p = \{x \in \mathcal{H} \mid \|x\|_p \leq 1\}$ , where

$$\|x\|_p = \begin{cases} \sqrt[p]{\sum_{k=1}^N |\xi_k|^p}, & \text{if } 1 \leq p < +\infty; \\ \max_{1 \leq k \leq N} |\xi_k|, & \text{if } p = +\infty. \end{cases} \quad (7.8)$$

Compute  $C_p^\circ$ .

**Exercise 7.10** Let  $C$  be a subset of  $\mathcal{H}$ . Show that if  $C$  is a cone, then  $C^\circ = C^\circ$ . Show that this identity fails if  $C$  is not a cone.

# Chapter 8

## Convex Functions

Convex functions, which lie at the heart of modern optimization, are introduced in this chapter. We study operations that preserve convexity, and the interplay between various continuity properties.

### 8.1 Definition and Examples

**Definition 8.1** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is *convex* if its epigraph  $\text{epi } f = \{(x, \xi) \in \mathcal{H} \times \mathbb{R} \mid f(x) \leq \xi\}$  is a convex subset of  $\mathcal{H} \times \mathbb{R}$ . Moreover,  $f$  is *concave* if  $-f$  is convex.

**Proposition 8.2** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be convex. Then its domain  $\text{dom } f = \{x \in \mathcal{H} \mid f(x) < +\infty\}$  is convex.

*Proof.* Set  $L: \mathcal{H} \times \mathbb{R} \rightarrow \mathcal{H}: (x, \xi) \mapsto x$ . Then  $L$  is linear and  $\text{dom } f = L(\text{epi } f)$ . It therefore follows from Proposition 3.5 that  $\text{dom } f$  is convex.  $\square$

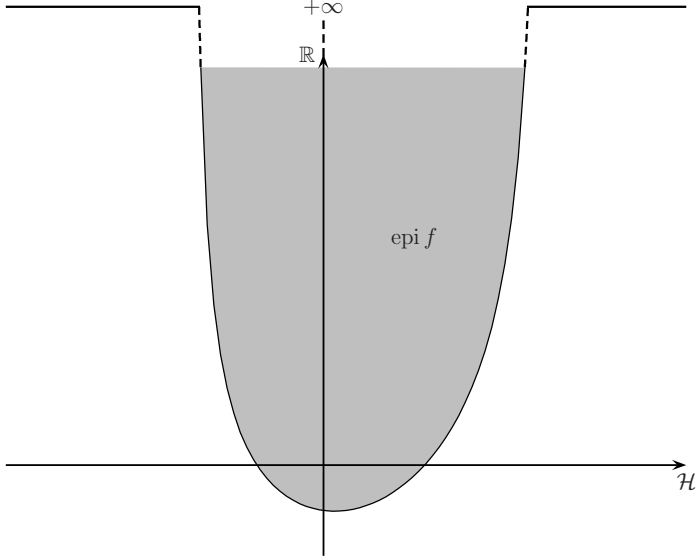
**Example 8.3** Let  $C$  be a subset of  $\mathcal{H}$ . Then  $\text{epi } \iota_C = C \times \mathbb{R}_+$ , and hence  $\iota_C$  is a convex function if and only if  $C$  is a convex set.

**Proposition 8.4** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is convex if and only if

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y). \quad (8.1)$$

*Proof.* Note that  $f \equiv +\infty \Leftrightarrow \text{epi } f = \emptyset \Leftrightarrow \text{dom } f = \emptyset$ , in which case  $f$  is convex and (8.1) holds. So we assume that  $\text{dom } f \neq \emptyset$ , and take  $(x, \xi) \in \text{epi } f$ ,  $(y, \eta) \in \text{epi } f$ , and  $\alpha \in ]0, 1[$ . First, suppose that  $f$  is convex. Then  $\alpha(x, \xi) + (1 - \alpha)(y, \eta) \in \text{epi } f$  and therefore

$$f(\alpha x + (1 - \alpha)y) \leq \alpha \xi + (1 - \alpha)\eta. \quad (8.2)$$



**Fig. 8.1** A function  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  is convex if  $\text{epi } f$  is convex in  $\mathcal{H} \times \mathbb{R}$ .

Letting  $\xi \downarrow f(x)$  and  $\eta \downarrow f(y)$  in (8.2), we obtain (8.1). Now assume that  $f$  satisfies (8.1). Then  $f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y) \leq \alpha \xi + (1 - \alpha)\eta$ , and therefore  $\alpha(x, \xi) + (1 - \alpha)(y, \eta) \in \text{epi } f$ .  $\square$

**Corollary 8.5** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be convex. Then, for every  $\xi \in \mathbb{R}$ ,  $\text{lev}_{\leq \xi} f$  is convex.

**Definition 8.6** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper function. Then  $f$  is *strictly convex* if

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ x \neq y \Rightarrow f(\alpha x + (1 - \alpha)y) < \alpha f(x) + (1 - \alpha)f(y). \quad (8.3)$$

Now let  $C$  be a nonempty subset of  $\text{dom } f$ . Then  $f$  is *convex on  $C$*  if

$$(\forall x \in C)(\forall y \in C)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y), \quad (8.4)$$

and  $f$  is *strictly convex on  $C$*  if

$$(\forall x \in C)(\forall y \in C)(\forall \alpha \in ]0, 1[) \\ x \neq y \Rightarrow f(\alpha x + (1 - \alpha)y) < \alpha f(x) + (1 - \alpha)f(y). \quad (8.5)$$

**Example 8.7** The function  $\|\cdot\|$  is convex. If  $\mathcal{H} \neq \{0\}$ , then  $\|\cdot\|$  is not strictly convex.

*Proof.* Convexity is clear. Now take  $x \in \mathcal{H} \setminus \{0\}$  and  $\alpha \in ]0, 1[$ . Then  $\|\alpha x + (1 - \alpha)0\| = \alpha\|x\| + (1 - \alpha)\|0\|$ . Hence  $\|\cdot\|$  is not strictly convex.  $\square$

**Example 8.8** The function  $\|\cdot\|^2$  is strictly convex.

*Proof.* This follows from Corollary 2.14.  $\square$

**Proposition 8.9** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is convex if and only if, for all finite families  $(\alpha_i)_{i \in I}$  in  $]0, 1[$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $(x_i)_{i \in I}$  in  $\text{dom } f$ , we have

$$f\left(\sum_{i \in I} \alpha_i x_i\right) \leq \sum_{i \in I} \alpha_i f(x_i). \quad (8.6)$$

*Proof.* Assume that  $f$  is convex and fix finite families  $(x_i)_{i \in I}$  in  $\text{dom } f$  and  $(\alpha_i)_{i \in I}$  in  $]0, 1[$  such that  $\sum_{i \in I} \alpha_i = 1$ . Then  $(x_i, f(x_i))_{i \in I}$  lies in  $\text{epi } f$ . Hence, by convexity,  $(\sum_{i \in I} \alpha_i x_i, \sum_{i \in I} \alpha_i f(x_i)) \in \text{conv epi } f = \text{epi } f$ , which gives (8.6). The converse implication follows at once from Proposition 8.4.  $\square$

**Corollary 8.10** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ . Then the following are equivalent:

- (i)  $f$  is convex.
- (ii) For all finite families  $(\alpha_i)_{i \in I}$  in  $]0, 1[$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $(x_i)_{i \in I}$  in  $\text{dom } f$ , we have  $f(\sum_{i \in I} \alpha_i x_i) \leq \sum_{i \in I} \alpha_i f(x_i)$ .
- (iii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H})(\forall \alpha \in ]0, 1[) f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y)$ .

**Proposition 8.11** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $f$  is strictly convex if and only if for all finite families  $(\alpha_i)_{i \in I}$  in  $]0, 1[$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $(x_i)_{i \in I}$  in  $\text{dom } f$ , we have  $f(\sum_{i \in I} \alpha_i x_i) \leq \sum_{i \in I} \alpha_i f(x_i)$ , and equality holds if and only if  $\{x_i\}_{i \in I}$  is a singleton.

*Proof.* Assume first that  $f$  is strictly convex. We prove the corresponding implication by induction on  $m$ , the number of elements in  $I$ . The result is clear for  $m = 2$ . We assume now that  $m \geq 3$ , that  $I = \{1, \dots, m\}$ , and that the result is true for families containing  $m - 1$  or fewer points, and we set  $\mu = f(\sum_{i \in I} \alpha_i x_i) = \sum_{i \in I} \alpha_i f(x_i)$ . Then

$$\mu \leq (1 - \alpha_m) f\left(\sum_{i=1}^{m-1} \frac{\alpha_i}{1 - \alpha_m} x_i\right) + \alpha_m f(x_m) \quad (8.7)$$

$$\leq (1 - \alpha_m) \sum_{i=1}^{m-1} \frac{\alpha_i}{1 - \alpha_m} f(x_i) + \alpha_m f(x_m) \quad (8.8)$$

$$= \mu. \quad (8.9)$$

Hence, the inequalities (8.7) and (8.8) are actually equalities, and the induction hypothesis yields  $(1 - \alpha_m)^{-1}(\sum_{i=1}^{m-1} \alpha_i x_i) = x_m$  and  $x_1 = \dots = x_{m-1}$ .

Therefore,  $x_1 = \cdots = x_m$ , as required. The reverse implication is clear by considering the case in which  $I$  contains exactly two elements.  $\square$

We now state a simple convexity condition for functions on the real line (for an extension, see Proposition 17.10).

**Proposition 8.12** *Let  $\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be a proper function that is differentiable on a nonempty open interval  $I$  in  $\text{dom } \phi$ . Then the following hold:*

- (i) *Suppose that  $\phi'$  is increasing on  $I$ . Then  $\phi + \iota_I$  is convex.*
- (ii) *Suppose that  $\phi'$  is strictly increasing on  $I$ . Then  $\phi$  is strictly convex on  $I$ .*

*Proof.* Fix  $x$  and  $y$  in  $I$ , and  $\alpha \in ]0, 1[$ . Set  $\psi: \mathbb{R} \rightarrow ]-\infty, +\infty]: z \mapsto \alpha\phi(x) + (1 - \alpha)\phi(z) - \phi(\alpha x + (1 - \alpha)z)$ . Now let  $z \in I$ . Then

$$\psi'(z) = (1 - \alpha)(\phi'(z) - \phi'(\alpha x + (1 - \alpha)z)) \quad (8.10)$$

and  $\psi'(x) = 0$ .

(i): It follows from (8.10) that  $\psi'(z) \leq 0$  if  $z < x$ , and that  $\psi'(z) \geq 0$  if  $z > x$ . Thus,  $\psi$  achieves its minimum on  $I$  at  $x$ . In particular,  $\psi(y) \geq \psi(x) = 0$ , and the convexity of  $\phi + \iota_I$  follows from Proposition 8.4.

(ii): It follows from (8.10) that  $\psi'(z) < 0$  if  $z < x$ , and  $\psi'(z) > 0$  if  $z > x$ . Thus,  $\psi$  achieves its strict minimum on  $I$  at  $x$ . Hence, if  $y \neq x$ ,  $\psi(y) > \psi(x) = 0$ , and  $\phi + \iota_I$  is strictly convex by (8.3).  $\square$

**Example 8.13** Let  $\psi: \mathbb{R} \rightarrow \mathbb{R}$  be increasing, let  $\alpha \in \mathbb{R}$ , and set

$$\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} \int_{\alpha}^x \psi(t) dt, & \text{if } x \geq \alpha; \\ +\infty, & \text{otherwise.} \end{cases} \quad (8.11)$$

Then  $\phi$  is convex. If  $\psi$  is strictly increasing, then  $\phi$  is strictly convex.

*Proof.* By Proposition 8.12,  $\varphi: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto \int_{\alpha}^x \psi(t) dt$  is convex since  $\varphi' = \psi$ . Hence, the convexity of  $\phi = \varphi + \iota_{[\alpha, +\infty[}$  follows from Proposition 8.4. The variant in which  $\psi$  is strictly convex is proved similarly.  $\square$

## 8.2 Convexity–Preserving Operations

In this section we describe some basic operations that preserve convexity. Further results will be provided in Proposition 12.11 and Proposition 12.34(ii) (see also Exercise 8.13).

**Proposition 8.14** *Let  $(f_i)_{i \in I}$  be a family of convex functions from  $\mathcal{H}$  to  $[-\infty, +\infty]$ . Then  $\sup_{i \in I} f_i$  is convex.*

*Proof.* By Lemma 1.6(i),  $\text{epi} \left( \sup_{i \in I} f_i \right) = \bigcap_{i \in I} \text{epi} f_i$ , which is convex as an intersection of convex sets by Example 3.2(iv).  $\square$

**Proposition 8.15** *The set of convex functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$  is closed under addition and multiplication by strictly positive real numbers.*

*Proof.* This is an easy consequence of Proposition 8.4.  $\square$

**Proposition 8.16** *Let  $(f_n)_{n \in \mathbb{N}}$  be a sequence of convex functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$  such that  $(f_n)_{n \in \mathbb{N}}$  is pointwise convergent. Then  $\lim f_n$  is convex.*

*Proof.* Set  $f = \lim f_n$ , and take  $x$  and  $y$  in  $\text{dom } f$ , and  $\alpha \in ]0, 1[$ . By Corollary 8.10,  $(\forall n \in \mathbb{N}) f_n(\alpha x + (1 - \alpha)y) \leq \alpha f_n(x) + (1 - \alpha)f_n(y)$ . Passing to the limit, we obtain  $f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y)$ , and using the same result, we conclude that  $f$  is convex.  $\square$

**Example 8.17** Let  $(z_n)_{n \in \mathbb{N}}$  be a bounded sequence in  $\mathcal{H}$ . Then  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \overline{\lim} \|x - z_n\|$  is convex and Lipschitz continuous with constant 1.

*Proof.* Set  $(\forall n \in \mathbb{N}) f_n = \sup \{ \|\cdot - z_k\| \mid k \in \mathbb{N}, k \geq n \}$ . Then it follows from Example 8.7 and Proposition 8.14 that each  $f_n$  is convex. In addition,  $(f_n)_{n \in \mathbb{N}}$  decreases pointwise to  $f = \inf_{n \in \mathbb{N}} f_n$ . Hence  $f$  is convex by Proposition 8.16. Now take  $x$  and  $y$  in  $\mathcal{H}$ . Then  $(\forall n \in \mathbb{N}) \|x - z_n\| \leq \|x - y\| + \|y - z_n\|$ . Thus

$$\overline{\lim} \|x - z_n\| \leq \|x - y\| + \overline{\lim} \|y - z_n\| \quad (8.12)$$

and hence  $f(x) - f(y) \leq \|x - y\|$ . Analogously,  $f(y) - f(x) \leq \|y - x\|$ . Altogether,  $|f(x) - f(y)| \leq \|x - y\|$ .  $\square$

**Proposition 8.18** *Let  $\mathcal{K}$  be a real Hilbert space, let  $L: \mathcal{H} \rightarrow \mathcal{K}$  be affine, and let  $f: \mathcal{K} \rightarrow ] -\infty, +\infty]$  be convex. Then  $f \circ L$  is convex.*

*Proof.* This is an immediate consequence of Proposition 8.4.  $\square$

**Proposition 8.19** *Let  $f: \mathcal{H} \rightarrow ] -\infty, +\infty]$  and  $\phi: \mathbb{R} \rightarrow ] -\infty, +\infty]$  be convex. Set  $C = \text{conv}(\mathbb{R} \cap \text{ran } f)$  and extend  $\phi$  to  $\tilde{\phi}: ] -\infty, +\infty] \rightarrow ] -\infty, +\infty]$  by setting  $\tilde{\phi}(+\infty) = +\infty$ . Suppose that  $C \subset \text{dom } \phi$  and that  $\phi$  is increasing on  $C$ . Then  $\tilde{\phi} \circ f$  is convex.*

*Proof.* Note that  $\text{dom}(\tilde{\phi} \circ f) = f^{-1}(\text{dom } \phi) \subset \text{dom } f$  and that  $\tilde{\phi} \circ f$  coincides with  $\phi \circ f$  on  $\text{dom}(\tilde{\phi} \circ f)$ . Take  $x$  and  $y$  in  $\text{dom}(\tilde{\phi} \circ f)$ , and  $\alpha \in ]0, 1[$ . Using the convexity of  $f$ , we obtain  $f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y) < +\infty$ . Since  $\phi$  is increasing and convex on  $C$ , we deduce that  $(\phi \circ f)(\alpha x + (1 - \alpha)y) \leq \phi(\alpha f(x) + (1 - \alpha)f(y)) \leq \alpha(\phi \circ f)(x) + (1 - \alpha)(\phi \circ f)(y)$ . Hence, the result follows from Proposition 8.4.  $\square$

**Example 8.20** Let  $\psi: \mathbb{R} \rightarrow \mathbb{R}$  be a continuous increasing function such that  $\psi(0) \geq 0$ , and set

$$f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \int_0^{\|x\|} \psi(t) dt. \quad (8.13)$$

Then the following hold:

- (i)  $f$  is convex.
- (ii) Suppose that  $\psi$  is strictly increasing. Then  $f$  is strictly convex.

*Proof.* Let  $\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be the function obtained by setting  $\alpha = 0$  in (8.11). Then, since  $\psi \geq 0$  on  $\mathbb{R}_+$ ,  $\phi$  is increasing on  $\text{dom } \phi = \mathbb{R}_+$  and  $f = \phi \circ \|\cdot\|$ .

(i): It follows from Example 8.13 that  $\phi$  is convex, and in turn, from Example 8.7 and Proposition 8.19 that  $f$  is convex.

(ii): Let  $x$  and  $y$  be two distinct points in  $\mathcal{H}$  and let  $\alpha \in ]0, 1[$ . First, assume that  $\|x\| \neq \|y\|$ . Since  $\phi$  is increasing and, by Example 8.13, strictly convex, we have  $f(\alpha x + (1 - \alpha)y) = \phi(\|\alpha x + (1 - \alpha)y\|) \leq \phi(\alpha\|x\| + (1 - \alpha)\|y\|) < \alpha\phi(\|x\|) + (1 - \alpha)\phi(\|y\|) = \alpha f(x) + (1 - \alpha)f(y)$ . Now assume that  $\|x\| = \|y\|$ . Then Corollary 2.14 asserts that  $\|\alpha x + (1 - \alpha)y\|^2 = \alpha\|x\|^2 + (1 - \alpha)\|y\|^2 - \alpha(1 - \alpha)\|x - y\|^2 = \|x\|^2 - \alpha(1 - \alpha)\|x - y\|^2 < \|x\|^2$ . Hence, since  $\phi$  is strictly increasing,

$$\begin{aligned} f(\alpha x + (1 - \alpha)y) &= \phi(\|\alpha x + (1 - \alpha)y\|) \\ &< \phi(\|x\|) \\ &= \alpha\phi(\|x\|) + (1 - \alpha)\phi(\|y\|) \\ &= \alpha f(x) + (1 - \alpha)f(y), \end{aligned} \quad (8.14)$$

which completes the proof.  $\square$

**Example 8.21** Let  $p \in ]1, +\infty[$ . Then  $\|\cdot\|^p$  is strictly convex and

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad \|x + y\|^p \leq 2^{p-1}(\|x\|^p + \|y\|^p). \quad (8.15)$$

*Proof.* Set  $\psi(t) = t$  if  $t < 0$ , and  $\psi(t) = pt^{p-1}$  if  $t \geq 0$ . Then  $\psi$  is continuous, strictly increasing, and  $\psi(0) = 0$ . Hence, we deduce from Example 8.20(ii) that  $\|\cdot\|^p$  is strictly convex. As a result, for every  $x$  and  $y$  in  $\mathcal{H}$ ,  $\|(x+y)/2\|^p \leq (\|x\|^p + \|y\|^p)/2$  and (8.15) follows.  $\square$

The next proposition describes a convex integral function in the setting of Example 2.5.

**Proposition 8.22** Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space and let  $(\mathbf{H}, \langle \cdot | \cdot \rangle_{\mathbf{H}})$  be a real Hilbert space. Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); \mathbf{H})$ , and let  $\varphi: \mathbf{H} \rightarrow ]-\infty, +\infty]$  be a measurable convex function such that

$$(\forall x \in \mathcal{H})(\exists \varrho \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R})) \quad \varphi \circ x \geq \varrho \quad \mu\text{-a.e.} \quad (8.16)$$



Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \int_{\Omega} \varphi(x(\omega)) \mu(d\omega). \quad (8.17)$$

Then  $\text{dom } f = \{x \in \mathcal{H} \mid \varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R})\}$  and  $f$  is convex.

*Proof.* For every  $x \in \mathcal{H}$ ,  $\varphi \circ x$  is measurable and, in view of (8.16),  $\int_{\Omega} (\varphi \circ x) d\mu$  is well defined and it never takes on the value  $-\infty$ . Hence,  $f$  is well defined and we obtain the above expression for its domain. Now take  $x$  and  $y$  in  $\mathcal{H}$  and  $\alpha \in ]0, 1[$ . It follows from Corollary 8.10 that, for  $\mu$ -almost every  $\omega \in \Omega$ , we have  $\varphi(\alpha x(\omega) + (1 - \alpha)y(\omega)) \leq \alpha \varphi(x(\omega)) + (1 - \alpha)\varphi(y(\omega))$ . Upon integrating these inequalities over  $\Omega$  with respect to  $\mu$ , we obtain  $f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y)$ , hence the convexity of  $f$ .  $\square$

**Proposition 8.23** *Let  $\varphi: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and let*

$$f: \mathbb{R} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi, x) \mapsto \begin{cases} \xi \varphi(x/\xi), & \text{if } \xi > 0; \\ +\infty, & \text{otherwise,} \end{cases} \quad (8.18)$$

*be the perspective function of  $\varphi$ . Then  $f$  is convex.*

*Proof.* Set  $C = \{1\} \times \text{epi } \varphi$ . It follows from the convexity of  $\varphi$  that  $C$  is a convex subset of  $\mathbb{R} \times \mathcal{H} \times \mathbb{R}$ . Furthermore,

$$\begin{aligned} \text{epi } f &= \{(\xi, x, \zeta) \in \mathbb{R}_{++} \times \mathcal{H} \times \mathbb{R} \mid \xi \varphi(x/\xi) \leq \zeta\} \\ &= \{(\xi, x, \zeta) \in \mathbb{R}_{++} \times \mathcal{H} \times \mathbb{R} \mid (x/\xi, \zeta/\xi) \in \text{epi } \varphi\} \\ &= \{\xi(1, y, \eta) \mid \xi \in \mathbb{R}_{++} \text{ and } (y, \eta) \in \text{epi } \varphi\} \\ &= \mathbb{R}_{++} C \\ &= \text{cone } C. \end{aligned} \quad (8.19)$$

In view of Proposition 6.2(iii),  $\text{epi } f$  is therefore a convex set.  $\square$

**Example 8.24** Let  $C$  be a convex subset of  $\mathcal{H}$  and set

$$f: \mathbb{R} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi, x) \mapsto \begin{cases} \xi, & \text{if } \xi > 0 \text{ and } x \in \xi C; \\ +\infty, & \text{otherwise.} \end{cases} \quad (8.20)$$

Then  $f$  is convex.

*Proof.* Apply Proposition 8.23 to  $1 + \iota_C$ .  $\square$

**Proposition 8.25** *Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered family of real Hilbert spaces, set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ , and let, for every  $i$  in  $I$ ,  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$  be convex. Suppose that  $I$  is finite or that  $\inf_{i \in I} f_i \geq 0$ . Then  $\bigoplus_{i \in I} f_i$  is convex.*

*Proof.* This is an immediate consequence of Proposition 8.4.  $\square$

**Proposition 8.26** *Let  $\mathcal{K}$  be a real Hilbert space and let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$  be convex. Then the marginal function*

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty] : x \mapsto \inf F(x, \mathcal{K}) \quad (8.21)$$

*is convex.*

*Proof.* Take  $x_1$  and  $x_2$  in  $\text{dom } f$ , and  $\alpha \in ]0, 1[$ . Furthermore, let  $\xi_1 \in ]f(x_1), +\infty[$  and  $\xi_2 \in ]f(x_2), +\infty[$ . Then there exist  $y_1$  and  $y_2$  in  $\mathcal{K}$  such that  $F(x_1, y_1) < \xi_1$  and  $F(x_2, y_2) < \xi_2$ . In turn, the convexity of  $F$  gives

$$\begin{aligned} f(\alpha x_1 + (1 - \alpha)x_2) &\leq F(\alpha x_1 + (1 - \alpha)x_2, \alpha y_1 + (1 - \alpha)y_2) \\ &\leq \alpha F(x_1, y_1) + (1 - \alpha)F(x_2, y_2) \\ &< \alpha \xi_1 + (1 - \alpha)\xi_2. \end{aligned} \quad (8.22)$$

Thus, letting  $\xi_1 \downarrow f(x_1)$  and  $\xi_2 \downarrow f(x_2)$ , we obtain  $f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2)$ .  $\square$

**Example 8.27** Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $0 \in C$ . The *Minkowski gauge* of  $C$ , i.e., the function

$$m_C: \mathcal{H} \rightarrow ]-\infty, +\infty] : x \mapsto \inf \{ \xi \in \mathbb{R}_{++} \mid x \in \xi C \}, \quad (8.23)$$

is convex. Furthermore,  $m_C(0) = 0$ ,

$$(\forall x \in \mathcal{H})(\forall \lambda \in \mathbb{R}_{++}) \quad m_C(\lambda x) = \lambda m_C(x) = m_{(1/\lambda)C}(x), \quad (8.24)$$

and

$$(\forall x \in \text{dom } m_C)(\forall \lambda \in ]m_C(x), +\infty[) \quad x \in \lambda C. \quad (8.25)$$

*Proof.* The convexity of  $m_C$  follows from Example 8.24 and Proposition 8.26. Since  $0 \in C$ , it is clear that  $m_C(0) = 0$ . Now take  $x \in \mathcal{H}$  and  $\lambda \in \mathbb{R}_{++}$ . Then

$$\begin{aligned} \{ \xi \in \mathbb{R}_{++} \mid \lambda x \in \xi C \} &= \lambda \{ \xi/\lambda \in \mathbb{R}_{++} \mid x \in (\xi/\lambda)C \} \\ &= \{ \xi \in \mathbb{R}_{++} \mid x \in \xi((1/\lambda)C) \}, \end{aligned} \quad (8.26)$$

and (8.24) follows by taking the infimum. Finally, suppose that  $x \in \text{dom } m_C$ ,  $\lambda > m_C(x)$ , and  $x \notin \lambda C$ . Now let  $\mu \in ]0, \lambda]$ . Since  $0 \in C$  and  $C$  is convex, we have  $\mu C \subset \lambda C$  and therefore  $x \notin \mu C$ . Consequently,  $m_C(x) \geq \lambda$ , which is a contradiction.  $\square$

### 8.3 Topological Properties

We start with a technical fact.

**Proposition 8.28** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x_0 \in \text{dom } f$ , and let  $\rho \in \mathbb{R}_{++}$ . Then the following hold:*

(i) *Suppose that  $\eta = \sup f(B(x_0; \rho)) < +\infty$  and let  $\alpha \in ]0, 1[$ . Then*

$$(\forall x \in B(x_0; \alpha\rho)) \quad |f(x) - f(x_0)| \leq \alpha(\eta - f(x_0)). \quad (8.27)$$

(ii) *Suppose that  $\delta = \text{diam } f(B(x_0; 2\rho)) < +\infty$ . Then  $f$  is Lipschitz continuous relative to  $B(x_0; \rho)$  with constant  $\delta/\rho$ .*

*Proof.* (i): Take  $x \in B(x_0; \alpha\rho)$ . The convexity of  $f$  yields

$$\begin{aligned} f(x) - f(x_0) &= f((1 - \alpha)x_0 + \alpha(x - (1 - \alpha)x_0)/\alpha) - f(x_0) \\ &\leq \alpha(f(x_0 + (x - x_0)/\alpha) - f(x_0)) \\ &\leq \alpha(\eta - f(x_0)). \end{aligned} \quad (8.28)$$

Likewise,

$$\begin{aligned} f(x_0) - f(x) &= f\left(\frac{x}{1 + \alpha} + \frac{\alpha}{1 + \alpha} \frac{(1 + \alpha)x_0 - x}{\alpha}\right) - f(x) \\ &\leq \frac{\alpha}{1 + \alpha}(f(x_0 + (x_0 - x)/\alpha) - f(x)) \\ &\leq \frac{\alpha}{1 + \alpha}((\eta - f(x_0)) + (f(x_0) - f(x))), \end{aligned} \quad (8.29)$$

which after rearranging implies that  $f(x_0) - f(x) \leq \alpha(\eta - f(x_0))$ . Altogether,  $|f(x) - f(x_0)| \leq \alpha(\eta - f(x_0))$ .

(ii): Take distinct points  $x$  and  $y$  in  $B(x_0; \rho)$  and set

$$z = x + \left(\frac{1}{\alpha} - 1\right)(x - y), \quad \text{where } \alpha = \frac{\|x - y\|}{\|x - y\| + \rho} < \frac{\|x - y\|}{\rho}. \quad (8.30)$$

Then  $x = \alpha z + (1 - \alpha)y$  and  $\|z - x_0\| \leq \|z - x\| + \|x - x_0\| = \rho + \|x - x_0\| \leq 2\rho$ . Therefore,  $y$  and  $z$  belong to  $B(x_0; 2\rho)$  and hence, by convexity of  $f$ ,

$$\begin{aligned} f(x) &= f(\alpha z + (1 - \alpha)y) \\ &\leq f(y) + \alpha(f(z) - f(y)) \\ &\leq f(y) + \alpha\delta \\ &\leq f(y) + (\delta/\rho)\|x - y\|. \end{aligned} \quad (8.31)$$

Thus  $f(x) - f(y) \leq (\delta/\rho)\|x - y\|$ . Interchanging the roles of  $x$  and  $y$ , we conclude that  $|f(x) - f(y)| \leq (\delta/\rho)\|x - y\|$ .  $\square$

The following theorem captures the main continuity properties of convex functions.

**Theorem 8.29** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex and let  $x_0 \in \text{dom } f$ . Then the following are equivalent:*

- (i)  $f$  is locally Lipschitz continuous near  $x_0$ .
- (ii)  $f$  is continuous at  $x_0$ .
- (iii)  $f$  is bounded on a neighborhood of  $x_0$ .
- (iv)  $f$  is bounded above on a neighborhood of  $x_0$ .

Moreover, if one of these conditions holds, then  $f$  is locally Lipschitz continuous on  $\text{int dom } f$ .

*Proof.* (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (iv): Clear.

(iv) $\Rightarrow$ (ii): Take  $\rho \in \mathbb{R}_{++}$  such that  $\eta = \sup f(B(x_0; \rho)) < +\infty$ ,  $\alpha \in ]0, 1[$ , and  $x \in B(x_0; \alpha\rho)$ . Proposition 8.28(i) implies that  $|f(x) - f(x_0)| \leq \alpha(\eta - f(x_0))$ . Therefore,  $f$  is continuous at  $x_0$ .

(iii) $\Rightarrow$ (i): An immediate consequence of Proposition 8.28(ii).

We have shown that items (i)–(iv) are equivalent. Now assume that (iv) is satisfied, say  $\eta = \sup f(B(x_0; \rho)) < +\infty$  for some  $\rho \in \mathbb{R}_{++}$ . Then  $f$  is locally Lipschitz continuous near  $x_0$ . Take  $x \in \text{int dom } f \setminus \{x_0\}$  and  $\gamma \in \mathbb{R}_{++}$  such that  $B(x; \gamma) \subset \text{dom } f$ . Now set

$$y = x_0 + \frac{1}{1 - \alpha}(x - x_0), \quad \text{where } \alpha = \frac{\gamma}{\gamma + \|x - x_0\|} \in ]0, 1[. \quad (8.32)$$

Then  $y \in B(x; \gamma)$ . Now take  $z \in B(x; \alpha\rho)$  and set  $w = x_0 + (z - x)/\alpha = (z - (1 - \alpha)y)/\alpha$ . Then  $w \in B(x_0; \rho)$  and  $z = \alpha w + (1 - \alpha)y$ . Consequently,

$$f(z) \leq \alpha f(w) + (1 - \alpha)f(y) \leq \alpha\eta + (1 - \alpha)f(y), \quad (8.33)$$

and  $f$  is therefore bounded above on  $B(x; \alpha\rho)$ . By the already established equivalence between (i) and (iv) (applied to the point  $x$ ), we conclude that  $f$  is locally Lipschitz continuous near  $x$ .  $\square$

**Corollary 8.30** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and suppose that one of the following holds:*

- (i)  $f$  is bounded above on some neighborhood.
- (ii)  $f$  is lower semicontinuous.
- (iii)  $\mathcal{H}$  is finite-dimensional.

Then  $\text{cont } f = \text{int dom } f$ .

*Proof.* Since the inclusion  $\text{cont } f \subset \text{int dom } f$  always holds, we assume that  $\text{int dom } f \neq \emptyset$ .

(i): Clear from Theorem 8.29.

(ii): Define a sequence  $(C_n)_{n \in \mathbb{N}}$  of closed convex subsets of  $\mathcal{H}$  by  $(\forall n \in \mathbb{N})$   $C_n = \text{lev}_{\leq n} f$ . Then  $\text{dom } f = \bigcup_{n \in \mathbb{N}} C_n$  and, by Lemma 1.43(i),  $\emptyset \neq \overline{\text{int dom } f} = \overline{\text{int } \bigcup_{n \in \mathbb{N}} C_n} = \bigcup_{n \in \mathbb{N}} \overline{\text{int } C_n}$ . Hence there exists  $n \in \mathbb{N}$  such that  $\text{int } C_n \neq \emptyset$ , say  $B(x; \rho) \subset C_n$ , where  $x \in C_n$  and  $\rho \in \mathbb{R}_{++}$ . Thus  $\sup f(B(x; \rho)) \leq n$  and we apply (i).

(iii): Let  $x \in \text{int dom } f$ . Since  $\mathcal{H}$  is finite-dimensional, there exist a finite family  $(y_i)_{i \in I}$  in  $\text{dom } f$  and  $\rho \in \mathbb{R}_{++}$  such that  $B(x; \rho) \subset \text{conv}\{y_i\}_{i \in I}$

(see Exercise 3.18). Consequently, by Proposition 8.9,  $\sup f(B(x; \rho)) \leq \sup f(\text{conv}\{y_i\}_{i \in I}) \leq \sup \text{conv}\{f(y_i)\}_{i \in I} = \max_{i \in I} f(y_i) < +\infty$ . The conclusion follows from (i).  $\square$

**Corollary 8.31** *Suppose that  $\mathcal{H}$  is finite-dimensional and let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be convex. Then  $f$  is continuous.*

**Corollary 8.32** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $C$  be a nonempty closed bounded subset of  $\text{ri dom } f$ . Then  $f$  is Lipschitz continuous relative to  $C$ .*

*Proof.* Let  $z \in \text{dom } f$ . We work in the Euclidean space  $\text{span}(z - \text{dom } f)$  and therefore assume, without loss of generality, that  $C \subset \text{int dom } f$ . By Corollary 8.30(iii),  $f$  is continuous on  $\text{int dom } f$ . Hence, by Theorem 8.29, for every  $x \in C$ , there exists an open ball  $B_x$  of center  $x$  such that  $f$  is  $\beta_x$ -Lipschitz continuous relative to  $B_x$  for some  $\beta_x \in \mathbb{R}_{++}$ . Thus,  $C \subset \bigcup_{x \in C} B_x$ , and since  $C$  is compact, it follows that there exists a finite subset  $D$  of  $C$  such that  $C \subset \bigcup_{x \in D} B_x$ . We therefore conclude that  $f$  is Lipschitz continuous relative to  $C$  with constant  $\max_{x \in D} \beta_x$ .  $\square$

**Example 8.33** Suppose that  $\mathcal{H}$  is infinite-dimensional and let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be a linear functional that is everywhere discontinuous; see Example 2.20 for a concrete construction. Set  $C = \text{lev}_{\leq 1} f$  and  $H = \{x \in \mathcal{H} \mid f(x) = 0\}$ . Then the following hold:

- (i)  $f$  and  $-f$  are convex,  $\text{int dom } f = \text{int dom }(-f) = \mathcal{H}$ , and  $\text{cont } f = \text{cont }(-f) = \emptyset$ .
- (ii)  $f$  is neither lower semicontinuous nor upper semicontinuous.
- (iii)  $C$  is convex,  $0 \in \text{core } C$ , and  $\text{int } C = \emptyset$ .
- (iv)  $f$  is unbounded above and below on every nonempty open subset of  $\mathcal{H}$ .
- (v)  $H$  is a hyperplane that is dense in  $\mathcal{H}$ .

*Proof.* (i): Since  $f$  is linear and  $\text{dom } f = \mathcal{H}$ , it follows that  $f$  and  $-f$  are convex with  $\text{int dom } f = \text{int dom }(-f) = \text{int } \mathcal{H} = \mathcal{H}$ . The assumption that  $f$  is everywhere discontinuous implies that  $\text{cont } f = \text{cont }(-f) = \emptyset$ .

(ii): If  $f$  were lower semicontinuous, then (i) and Corollary 8.30(ii) would imply that  $\emptyset = \text{cont } f = \text{int dom } f = \mathcal{H}$ , which is absurd. Hence  $f$  is not lower semicontinuous. Analogously, we deduce that  $-f$  is not lower semicontinuous, i.e., that  $f$  is not upper semicontinuous.

(iii): The convexity of  $C$  follows from (i) and Corollary 8.5. Assume that  $x_0 \in \text{int } C$ . Since  $f$  is bounded above on  $C$  by 1, it follows that  $f$  is bounded above on a neighborhood of  $x_0$ . Then Corollary 8.30(i) implies that  $\text{cont } f = \text{int dom } f$ , which contradicts (i). Hence  $\text{int } C = \emptyset$ . Now take  $x \in \mathcal{H}$ . If  $x \in C$ , then  $[0, x] \subset C$  by linearity of  $f$ . Likewise, if  $x \in \mathcal{H} \setminus C$ , then  $[0, x/f(x)] \subset C$ . Thus  $\text{cone } C = \mathcal{H}$  and therefore  $0 \in \text{core } C$ .

(iv)&(v): Take  $x_0 \in \mathcal{H}$  and  $\varepsilon \in \mathbb{R}_{++}$ . Corollary 8.30(i) and item (i) imply that  $f$  is unbounded both above and below on  $B(x_0; \varepsilon)$ . In particular, there

exist points  $y_-$  and  $y_+$  in  $B(x_0; \varepsilon)$  such that  $f(y_-) < 0 < f(y_+)$ . Now set  $y_0 = (f(y_+)y_- - f(y_-)y_+)/ (f(y_+) - f(y_-))$ . Then, by linearity of  $f$ ,  $y_0 \in [y_-, y_+] \subset B(x_0; \varepsilon)$  and  $f(y_0) = 0$ .  $\square$

**Example 8.34** Suppose that  $C$  is a convex subset of  $\mathcal{H}$  such that  $0 \in \text{int } C$ . Then  $\text{dom } m_C = \mathcal{H}$  and  $m_C$  is continuous.

*Proof.* Let  $x \in \mathcal{H}$ . Then there exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $\varepsilon x \in C$ , which implies that  $m_C(x) \leq 1/\varepsilon$ . Thus  $\text{dom } m_C = \mathcal{H}$ . Since  $C$  is a neighborhood of 0 and  $m_C(C) \subset [0, 1]$ , Corollary 8.30(i) implies that  $m_C$  is continuous.  $\square$

**Example 8.35 (Huber's function)** Let  $\rho \in \mathbb{R}_{++}$  and set

$$f: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto \begin{cases} \frac{x^2}{2}, & \text{if } |x| \leq \rho; \\ \rho|x| - \frac{\rho^2}{2}, & \text{otherwise.} \end{cases} \quad (8.34)$$

Then  $f$  is continuous and convex.

*Proof.* The convexity of  $f$  follows from Proposition 8.12(i) and its continuity from Corollary 8.31.  $\square$

**Proposition 8.36** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Suppose that  $f$  is bounded above on a neighborhood of  $x \in \text{dom } f$ . Then  $\text{int epi } f \neq \emptyset$ .

*Proof.* Using Theorem 8.29, we obtain  $\delta \in \mathbb{R}_{++}$  and  $\beta \in \mathbb{R}_+$  such that

$$(\forall y \in B(x; \delta)) \quad |f(x) - f(y)| \leq \beta \|x - y\| \leq \beta \delta. \quad (8.35)$$

Now fix  $\rho \in ]2\beta\delta, +\infty[$ , set  $\gamma = \min\{\delta, \rho/2\} > 0$ , and take  $(y, \eta) \in \mathcal{H} \times \mathbb{R}$  such that

$$\|(y, \eta) - (x, f(x) + \rho)\|^2 \leq \gamma^2. \quad (8.36)$$

Then  $\|y - x\| \leq \gamma \leq \delta$  and  $|\eta - (f(x) + \rho)| \leq \gamma \leq \rho/2$ . It follows from (8.35) that  $y \in \text{dom } f$  and

$$f(y) < f(x) + \rho/2 = f(x) + \rho - \rho/2 \leq f(x) + \rho - \gamma \leq \eta. \quad (8.37)$$

Thus  $(y, \eta) \in \text{epi } f$  and hence  $B((x, f(x) + \rho); \gamma) \subset \text{epi } f$ . We conclude that  $\text{int epi } f \neq \emptyset$ .  $\square$

**Proposition 8.37** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ . Then the following hold:

- (i) Suppose that  $f$  is upper semicontinuous at a point  $x_0 \in \mathcal{H}$  such that  $f(x_0) < 0$ . Then  $x_0 \in \text{int lev}_{\leq 0} f$ .
- (ii) Suppose that  $f$  is convex and that there exists a point  $x_0 \in \mathcal{H}$  such that  $f(x_0) < 0$ . Then  $\text{int lev}_{\leq 0} f \subset \text{lev}_{< 0} f$ .

*Proof.* (i): If  $x_0 \notin \text{int lev}_{\leq 0} f$ , then there exists a sequence  $(y_n)_{n \in \mathbb{N}}$  in  $\mathcal{H} \setminus \text{lev}_{\leq 0} f = \text{lev}_{> 0} f$  such that  $y_n \rightarrow x_0$ . In turn, the upper semicontinuity of  $f$  at  $x_0$  yields  $f(x_0) \geq \overline{\lim} f(y_n) \geq 0 > f(x_0)$ , which is absurd.

(ii): Fix  $x \in \text{int lev}_{\leq 0} f$ . Then  $(\exists \rho \in \mathbb{R}_{++}) B(x; \rho) \subset \text{lev}_{\leq 0} f$ . We must show that  $f(x) < 0$ . We assume that  $x \neq x_0$  since, if  $x = x_0$ , there is nothing to prove. Let  $\delta \in ]0, \rho/\|x - x_0\|]$ . Set  $y = x_0 + (1 + \delta)(x - x_0) = x + \delta(x - x_0) \in \text{lev}_{\leq 0} f$  and observe that  $x = \alpha x_0 + (1 - \alpha)y$ , where  $\alpha = \delta/(1 + \delta) \in ]0, 1[$ . Hence  $f(x) \leq \alpha f(x_0) + (1 - \alpha)f(y) < 0$ .  $\square$

**Corollary 8.38** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a convex function such that  $\text{lev}_{< 0} f \neq \emptyset$ . Suppose that one of the following holds:*

- (i)  *$f$  is upper semicontinuous on  $\text{lev}_{< 0} f$ .*
- (ii)  *$f$  is lower semicontinuous and  $\text{dom } f$  is open.*
- (iii)  *$\mathcal{H}$  is finite-dimensional and  $\text{dom } f$  is open.*

*Then  $\text{int lev}_{\leq 0} f = \text{lev}_{< 0} f$ .*

*Proof.* (i): This follows from Proposition 8.37.

(ii)&(iii): By Corollary 8.30,  $f$  is continuous on  $\text{int dom } f = \text{dom } f$  and hence upper semicontinuous on  $\text{lev}_{< 0} f$ . Thus, the claim follows from (i).  $\square$

## Exercises

**Exercise 8.1** Let  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be convex, let  $x_0, x_\beta$ , and  $x_1$  be points in  $\text{dom } f$  such that  $x_0 < x_1$  and  $x_\beta = (1 - \beta)x_0 + \beta x_1$ , where  $\beta \in ]0, 1[$ , and suppose that the points  $(x_0, f(x_0))$ ,  $(x_\beta, f(x_\beta))$ , and  $(x_1, f(x_1))$  lie on a line. Show that  $(\forall \alpha \in [0, 1]) f((1 - \alpha)x_0 + \alpha x_1) = (1 - \alpha)f(x_0) + \alpha f(x_1)$ .

**Exercise 8.2** Provide examples of a function  $f: \mathbb{R} \rightarrow \mathbb{R}$  and a nonempty set  $C \subset \mathbb{R}$  illustrating each of the following:

- (i)  $f$  is not convex,  $C$  is convex, and  $f$  is convex on  $C$ .
- (ii)  $f$  is not convex,  $C$  is not convex, and  $f$  is convex on  $C$ .

**Exercise 8.3** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper function. Show that  $f$  is convex if and only if  $f$  is convex on  $\text{dom } f$ .

**Exercise 8.4** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper function, let  $C$  be a nonempty subset of  $\text{dom } f$ , and suppose that  $f + \iota_C$  is convex. Show that  $C$  is convex and that  $f$  is convex on  $C$ .

**Exercise 8.5** Provide an example of a proper function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and a nonempty subset  $C$  of  $\text{dom } f$  such that  $f$  is convex on  $C$  but  $f + \iota_C$  is not convex.

**Exercise 8.6** Provide an example of a function  $f: \mathcal{H} \rightarrow \mathbb{R}$  and a subset  $C$  of  $\text{dom } f$  such that  $f$  is convex on  $C$  but there exist distinct points  $x_1, x_2$ , and  $x_3$  in  $C$  and real numbers  $\alpha_1, \alpha_2$ , and  $\alpha_3$  in  $]0, 1[$  such that  $f(\sum_{i=1}^3 \alpha_i x_i) > \sum_{i=1}^3 \alpha_i f(x_i)$ .

**Exercise 8.7** Let  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be proper and convex. Show that  $f$  is strictly convex if and only if it is strictly convex on  $\text{int dom } f$ .

**Exercise 8.8** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ . Show that  $f$  is convex if and only if

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \quad x \neq y \\ \Rightarrow (\forall z \in ]x, y[) \quad \frac{f(z) - f(x)}{\|z - x\|} + \frac{f(z) - f(y)}{\|z - y\|} \leq 0. \quad (8.38)$$

**Exercise 8.9** Let  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x_0 \in \text{dom } f$ , and let  $\alpha \in ]0, 1[$ . Set  $g: \text{dom } f \rightarrow \mathbb{R}: x \mapsto \alpha f(x) - f(x_0 + \alpha(x - x_0))$ . Show that  $g$  is decreasing on  $] -\infty, x_0] \cap \text{dom } f$  and increasing on  $[x_0, +\infty[ \cap \text{dom } f$ . In addition show that, if  $f$  is strictly convex, then  $g$  is strictly decreasing on  $] -\infty, x_0] \cap \text{dom } f$  and strictly increasing on  $[x_0, +\infty[ \cap \text{dom } f$ .

**Exercise 8.10** Show that the converse of Corollary 8.5 is false by providing an example of a function that is not convex and the lower level sets of which are all convex.

**Exercise 8.11 (arithmetic mean–geometric mean inequality)** Let  $(x_i)_{1 \leq i \leq m}$  be a finite family in  $\mathbb{R}_+$ . Show that

$$\sqrt[m]{x_1 \cdots x_m} \leq \frac{x_1 + \cdots + x_m}{m}, \quad (8.39)$$

and that equality occurs in (8.39) if and only if  $x_1 = \cdots = x_m$ .

**Exercise 8.12** Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces, and let, for every  $i$  in  $I$ ,  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$  be proper. Show that  $\bigoplus_{i \in I} f_i$  is convex if and only if the functions  $(f_i)_{i \in I}$  are. Furthermore, prove the corresponding statement for strict convexity.

**Exercise 8.13** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$ , and set

$$h: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \inf_{y \in \mathcal{H}} \max\{f(y), g(x - y)\}. \quad (8.40)$$

Prove the following:

- (i)  $\text{dom } h = \text{dom } f + \text{dom } g$ .
- (ii)  $\inf h(\mathcal{H}) = \max\{\inf f(\mathcal{H}), \inf g(\mathcal{H})\}$ .
- (iii)  $(\forall \eta \in \mathbb{R}) \quad \text{lev}_{h < \eta} = \text{lev}_{< \eta} f + \text{lev}_{< \eta} g$ .
- (iv)  $h$  is convex if  $f$  and  $g$  are.



**Exercise 8.14** Determine the Minkowski gauge of the closed unit ball.

**Exercise 8.15** Set  $C = \{(\xi_1, \xi_2) \in \mathbb{R}^2 \mid \max\{|\xi_1|, |\xi_2|\} \leq 1\}$ . Show that  $m_C: \mathbb{R}^2 \rightarrow \mathbb{R}: (\xi_1, \xi_2) \mapsto \max\{|\xi_1|, |\xi_2|\}$  and that  $m_{(1/2)C} = 2m_C$ .

**Exercise 8.16** Let  $C$  and  $D$  be convex subsets of  $\mathcal{H}$  such that  $0 \in C \subset D$ . Show that  $m_C \geq m_D$ .

**Exercise 8.17** Set  $C = \{(\xi_1, \xi_2) \in \mathbb{R}^2 \mid \max\{|\xi_1|, |\xi_2|\} \leq 1\}$ ,  $D = B(0; 1)$ , and  $Q = \frac{1}{2}C + \frac{1}{2}D$ . Show that  $0 \in \text{int } Q$ , that  $Q$  is closed and convex, that  $m_Q: \mathbb{R}^2 \rightarrow \mathbb{R}$  is convex and continuous, that

$$(\forall (\xi_1, \xi_2) \in \mathbb{R}^2) \quad \max\{|\xi_1|, |\xi_2|\} \leq m_Q(\xi_1, \xi_2) \leq 2 \max\{|\xi_1|, |\xi_2|\}, \quad (8.41)$$

that

$$(\forall \xi_1 \in \mathbb{R}_{++})(\forall \xi_2 \in [0, \frac{1}{2}\xi_1[) \quad m_Q(\xi_1, \xi_2) = \xi_1, \quad (8.42)$$

and that

$$(\forall (\xi_1, \xi_2) \in \mathbb{R}_+^2)(\forall (\eta_1, \eta_2) \in \mathbb{R}_+^2) \quad \left. \begin{array}{l} \xi_1 \leq \eta_1 \\ \xi_2 \leq \eta_2 \end{array} \right\} \Rightarrow m_Q(\xi_1, \xi_2) \leq m_Q(\eta_1, \eta_2). \quad (8.43)$$

**Exercise 8.18** Suppose that  $\mathcal{H}$  is finite-dimensional and let  $C$  be a convex subset of  $\mathcal{H}$  such that  $0 \in C$ . Show that  $m_C$  is lower semicontinuous.

**Exercise 8.19** Let  $f$  and  $C$  be as in Example 8.33. Determine  $m_C$ .

**Exercise 8.20** Provide an example of a convex subset  $C$  of  $\mathcal{H}$  such that  $0 \in C$  but  $m_C$  is not lower semicontinuous.

**Exercise 8.21** Show that the conclusion of Corollary 8.38(i) fails if  $\text{dom } f$  is not open.



# Chapter 9

## Lower Semicontinuous Convex Functions

The theory of convex functions is most powerful in the presence of lower semicontinuity. A key property of lower semicontinuous convex functions is the existence of a continuous affine minorant, which we establish in this chapter by projecting onto the epigraph of the function.

### 9.1 Lower Semicontinuous Convex Functions

We start by observing that various types of lower semicontinuity coincide for convex functions.

**Theorem 9.1** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex. Then the following are equivalent:*

- (i)  *$f$  is weakly sequentially lower semicontinuous.*
- (ii)  *$f$  is sequentially lower semicontinuous.*
- (iii)  *$f$  is lower semicontinuous.*
- (iv)  *$f$  is weakly lower semicontinuous.*

*Proof.* The set  $\text{epi } f$  is convex by Definition 8.1. Hence, the equivalences follow from Lemma 1.24, Lemma 1.35, and Theorem 3.32.  $\square$

**Definition 9.2** The set of lower semicontinuous convex functions from  $\mathcal{H}$  to  $[-\infty, +\infty]$  is denoted by  $\Gamma(\mathcal{H})$ .

The set  $\Gamma(\mathcal{H})$  is closed under several important operations. For instance, it is straightforward to verify that  $\Gamma(\mathcal{H})$  is closed under multiplication by strictly positive real numbers.

**Proposition 9.3** *Let  $(f_i)_{i \in I}$  be a family in  $\Gamma(\mathcal{H})$ . Then  $\sup_{i \in I} f_i \in \Gamma(\mathcal{H})$ .*

*Proof.* Combine Lemma 1.26 and Proposition 8.14.  $\square$

**Corollary 9.4** *Let  $(f_i)_{i \in I}$  be a family in  $\Gamma(\mathcal{H})$ . Suppose that one of the following holds:*

- (i)  *$I$  is finite and  $-\infty \notin \bigcup_{i \in I} f_i(\mathcal{H})$ .*
- (ii)  *$\inf_{i \in I} f_i \geq 0$ .*

*Then  $\sum_{i \in I} f_i \in \Gamma(\mathcal{H})$ .*

*Proof.* (i): A consequence of Lemma 1.27 and Proposition 8.15.

(ii): Let  $\mathcal{I}$  be the class of nonempty finite subsets of  $I$  and set  $(\forall J \in \mathcal{I})$   $g_J = \sum_{i \in J} f_i$ . Then it follows from (i) that  $(\forall J \in \mathcal{I})$   $g_J \in \Gamma(\mathcal{H})$ . However, (2.4) yields  $\sum_{i \in I} f_i = \sup_{J \in \mathcal{I}} g_J$ . In view of Proposition 9.3, the proof is complete.  $\square$

**Proposition 9.5** *Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $f \in \Gamma(\mathcal{K})$ . Then  $f \circ L \in \Gamma(\mathcal{H})$ .*

*Proof.* This is a consequence of Proposition 8.18.  $\square$

**Proposition 9.6** *Let  $f \in \Gamma(\mathcal{H})$  and suppose that  $-\infty \in f(\mathcal{H})$ . Then  $f$  is nowhere real-valued, i.e.,  $f(\mathcal{H}) \subset \{-\infty, +\infty\}$ .*

*Proof.* Let  $x \in \mathcal{H}$  be such that  $f(x) = -\infty$ , let  $y \in \mathcal{H}$ , and let  $\alpha \in ]0, 1[$ . If  $f(y) \neq +\infty$ , then Proposition 8.4 yields  $f(\alpha x + (1 - \alpha)y) = -\infty$ . In turn, since  $f$  is lower semicontinuous,  $f(y) \leq \lim_{\alpha \downarrow 0} f(\alpha x + (1 - \alpha)y) = -\infty$ , i.e.,  $f(y) = -\infty$ .  $\square$

The function  $x \mapsto -\infty$  belongs to  $\Gamma(\mathcal{H})$ , which makes the following notion well defined.

**Definition 9.7** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then

$$\check{f} = \sup \{g \in \Gamma(\mathcal{H}) \mid g \leq f\} \quad (9.1)$$

is the *lower semicontinuous convex envelope* of  $f$ .

**Proposition 9.8** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then the following hold:*

- (i)  *$\check{f}$  is the largest lower semicontinuous convex function majorized by  $f$ .*
- (ii)  *$(\forall x \in \mathcal{H})$   $\check{f}(x) = \lim_{y \rightarrow x} \check{f}(y)$ .*
- (iii)  *$\text{epi } \check{f}$  is closed and convex.*
- (iv)  *$\text{conv dom } f \subset \text{dom } \check{f} \subset \overline{\text{conv}} \text{ dom } f$ .*

*Proof.* (i): This is a consequence of (9.1) and Proposition 9.3.

(ii): This follows from (i) and Lemma 1.31(iv).

(iii): Combine (i), Lemma 1.24, and Definition 8.1.

(iv): By (i),  $\check{f} \leq f$  and  $\check{f}$  is convex. Hence, Proposition 8.2 yields

$$\text{conv dom } f \subset \text{conv dom } \check{f} = \text{dom } \check{f}. \quad (9.2)$$

Now set  $C = \overline{\text{conv}} \text{ dom } f$  and

$$g: \mathcal{H} \rightarrow [-\infty, +\infty] : x \mapsto \begin{cases} \check{f}(x), & \text{if } x \in C; \\ +\infty, & \text{if } x \notin C. \end{cases} \quad (9.3)$$

Using (iii), we note that  $\text{epi } g = (\text{epi } \check{f}) \cap (C \times \mathbb{R})$  is closed and convex. It follows from Lemma 1.24 and Definition 8.1 that

$$g \in \Gamma(\mathcal{H}). \quad (9.4)$$

Now fix  $x \in \mathcal{H}$ . If  $x \in C$ , then  $g(x) = \check{f}(x) \leq f(x)$ ; otherwise,  $x \notin \text{dom } f \subset C$  and therefore  $g(x) = f(x) = +\infty$ . Altogether,  $g \leq f$  and, in view of (9.4), we obtain  $g \leq \check{f}$ . Thus,  $\text{dom } \check{f} \subset \text{dom } g \subset C = \overline{\text{conv}} \text{ dom } f$ .  $\square$

**Theorem 9.9** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $\text{epi } \check{f} = \overline{\text{conv}} \text{ epi } f$ .*

*Proof.* Set  $E = \overline{\text{conv}} \text{ epi } f$ . Since  $\check{f} \leq f$ , we have  $\text{epi } f \subset \text{epi } \check{f}$ . Hence  $E \subset \overline{\text{conv}} \text{ epi } \check{f} = \text{epi } \check{f}$  by Proposition 9.8(iii). It remains to show that  $\text{epi } \check{f} \subset E$ . We assume that  $f \not\equiv +\infty$ , since otherwise  $\check{f} = f$  and the conclusion is clear. Let us proceed by contradiction and assume that there exists

$$(x, \xi) \in \text{epi } \check{f} \setminus E. \quad (9.5)$$

Since  $E$  is a nonempty closed convex subset of  $\mathcal{H} \times \mathbb{R}$ , Theorem 3.14 implies that the projection  $(p, \pi)$  of  $(x, \xi)$  onto  $E$  satisfies

$$(\forall (y, \eta) \in E) \quad \langle y - p \mid x - p \rangle + (\eta - \pi)(\xi - \pi) \leq 0. \quad (9.6)$$

Letting  $\eta \uparrow +\infty$  in (9.6), we deduce that  $\xi \leq \pi$ . Let us first assume that  $\xi = \pi$ . Then (9.6) yields  $(\forall y \in \overline{\text{conv}} \text{ dom } f) \langle y - p \mid x - p \rangle \leq 0$ . Consequently, since Proposition 9.8(iv) asserts that  $x \in \text{dom } \check{f} \subset \overline{\text{conv}} \text{ dom } f$ , we obtain  $\|x - p\|^2 = 0$  and, in turn,  $(p, \pi) = (x, \xi)$ , which is impossible since  $(x, \xi) \notin E$  by (9.5). Therefore, we must have

$$\xi < \pi. \quad (9.7)$$

Setting  $u = (x - p)/(\pi - \xi)$  and letting  $\eta = f(y)$  in (9.6), we get

$$(\forall y \in \text{dom } f) \quad \langle y - p \mid u \rangle + \pi \leq f(y). \quad (9.8)$$

Consequently,  $f$  is minorized by the lower semicontinuous convex function  $g: y \mapsto \langle y - p \mid u \rangle + \pi$ , and it follows that  $g \leq \check{f}$ . In particular, since  $(x, \xi) \in \text{epi } \check{f}$ , we have

$$\pi \leq \frac{\|x - p\|^2}{\pi - \xi} + \pi = g(x) \leq \check{f}(x) \leq \xi, \quad (9.9)$$

which contradicts (9.7). We conclude that  $\text{epi } \check{f} \subset E$ .  $\square$

**Corollary 9.10** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be convex. Then  $\bar{f} = \check{f}$ .*

*Proof.* Combine Lemma 1.31(vi) and Theorem 9.9.  $\square$

**Corollary 9.11** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and such that  $\text{lev}_{<0} f \neq \emptyset$ . Then  $\text{lev}_{<0} f \subset \text{lev}_{\leq 0} f \subset \text{lev}_{\leq 0} \check{f}$  and  $\overline{\text{lev}_{<0} f} = \overline{\text{lev}_{\leq 0} f} = \text{lev}_{\leq 0} \check{f}$ .*

*Proof.* Take  $x \in \mathcal{H}$ . Then  $f(x) < 0 \Rightarrow f(x) \leq 0 \Rightarrow \check{f}(x) \leq 0$ , which shows the inclusions. Now assume that  $x \in \text{lev}_{\leq 0} \check{f}$ . Then, since  $f$  is convex, Theorem 9.9 yields  $(x, \check{f}(x)) \in \text{epi } \check{f} = \overline{\text{epi } f}$ . Hence there exists a sequence  $(x_n, \xi_n)_{n \in \mathbb{N}}$  in  $\text{epi } f$  that converges to  $(x, \check{f}(x))$ . Now fix  $z \in \text{lev}_{<0} f$  and define a sequence  $(\alpha_n)_{n \in \mathbb{N}}$  in  $]0, 1]$  by

$$(\forall n \in \mathbb{N}) \quad \alpha_n = \begin{cases} \frac{1}{n+1}, & \text{if } \xi_n \leq 0; \\ \min \left\{ 1, \frac{1}{n+1} + \frac{\xi_n}{\xi_n - f(z)} \right\}, & \text{otherwise.} \end{cases} \quad (9.10)$$

Then eventually

$$\begin{aligned} f(\alpha_n z + (1 - \alpha_n)x_n) &\leq \alpha_n f(z) + (1 - \alpha_n)f(x_n) \\ &\leq \alpha_n f(z) + (1 - \alpha_n)\xi_n \\ &< 0. \end{aligned} \quad (9.11)$$

Therefore the sequence  $(\alpha_n z + (1 - \alpha_n)x_n)_{n \in \mathbb{N}}$ , which converges to  $x$ , lies eventually in  $\text{lev}_{<0} f$ . The result follows.  $\square$

## 9.2 Proper Lower Semicontinuous Convex Functions

As illustrated in Proposition 9.6, nonproper lower semicontinuous convex functions are of limited use. By contrast, proper lower semicontinuous convex functions will play a central role in this book.

**Definition 9.12** The set of proper lower semicontinuous convex functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$  is denoted by  $\Gamma_0(\mathcal{H})$ .

**Example 9.13** Let  $(e_i)_{i \in I}$  be a family in  $\mathcal{H}$  and let  $(\phi_i)_{i \in I}$  be a family in  $\Gamma_0(\mathbb{R})$  such that  $(\forall i \in I) \phi_i \geq \phi_i(0) = 0$ . Set  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sum_{i \in I} \phi_i(\langle x | e_i \rangle)$ . Then  $f \in \Gamma_0(\mathcal{H})$ .

*Proof.* Set  $(\forall i \in I) f_i: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \phi_i(\langle x | e_i \rangle)$ . Then  $f = \sum_{i \in I} f_i$  and  $(\forall i \in I) 0 \leq f_i \in \Gamma_0(\mathcal{H})$ . Thus, it follows from Corollary 9.4(ii) that  $f \in \Gamma(\mathcal{H})$ . Finally, since  $f(0) = 0$ ,  $f$  is proper.  $\square$

**Proposition 9.14** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \mathcal{H}$ , and let  $y \in \text{dom } f$ . For every  $\alpha \in ]0, 1[$ , set  $x_\alpha = (1 - \alpha)x + \alpha y$ . Then  $\lim_{\alpha \downarrow 0} f(x_\alpha) = f(x)$ .*

*Proof.* Using the lower semicontinuity and the convexity of  $f$ , we obtain  $f(x) \leq \underline{\lim}_{\alpha \downarrow 0} f(x_\alpha) \leq \overline{\lim}_{\alpha \downarrow 0} f(x_\alpha) \leq \overline{\lim}_{\alpha \downarrow 0} (1 - \alpha)f(x) + \alpha f(y) = f(x)$ . Therefore,  $\lim_{\alpha \downarrow 0} f(x_\alpha) = f(x)$ .  $\square$

**Corollary 9.15** *Let  $f \in \Gamma_0(\mathbb{R})$ . Then  $f|_{\overline{\text{dom}} f}$  is continuous.*

The conclusion of Corollary 9.15 fails in general Hilbert spaces and even in the Euclidean plane (see Example 9.27 below).

We conclude this section with an extension of Fact 6.13.

**Fact 9.16** [233, Corollary 13.2] *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Then*

$$\text{int}(\text{dom } f - \text{dom } g) = \text{core}(\text{dom } f - \text{dom } g). \quad (9.12)$$

### 9.3 Affine Minorization

A key property of functions in  $\Gamma_0(\mathcal{H})$  is that they possess continuous affine minorants. To see this, we require the following two results.

**Proposition 9.17** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $(x, \xi) \in \mathcal{H} \times \mathbb{R}$ , and let  $(p, \pi) \in \mathcal{H} \times \mathbb{R}$ . Then  $(p, \pi) = P_{\text{epi } f}(x, \xi)$  if and only if*

$$\max\{\xi, f(p)\} \leq \pi \quad (9.13)$$

and

$$(\forall y \in \text{dom } f) \quad \langle y - p \mid x - p \rangle + (f(y) - \pi)(\xi - \pi) \leq 0. \quad (9.14)$$

*Proof.* Since  $f \in \Gamma_0(\mathcal{H})$ , the set  $\text{epi } f$  is nonempty, closed, and convex. Now set  $(p, \pi) = P_{\text{epi } f}(x, \xi)$ . Then Theorem 3.14 implies that  $(p, \pi)$  is characterized by  $(p, \pi) \in \text{epi } f$  and  $(\forall (y, \eta) \in \text{epi } f) \langle y - p \mid x - p \rangle + (\eta - \pi)(\xi - \pi) \leq 0$ , which is equivalent to  $f(p) \leq \pi$  and  $(\forall y \in \text{dom } f)(\forall \lambda \in \mathbb{R}_+) \langle y - p \mid x - p \rangle + (f(y) + \lambda - \pi)(\xi - \pi) \leq 0$ . By letting  $\lambda \uparrow +\infty$ , we deduce that  $\xi \leq \pi$ . The characterization follows.  $\square$

**Proposition 9.18** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \text{dom } f$ , let  $\xi \in ]-\infty, f(x)[$ , and let  $(p, \pi) \in \mathcal{H} \times \mathbb{R}$ . Then  $(p, \pi) = P_{\text{epi } f}(x, \xi)$  if and only if*

$$\xi < f(p) = \pi \quad (9.15)$$

and

$$(\forall y \in \text{dom } f) \quad \langle y - p \mid x - p \rangle \leq (f(y) - f(p))(f(p) - \xi). \quad (9.16)$$

*Proof.* Suppose first that  $(p, \pi) = P_{\text{epi } f}(x, \xi)$ . Since  $p \in \text{dom } f$ , (9.14) yields

$$(f(p) - \pi)(\xi - \pi) \leq 0. \quad (9.17)$$

To establish that  $\xi < f(p)$ , we argue by contradiction. Suppose that  $f(p) \leq \xi$ . Then  $f(p) - \pi \leq \xi - \pi$  and hence, since  $\xi - \pi \leq 0$  by (9.13), we obtain  $(f(p) - \pi)(\xi - \pi) \geq (\xi - \pi)^2$ . In view of (9.17), we deduce that  $\xi = \pi$ . In turn, since  $x \in \text{dom } f$ , (9.14) implies that  $\langle x - p \mid x - p \rangle \leq 0$ . Thus  $x = p$  and hence  $(p, \pi) = (x, \xi)$ . This is impossible, since  $(p, \pi) \in \text{epi } f$  and  $(x, \xi) \notin \text{epi } f$ . Thus,

$$\xi < f(p), \quad (9.18)$$

and (9.13) implies that  $\xi < \pi$  and  $f(p) \leq \pi$ . Hence, (9.17) yields  $f(p) = \pi$  and (9.15) holds. Combining (9.15) and Proposition 9.17, we obtain (9.16).

Conversely, if (9.15) and (9.16) hold, then Proposition 9.17 implies directly that  $(p, \pi) = P_{\text{epi } f}(x, \xi)$ .  $\square$

**Theorem 9.19** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f$  possesses a continuous affine minorant.*

*Proof.* Fix  $x \in \text{dom } f$  and  $\xi \in ]-\infty, f(x)[$ , and set  $(p, \pi) = P_{\text{epi } f}(x, \xi)$ . Then, by (9.15),  $f(p) > \xi$ . Now set  $u = (x - p)/(f(p) - \xi)$  and  $g: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \langle y - p \mid u \rangle + f(p)$ . Then (9.16) yields  $g \leq f$ .  $\square$

**Corollary 9.20** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f$  is bounded below on every nonempty bounded subset of  $\mathcal{H}$ .*

*Proof.* Let  $C$  be a nonempty bounded subset of  $\mathcal{H}$  and set  $\beta = \sup_{x \in C} \|x\|$ . Theorem 9.19 asserts that  $f$  has a continuous affine minorant, say  $\langle \cdot \mid u \rangle + \eta$ . Then, by Cauchy–Schwarz,  $(\forall x \in C) \ f(x) \geq \langle x \mid u \rangle + \eta \geq -\|x\| \|u\| + \eta \geq -\beta \|u\| + \eta > -\infty$ .  $\square$

**Example 9.21** Suppose that  $\mathcal{H}$  is infinite-dimensional and let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be a discontinuous linear functional (see Example 2.20 and Example 8.33). Then  $f$  has no continuous affine minorant.

*Proof.* Assume that the conclusion is false, i.e., that there exist  $u \in \mathcal{H}$  and  $\eta \in \mathbb{R}$  such that  $(\forall x \in \mathcal{H}) \ \langle x \mid u \rangle + \eta \leq f(x)$ . Then, since  $f$  is odd,  $(\forall x \in \mathcal{H}) \ f(x) \leq \langle x \mid u \rangle - \eta \leq \|x\| \|u\| - \eta$ . Consequently,  $\sup f(B(0; 1)) \leq \|u\| - \eta$  and therefore  $f$  is bounded above on a neighborhood of 0. This contradicts Corollary 8.30(i) since  $f$  is nowhere continuous.  $\square$

**Theorem 9.22** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $x \in \text{int dom } f$ . Then there exists a continuous affine minorant  $a$  of  $f$  such that  $a(x) = f(x)$ . In other words,  $(\exists u \in \mathcal{H})(\forall y \in \mathcal{H}) \ \langle y - x \mid u \rangle + f(x) \leq f(y)$ .*

*Proof.* In view of Corollary 8.30,  $x \in \text{cont } f$ . Hence, it follows from Theorem 8.29 and Proposition 8.36 that  $\text{int epi } f \neq \emptyset$ . In turn, Proposition 7.5 implies that  $(x, f(x)) \in \text{spts}(\text{epi } f)$ , and we therefore derive from Theorem 7.4 that there exists  $(z, \zeta) \in (\mathcal{H} \times \mathbb{R}) \setminus (\text{epi } f)$  such that  $(x, f(x)) = P_{\text{epi } f}(z, \zeta)$ . In view of Proposition 3.19 and since  $x \in \text{int dom } f$ , we assume that  $z \in \text{int dom } f$ . Thus, by Proposition 9.17,  $\max\{\zeta, f(x)\} \leq f(x)$ , i.e.,



$$f(x) \geq \zeta \quad (9.19)$$

and

$$(\forall y \in \text{dom } f) \quad \langle y - x \mid z - x \rangle + (f(y) - f(x))(\zeta - f(x)) \leq 0. \quad (9.20)$$

If  $f(x) = \zeta$ , then the above inequality evaluated at  $y = z$  yields  $z = x$ , which contradicts the fact that  $(z, \zeta) \neq (x, f(x))$ . Hence  $f(x) > \zeta$ . Now set  $u = (z - x)/(f(x) - \zeta)$ . Then (9.20) becomes  $(\forall y \in \text{dom } f) \langle y - x \mid u \rangle + f(x) - f(y) \leq 0$ , and the result follows.  $\square$

**Proposition 9.23 (Jensen's inequality)** *Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space such that  $\mu(\Omega) \in \mathbb{R}_{++}$ , let  $\phi \in \Gamma_0(\mathbb{R})$ , and let  $x: \Omega \rightarrow \mathbb{R}$  be a measurable function such that  $\mu(\Omega)^{-1} \int_{\Omega} x(\omega) \mu(d\omega) \in \text{int dom } \phi$ . Then*

$$\phi\left(\frac{1}{\mu(\Omega)} \int_{\Omega} x(\omega) \mu(d\omega)\right) \leq \frac{1}{\mu(\Omega)} \int_{\Omega} \phi(x(\omega)) \mu(d\omega). \quad (9.21)$$

*Proof.* Since  $\phi$  is lower semicontinuous, it is measurable, and so is therefore  $\phi \circ x$ . Now set  $\xi = \mu(\Omega)^{-1} \int_{\Omega} x d\mu$ . It follows from Theorem 9.22 that there exists  $\alpha \in \mathbb{R}$  such that  $(\forall \eta \in \mathbb{R}) \alpha(\eta - \xi) + \phi(\xi) \leq \phi(\eta)$ . Thus, for  $\mu$ -almost every  $\omega \in \Omega$ ,  $\alpha(x(\omega) - \xi) + \phi(\xi) \leq \phi(x(\omega))$ . Integrating these inequalities over  $\Omega$  with respect to  $\mu$  yields  $\phi(\xi)\mu(\Omega) \leq \int_{\Omega} \phi(x(\omega)) \mu(d\omega)$ .  $\square$

**Example 9.24** Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space such that  $\mu(\Omega) \in \mathbb{R}_{++}$ , let  $(\mathbf{H}, \|\cdot\|_{\mathbf{H}})$  be a real Hilbert space, and take  $p$  and  $q$  in  $\mathbb{R}_{++}$  such that  $p < q$ . Then the following hold:

(i) Let  $x \in L^p((\Omega, \mathcal{F}, \mu); \mathbf{H}))$ . Then

$$\left( \int_{\Omega} \|x(\omega)\|_{\mathbf{H}}^p \mu(d\omega) \right)^{1/p} \leq \mu(\Omega)^{1/p-1/q} \left( \int_{\Omega} \|x(\omega)\|_{\mathbf{H}}^q \mu(d\omega) \right)^{1/q}. \quad (9.22)$$

(ii)  $L^q((\Omega, \mathcal{F}, \mu); \mathbf{H}) \subset L^p((\Omega, \mathcal{F}, \mu); \mathbf{H})$ .

*Proof.* (i): Set  $\phi = |\cdot|^{q/p}$ . Then it follows from Example 8.21 that  $\phi$  is convex. Now let  $x \in L^p((\Omega, \mathcal{F}, \mu); \mathbf{H}))$  and set  $y: \omega \mapsto \|x(\omega)\|_{\mathbf{H}}^p$ . Since  $y$  is integrable,  $\mu(\Omega)^{-1} \int_{\Omega} y d\mu \in \mathbb{R} = \text{dom } \phi$ , and Proposition 9.23 applied to  $y$  yields (9.22).

(ii): An immediate consequence of (i).  $\square$

**Example 9.25** Let  $X$  be a random variable, and take  $p$  and  $q$  in  $\mathbb{R}_{++}$  such that  $p < q$  and  $\mathbb{E}|X|^p < +\infty$ . Then  $\mathbb{E}^{1/p}|X|^p \leq \mathbb{E}^{1/q}|X|^q$ .

*Proof.* Let  $\mu$  be a probability measure and set  $\mathbf{H} = \mathbb{R}$  in Example 9.24(i) (see Example 2.8).  $\square$

## 9.4 Construction of Functions in $\Gamma_0(\mathcal{H})$

We start with a basic tool for constructing functions in  $\Gamma_0(\mathcal{H})$ .

**Proposition 9.26** *Let  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper convex function such that  $\text{dom } g$  is open and  $g$  is continuous on  $\text{dom } g$ . Set*

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} g(x), & \text{if } x \in \text{dom } g; \\ \lim_{y \rightarrow x} g(y), & \text{if } x \in \text{bdry } \text{dom } g; \\ +\infty, & \text{if } x \in \mathcal{H} \setminus \overline{\text{dom } g}. \end{cases} \quad (9.23)$$

Then  $f = \check{g}$  and  $f \in \Gamma_0(\mathcal{H})$ .

*Proof.* Set  $C = \text{dom } g$ . To show that  $f = \check{g}$  we shall repeatedly utilize Proposition 9.8. Note that, since  $g \geq \check{g}$ , we have  $C \subset \text{dom } \check{g} \subset \overline{C}$ . Let  $x \in \mathcal{H}$  and assume first that  $x \in C$ . Then  $+\infty > g(x) \geq \check{g}(x)$ . By Theorem 9.9, there exists a sequence  $(x_n, \xi_n)_{n \in \mathbb{N}}$  in  $\text{epi } g$  such that  $(x_n, \xi_n) \rightarrow (x, \check{g}(x))$ . Hence  $\check{g}(x) = \lim \xi_n = \underline{\lim} \xi_n \geq \underline{\lim} g(x_n) \geq \underline{\lim} \check{g}(x_n) \geq \check{g}(x)$  and so  $f(x) = g(x) = \lim g(x_n) = \underline{\lim} g(x_n) = \check{g}(x)$ . Consequently,  $f = \check{g}$  on  $C$ . If  $x \in \mathcal{H} \setminus \overline{C}$ , then  $f(x) = +\infty = \check{g}(x)$  and thus  $f = \check{g}$  on  $\mathcal{H} \setminus \overline{C}$ . If  $x \in (\text{bdry } C) \setminus (\text{dom } \check{g})$ , then  $+\infty \geq f(x) = \underline{\lim}_{y \rightarrow x} g(y) \geq \underline{\lim}_{y \rightarrow x} \check{g}(y) = \check{g}(x) = +\infty$  and thus  $f(x) = \check{g}(x) = +\infty$ . Finally, we assume that  $x \in (\text{bdry } C) \cap (\text{dom } \check{g})$ . Using Theorem 9.9 again, we see that there exists a sequence  $(x_n, \xi_n)_{n \in \mathbb{N}}$  in  $\text{epi } g$  such that  $(x_n, \xi_n) \rightarrow (x, \check{g}(x))$ . Hence  $f(x) = \underline{\lim}_{y \rightarrow x} g(y) \geq \underline{\lim}_{y \rightarrow x} \check{g}(y) = \check{g}(x) = \lim \xi_n = \underline{\lim} \xi_n \geq \underline{\lim} g(x_n) \geq \underline{\lim}_{y \rightarrow x} g(y) = f(x)$  and therefore  $f(x) = \check{g}(x)$ . We have verified that  $f = \check{g}$ . It follows that  $f$  is lower semicontinuous and convex. Since  $f$  is real-valued on  $C$ , Proposition 9.6 implies that  $f$  is also proper.  $\square$

**Example 9.27** The function

$$f: \mathbb{R}^2 \rightarrow ]-\infty, +\infty]: (\xi, \eta) \mapsto \begin{cases} \eta^2/\xi, & \text{if } \xi > 0; \\ 0, & \text{if } (\xi, \eta) = (0, 0); \\ +\infty, & \text{otherwise,} \end{cases} \quad (9.24)$$

belongs to  $\Gamma_0(\mathbb{R}^2)$  and  $f|_{\text{dom } f}$  is not continuous at  $(0, 0)$ .

*Proof.* Set

$$g: \mathbb{R}^2 \rightarrow ]-\infty, +\infty]: (\xi, \eta) \mapsto \begin{cases} \eta^2/\xi, & \text{if } \xi > 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.25)$$

The convexity of  $t \mapsto t^2$  and Proposition 8.23 imply that  $g$  is proper and convex. Moreover, Proposition 9.26 yields  $\check{g} = f \in \Gamma_0(\mathbb{R}^2)$ . Now set  $x = (0, 0)$ , fix a sequence  $(\alpha_n)_{n \in \mathbb{N}}$  in  $\mathbb{R}_{++}$  such that  $\alpha_n \downarrow 0$ , and set  $(\forall n \in \mathbb{N})$

$x_n = (\alpha_n^2, \alpha_n)$ . Then  $(x_n)_{n \in \mathbb{N}}$  lies in  $\text{dom } f$  and  $x_n \rightarrow x$ , but  $\lim f(x_n) = 1 \neq 0 = f(x)$ .  $\square$

The following result concerns the construction of strictly convex functions in  $\Gamma_0(\mathbb{R})$ .

**Proposition 9.28** *Let  $g: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be strictly convex and proper, and suppose that  $\text{dom } g = ]\alpha, \beta[$ , where  $\alpha$  and  $\beta$  are in  $[-\infty, +\infty]$  and  $\alpha < \beta$ . Set*

$$f: \mathbb{R} \rightarrow ]-\infty, +\infty] : x \mapsto \begin{cases} g(x), & \text{if } x \in ]\alpha, \beta[; \\ \lim_{y \downarrow \alpha} g(y), & \text{if } x = \alpha; \\ \lim_{y \uparrow \beta} g(y), & \text{if } x = \beta; \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.26)$$

Then  $f$  is strictly convex,  $f = \check{g}$ , and  $f \in \Gamma_0(\mathbb{R})$ .

*Proof.* Proposition 9.14, Corollary 8.30(iii), and Proposition 9.26 imply that  $f$  is convex and that  $f = \check{g} \in \Gamma_0(\mathbb{R})$ . To verify strict convexity, suppose that  $x$  and  $y$  are distinct points in  $\text{dom } f$ , take  $\gamma \in ]0, 1[$ , and suppose that  $f(\gamma x + (1 - \gamma)y) = \gamma f(x) + (1 - \gamma)f(y)$ . By Exercise 8.1,  $(\forall \lambda \in ]0, 1[) f(\lambda x + (1 - \lambda)y) = \lambda f(x) + (1 - \lambda)f(y)$ . Since  $]x, y[ \subset ]\alpha, \beta[$  and  $f = g$  on  $]x, y[$ , this contradicts the strict convexity of  $g$ .  $\square$

The next two examples are consequences of Proposition 9.28 and Proposition 8.12(ii).

**Example 9.29 (entropy)** The negative *Boltzmann–Shannon entropy* function

$$\mathbb{R} \rightarrow ]-\infty, +\infty] : x \mapsto \begin{cases} x \ln(x) - x, & \text{if } x > 0; \\ 0, & \text{if } x = 0; \\ +\infty, & \text{if } x < 0, \end{cases} \quad (9.27)$$

is strictly convex and belongs to  $\Gamma_0(\mathbb{R})$ .

**Example 9.30** The following are strictly convex functions in  $\Gamma_0(\mathbb{R})$ :

- (i)  $x \mapsto \exp(x)$ .
- (ii)  $x \mapsto |x|^p$ , where  $p \in ]1, +\infty[$ .
- (iii)  $x \mapsto \begin{cases} 1/x^p, & \text{if } x > 0; \\ +\infty, & \text{otherwise,} \end{cases}$  where  $p \in [1, +\infty[$ .
- (iv)  $x \mapsto \begin{cases} -x^p, & \text{if } x \geq 0; \\ +\infty, & \text{otherwise,} \end{cases}$  where  $p \in ]0, 1[$ .
- (v)  $x \mapsto \begin{cases} 1/\sqrt{1-x^2}, & \text{if } |x| < 1; \\ +\infty, & \text{otherwise.} \end{cases}$

- (vi)  $x \mapsto \begin{cases} -\sqrt{1-x^2}, & \text{if } |x| \leq 1; \\ +\infty, & \text{otherwise.} \end{cases}$
- (vii)  $x \mapsto \begin{cases} x \ln(x) + (1-x) \ln(1-x), & \text{if } x \in ]0, 1[; \\ 0, & \text{if } x \in \{0, 1\}; \\ +\infty, & \text{otherwise.} \end{cases}$
- (viii)  $x \mapsto \begin{cases} -\ln(x), & \text{if } x > 0; \\ +\infty, & \text{otherwise} \end{cases} \quad (\text{negative Burg entropy function}).$

**Remark 9.31** By utilizing direct sum constructions (see Proposition 8.25 and Exercise 8.12), we can construct a (strictly) convex function in  $\Gamma_0(\mathbb{R}^N)$  from (strictly) convex functions in  $\Gamma_0(\mathbb{R})$ .

We now turn our attention to the construction of proper lower semicontinuous convex integral functions (see Example 2.5 for notation).

**Proposition 9.32** *Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space, let  $(\mathbf{H}, \langle \cdot | \cdot \rangle_{\mathbf{H}})$  be a real Hilbert space, and let  $\varphi \in \Gamma_0(\mathbf{H})$ . Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); \mathbf{H})$  and that one of the following holds:*

- (i)  $\mu(\Omega) < +\infty$ .
- (ii)  $\varphi \geq \varphi(0) = 0$ .

Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$x \mapsto \begin{cases} \int_{\Omega} \varphi(x(\omega)) \mu(d\omega), & \text{if } \varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R}); \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.28)$$

Then  $f \in \Gamma_0(\mathcal{H})$ .

*Proof.* We first observe that, since  $\varphi$  is lower semicontinuous, it is measurable, and so is therefore  $\varphi \circ x$  for every  $x \in \mathcal{H}$ . Let us now show that  $f \in \Gamma_0(\mathcal{H})$ .

(i): By Theorem 9.19, there exists a continuous affine function  $\psi: \mathbf{H} \rightarrow \mathbb{R}$  such that  $\varphi \geq \psi$ , say  $\psi = \langle \cdot | \mathbf{u} \rangle_{\mathbf{H}} + \eta$  for some  $\mathbf{u} \in \mathbf{H}$  and  $\eta \in \mathbb{R}$ . Let us set  $u: \Omega \rightarrow \mathbf{H}: \omega \mapsto \mathbf{u}$ . Then  $u \in \mathcal{H}$  since  $\int_{\Omega} \|u(\omega)\|_{\mathbf{H}}^2 \mu(d\omega) = \|\mathbf{u}\|_{\mathbf{H}}^2 \mu(\Omega) < +\infty$ . Moreover, for every  $x \in \mathcal{H}$ ,  $\varphi \circ x \geq \psi \circ x$  and

$$\int_{\Omega} \psi(x(\omega)) \mu(d\omega) = \int_{\Omega} \langle x(\omega) | \mathbf{u} \rangle_{\mathbf{H}} \mu(d\omega) + \eta \mu(\Omega) = \langle x | u \rangle + \eta \mu(\Omega) \in \mathbb{R}. \quad (9.29)$$

Thus, Proposition 8.22 asserts that  $f$  is well defined and convex, with  $\text{dom } f = \{x \in \mathcal{H} \mid \varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R})\}$ . It also follows from (9.28) and (9.29) that

$$(\forall x \in \text{dom } f) \quad f(x) = \int_{\Omega} (\varphi - \psi)(x(\omega)) \mu(d\omega) + \langle x | u \rangle + \eta \mu(\Omega). \quad (9.30)$$

Now take  $z \in \text{dom } \varphi$  and set  $z: \Omega \rightarrow \mathbb{H}: \omega \mapsto z$ . Then  $z \in \mathcal{H}$  and  $\int_{\Omega} |\varphi \circ z| d\mu = |\varphi(z)|\mu(\Omega) < +\infty$ . Hence,  $\varphi \circ z \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R})$ . This shows that  $f$  is proper. Next, to show that  $f$  is lower semicontinuous, let us fix  $\xi \in \mathbb{R}$  and a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\text{lev}_{\leq \xi} f$  that converges to some  $x \in \mathcal{H}$ . In view of Lemma 1.24, it suffices to show that  $f(x) \leq \xi$ . Since  $\|x_n(\cdot) - x(\cdot)\|_{\mathbb{H}} \rightarrow 0$  in  $L^2((\Omega, \mathcal{F}, \mu); \mathbb{R})$ , there exists a subsequence  $(x_{k_n})_{n \in \mathbb{N}}$  such that  $x_{k_n}(\omega) \xrightarrow{\mathbb{H}} x(\omega)$  for  $\mu$ -almost every  $\omega \in \Omega$  [3, Theorem 2.5.1 & Theorem 2.5.3]. Now set  $\phi = (\varphi - \psi) \circ x$  and  $(\forall n \in \mathbb{N}) \phi_n = (\varphi - \psi) \circ x_{k_n}$ . Since  $\varphi - \psi$  is lower semicontinuous, we have

$$\phi(\omega) = (\varphi - \psi)(x(\omega)) \leq \varliminf (\varphi - \psi)(x_{k_n}(\omega)) = \varliminf \phi_n(\omega) \quad \mu\text{-a.e. on } \Omega. \quad (9.31)$$

On the other hand, since  $\inf_{n \in \mathbb{N}} \phi_n \geq 0$ , Fatou's lemma [3, Lemma 1.6.8] yields  $\int_{\Omega} \varliminf \phi_n d\mu \leq \varliminf \int_{\Omega} \phi_n d\mu$ . Hence, we derive from (9.30) and (9.31) that

$$\begin{aligned} f(x) &= \int_{\Omega} \phi d\mu + \langle x \mid u \rangle + \eta\mu(\Omega) \\ &\leq \int_{\Omega} \varliminf \phi_n d\mu + \langle x \mid u \rangle + \eta\mu(\Omega) \\ &\leq \varliminf \int_{\Omega} \phi_n d\mu + \lim \langle x_{k_n} \mid u \rangle + \eta\mu(\Omega) \\ &= \varliminf \int_{\Omega} (\varphi \circ x_{k_n}) d\mu \\ &= \varliminf f(x_{k_n}) \\ &\leq \xi. \end{aligned} \quad (9.32)$$

(ii): Since (8.16) holds with  $\varrho = 0$ , it follows from Proposition 8.22 that  $f$  is a well-defined convex function. In addition, since  $\varphi(0) = 0$ , (9.28) yields  $f(0) = 0$ . Thus,  $f$  is proper. Finally, to prove that  $f$  is lower semicontinuous, we follow the same procedure as above with  $\psi = 0$ .  $\square$

**Example 9.33 (Boltzmann–Shannon entropy)** Let  $(\Omega, \mathcal{F}, \mu)$  be a finite measure space and suppose that  $\mathcal{H} = L^2(\Omega, \mathcal{F}, \mu)$  (see Example 2.6). Using the convention  $0 \ln(0) = 0$ , set

$$\begin{aligned} f: \mathcal{H} &\rightarrow ]-\infty, +\infty] \\ x &\mapsto \begin{cases} \int_{\Omega} (x(\omega) \ln(x(\omega)) - x(\omega)) \mu(d\omega), & \text{if } x \geq 0 \quad \mu\text{-a.e. on } \Omega; \\ +\infty, & \text{otherwise.} \end{cases} \end{aligned} \quad (9.33)$$

Then  $f \in \Gamma_0(\mathcal{H})$ . In particular, this is true in the following cases:

- (i) Entropy of a random variable:  $\mathcal{H} = L^2(\Omega, \mathcal{F}, \mathbb{P})$ , where  $(\Omega, \mathcal{F}, \mathbb{P})$  is a probability space (see Example 2.8), and

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$X \mapsto \begin{cases} \mathbb{E}(X \ln(X) - X), & \text{if } X \geq 0 \text{ a.s.}; \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.34)$$

(ii) Discrete entropy:  $\mathcal{H} = \mathbb{R}^N$  and

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$(\xi_k)_{1 \leq k \leq N} \mapsto \begin{cases} \sum_{k=1}^N \xi_k \ln(\xi_k) - \xi_k, & \text{if } \min_{1 \leq k \leq N} \xi_k \geq 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.35)$$

*Proof.* Denote by  $\varphi$  the function defined in (9.27). Then Example 9.29 asserts that  $\varphi \in \Gamma_0(\mathbb{R})$ . First, take  $x \in \mathcal{H}$  such that  $x \geq 0$   $\mu$ -a.e., and set  $C = \{\omega \in \Omega \mid 0 \leq x(\omega) < 1\}$  and  $D = \{\omega \in \Omega \mid x(\omega) \geq 1\}$ . Since, for every  $\xi \in \mathbb{R}_+$ ,  $|\varphi(\xi)| = |\xi \ln(\xi) - \xi| \leq 1_{[0,1[}(\xi) + \xi^2 1_{[1,+\infty[}(\xi)$ , we have

$$\begin{aligned} \int_{\Omega} |\varphi(x(\omega))| \mu(d\omega) &= \int_C |\varphi(x(\omega))| \mu(d\omega) + \int_D |\varphi(x(\omega))| \mu(d\omega) \\ &\leq \mu(C) + \int_D |x(\omega)|^2 \mu(d\omega) \\ &\leq \mu(\Omega) + \|x\|^2 \\ &< +\infty, \end{aligned} \quad (9.36)$$

and therefore  $\varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R})$ . Now take  $x \in \mathcal{H}$  and set  $A = \{\omega \in \Omega \mid x(\omega) \geq 0\}$  and  $B = \{\omega \in \Omega \mid x(\omega) < 0\}$ . Then

$$\begin{aligned} \int_{\Omega} \varphi(x(\omega)) \mu(d\omega) &= \int_A \varphi(x(\omega)) \mu(d\omega) + \int_B \varphi(x(\omega)) \mu(d\omega) \\ &= \begin{cases} \int_{\Omega} x(\omega) (\ln(x(\omega)) - 1) \mu(d\omega), & \text{if } x \geq 0 \text{ } \mu\text{-a.e. on } \Omega; \\ +\infty, & \text{otherwise} \end{cases} \\ &= f(x). \end{aligned} \quad (9.37)$$

Altogether, it follows from Proposition 9.32(i) with  $\mathbb{H} = \mathbb{R}$  that  $f \in \Gamma_0(\mathcal{H})$ .

(i): Special case when  $\mu$  is a probability measure.

(ii): Special case when  $\Omega = \{1, \dots, N\}$ ,  $\mathcal{F} = 2^{\Omega}$ , and  $\mu$  is the counting measure, i.e., for every  $C \in 2^{\Omega}$ ,  $\mu(C)$  is the cardinality of  $C$ .  $\square$

## Exercises

**Exercise 9.1** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be lower semicontinuous and *midpoint convex* in the sense that

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \quad f\left(\frac{x+y}{2}\right) \leq \frac{f(x) + f(y)}{2}. \quad (9.38)$$

Show that  $f$  is convex.

**Exercise 9.2** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be midpoint convex. Show that  $f$  need not be convex.

**Exercise 9.3** Provide a family of continuous linear functions the supremum of which is neither continuous nor linear.

**Exercise 9.4** Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $\mathbb{R} \cap \text{ran } f$  is convex, and provide an example in which  $\text{ran } f$  is not convex.

**Exercise 9.5** Provide an example of a convex function  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  such that  $\text{ran } f = \{-\infty, 0, +\infty\}$ . Compare with Proposition 9.6.

**Exercise 9.6** Set  $\mathcal{C} = \{C \subset \mathcal{H} \mid C \text{ is nonempty, closed, and convex}\}$  and set

$$(\forall C \in \mathcal{C}) \quad \gamma_C: \mathcal{H} \rightarrow [-\infty, +\infty] : x \mapsto \begin{cases} -\infty, & \text{if } x \in C; \\ +\infty, & \text{otherwise.} \end{cases} \quad (9.39)$$

Prove that  $\mathcal{C} \rightarrow \{f \in \Gamma(\mathcal{H}) \mid -\infty \in f(\mathcal{H})\} : C \mapsto \gamma_C$  is a bijection.

**Exercise 9.7** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex. Show that  $f$  is continuous if and only if it is lower semicontinuous and  $\text{cont } f = \text{dom } f$ .

**Exercise 9.8** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and set  $\mu = \inf f(\mathcal{H})$ . Prove the following statements:

- (i)  $f \in \Gamma(\mathcal{H}) \Leftrightarrow (\forall \xi \in ]\mu, +\infty[) \text{ lev}_{\leq \xi} f = \overline{\text{lev}_{< \xi} f}$ .
- (ii)  $\text{cont } f = \text{dom } f \Leftrightarrow (\forall \xi \in ]\mu, +\infty[) \text{ lev}_{< \xi} f = \text{int lev}_{\leq \xi} f$ .
- (iii)  $f$  is continuous  $\Leftrightarrow (\forall \xi \in ]\mu, +\infty[) \text{ lev}_{= \xi} f = \text{bdry lev}_{\leq \xi} f$ .

**Exercise 9.9** Let  $(e_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$ , let  $(\omega_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_+$ , and let  $(p_n)_{n \in \mathbb{N}}$  be a sequence in  $[1, +\infty[$ . Set  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  :  $x \mapsto \sum_{n \in \mathbb{N}} \omega_n |\langle x \mid e_n \rangle|^{p_n}$ . Show that  $f \in \Gamma_0(\mathcal{H})$ .

**Exercise 9.10** Use Proposition 8.12(ii) and Proposition 9.28 to verify Example 9.29 and Example 9.30.





# Chapter 10

## Convex Functions: Variants

In this chapter we present variants of the notion of convexity for functions. The most important are the weaker notion of quasiconvexity, and the stronger notions of uniform and strong convexity.

### 10.1 Between Linearity and Convexity

**Definition 10.1** A function  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  is

- (i) *positively homogeneous* if  $(\forall x \in \mathcal{H})(\forall \lambda \in \mathbb{R}_{++}) f(\lambda x) = \lambda f(x)$ ;
- (ii) *subadditive* if  $(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) f(x + y) \leq f(x) + f(y)$ ;
- (iii) *sublinear* if it is positively homogeneous and subadditive.

The proofs of the following results are left as Exercise 10.1 and Exercise 10.2, respectively.

**Proposition 10.2** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is positively homogeneous if and only if  $\text{epi } f$  is a cone; in this case,  $\text{dom } f$  is also a cone.*

**Proposition 10.3** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be positively homogeneous. Then  $f$  is sublinear if and only if it is convex.*

Clearly, linearity implies sublinearity, which in turn implies convexity. However, as we now illustrate, neither implication is reversible.

**Example 10.4** Suppose that  $\mathcal{H} \neq \{0\}$ . Then the following hold:

- (i)  $\|\cdot\|$  is sublinear, but not linear.
- (ii)  $\|\cdot\|^2$  is convex, but not sublinear.

## 10.2 Uniform and Strong Convexity

We introduce more restrictive versions of strict convexity.

**Definition 10.5** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $f$  is *uniformly convex* with modulus  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  if  $\phi$  is increasing,  $\phi$  vanishes only at 0, and

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) + \alpha(1 - \alpha)\phi(\|x - y\|) \leq \alpha f(x) + (1 - \alpha)f(y). \quad (10.1)$$

If (10.1) holds with  $\phi = (\beta/2) \cdot \|\cdot\|^2$  for some  $\beta \in \mathbb{R}_{++}$ , then  $f$  is *strongly convex* with constant  $\beta$ . Now let  $C$  be a nonempty subset of  $\text{dom } f$ . Then  $f$  is uniformly convex on  $C$  if

$$(\forall x \in C)(\forall y \in C)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) + \alpha(1 - \alpha)\phi(\|x - y\|) \leq \alpha f(x) + (1 - \alpha)f(y). \quad (10.2)$$

Clearly, strong convexity implies uniformly convexity, uniformly convexity implies strict convexity, and strict convexity implies convexity. In the exercises, we provide examples that show that none of these implications is reversible.

**Proposition 10.6** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and let  $\beta \in \mathbb{R}_{++}$ . Then  $f$  is strongly convex with constant  $\beta$  if and only if  $f - (\beta/2)\|\cdot\|^2$  is convex.

*Proof.* A direct consequence of Corollary 2.14. □

**Example 10.7** Suppose that  $\mathcal{H} \neq \{0\}$ . Then the following hold:

- (i)  $\|\cdot\|$  is sublinear, but not strictly (hence not uniformly) convex.
- (ii)  $\|\cdot\|^2$  is strongly convex with constant 2, but not positively homogeneous.

**Example 10.8** Let  $\phi: \mathbb{R}_+ \rightarrow \mathbb{R}_+$  be increasing and let  $C$  be a nonempty bounded convex subset of  $\mathcal{H}$ . Set

$$f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \int_0^{\|x\|} \phi(t) dt. \quad (10.3)$$

Then  $f$  is uniformly convex on  $C$  [260, Theorem 4.1(ii)]. In particular, taking  $p \in ]1, +\infty[$  and  $\phi: t \mapsto pt^{p-1}$ , we obtain that  $\|\cdot\|^p + \iota_C$  is uniformly convex.

**Definition 10.9** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. The *exact modulus of convexity* of  $f$  is

$$\begin{aligned} \varphi: \mathbb{R}_+ &\rightarrow [0, +\infty] \\ t &\mapsto \inf_{\substack{x \in \text{dom } f, y \in \text{dom } f, \\ \|x-y\|=t, \alpha \in ]0,1[}} \frac{\alpha f(x) + (1-\alpha)f(y) - f(\alpha x + (1-\alpha)y)}{\alpha(1-\alpha)}. \end{aligned} \quad (10.4)$$

**Proposition 10.10** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, with exact modulus of convexity  $\varphi$ . Then  $\varphi(0) = 0$ ,*

$$(\forall t \in \mathbb{R}_+)(\forall \gamma \in [1, +\infty[) \quad \varphi(\gamma t) \geq \gamma^2 \varphi(t), \quad (10.5)$$

and  $\varphi$  is increasing.

*Proof.* It is clear that  $\varphi(\mathbb{R}_+) \subset [0, +\infty]$ , that  $\varphi(0) = 0$ , that (10.5) holds when  $t = 0$  or  $\gamma = 1$ , and that (10.5) implies that  $\varphi$  is increasing. To show (10.5), we fix  $t \in \mathbb{R}_{++}$  and  $\gamma \in ]1, +\infty[$  such that  $\varphi(\gamma t) < +\infty$ , and we verify that

$$\varphi(\gamma t) \geq \gamma^2 \varphi(t). \quad (10.6)$$

We consider two cases.

(a)  $\gamma \in ]1, 2[$ : Fix  $\varepsilon \in \mathbb{R}_{++}$ . Since  $\varphi(\gamma t) < +\infty$ , there exist  $x \in \text{dom } f$ ,  $y \in \text{dom } f$ , and  $\alpha \in ]0, 1/2]$  such that  $\|x - y\| = \gamma t$  and

$$\varphi(\gamma t) + \varepsilon > \frac{\alpha f(x) + (1-\alpha)f(y) - f(\alpha x + (1-\alpha)y)}{\alpha(1-\alpha)}. \quad (10.7)$$

Now set  $z_\alpha = \alpha x + (1-\alpha)y$ ,  $\delta = 1/\gamma$ , and  $z_\delta = \delta x + (1-\delta)y$ . Then  $\|z_\delta - y\| = t$ ,  $\gamma\alpha \in ]0, 1[$ , and  $z_\alpha = \gamma\alpha z_\delta + (1-\gamma\alpha)y$ . We derive from (10.4) that

$$f(z_\delta) \leq \delta f(x) + (1-\delta)f(y) - \delta(1-\delta)\varphi(\gamma t) \quad (10.8)$$

and that

$$f(z_\alpha) \leq \gamma\alpha f(z_\delta) + (1-\gamma\alpha)f(y) - \gamma\alpha(1-\gamma\alpha)\varphi(t). \quad (10.9)$$

Furthermore, (10.7) is equivalent to

$$f(z_\alpha) > \alpha f(x) + (1-\alpha)f(y) - \alpha(1-\alpha)\varphi(\gamma t) - \varepsilon\alpha(1-\alpha). \quad (10.10)$$

Combining (10.8)–(10.10), we deduce that

$$\gamma^2 \varphi(t) < \varphi(\gamma t) + \frac{\varepsilon\gamma(1-\alpha)}{1-\gamma\alpha}. \quad (10.11)$$

However, since  $\alpha \in ]0, 1/2]$  and  $\gamma \in ]1, 2[$ , we have  $(1-\alpha)/(1-\gamma\alpha) \leq 1/(2-\gamma)$ . Thus,  $\gamma^2 \varphi(t) < \varphi(\gamma t) + \varepsilon\gamma/(2-\gamma)$  and, letting  $\varepsilon \downarrow 0$ , we obtain (10.6).

(b)  $\gamma \in [2, +\infty[$ : We write  $\gamma$  as a product of finitely many factors in  $]1, 2[$  and invoke (a) repeatedly to obtain (10.6).  $\square$

**Corollary 10.11** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, with exact modulus of convexity  $\varphi$ . Then  $f$  is uniformly convex if and only if  $\varphi$  vanishes only at 0; in this case,  $f$  is uniformly convex with modulus  $\varphi$ .*

*Proof.* Suppose that  $f$  is uniformly convex with modulus  $\phi$ . Then

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ \phi(\|x - y\|) \leq \frac{\alpha f(x) + (1 - \alpha)f(y) - f(\alpha x + (1 - \alpha)y)}{\alpha(1 - \alpha)} \quad (10.12)$$

and hence  $\phi \leq \varphi$ . Since  $\phi$  vanishes only at 0, so does  $\varphi$ , since Proposition 10.10 asserts that  $\varphi(0) = 0$ . Conversely, if  $\varphi$  vanishes only at 0, then Proposition 10.10 implies that  $f$  is uniformly convex with modulus  $\varphi$ .  $\square$

**Proposition 10.12** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, with exact modulus of convexity  $\varphi$ , and set*

$$\psi: \mathbb{R}_+ \rightarrow [0, +\infty] \\ t \mapsto \inf_{\substack{x \in \text{dom } f, y \in \text{dom } f \\ \|x - y\| = t}} \frac{1}{2}f(x) + \frac{1}{2}f(y) - f\left(\frac{1}{2}x + \frac{1}{2}y\right). \quad (10.13)$$

*Then the following hold:*

- (i)  $2\psi \leq \varphi \leq 4\psi$ .
- (ii)  $f$  is uniformly convex if and only if  $\psi$  vanishes only at 0.

*Proof.* (i): Let  $t \in \mathbb{R}_+$ . Since  $\varphi(0) = \psi(0) = 0$ , we assume that  $\text{dom } f$  is not a singleton and that  $t > 0$ . Take  $\alpha \in ]0, 1/2]$  and two points  $x_0$  and  $y_0$  in  $\text{dom } f$  such that  $\|x_0 - y_0\| = t$  (if no such points exist then  $\varphi(t) = \psi(t) = +\infty$  and hence (i) holds at  $t$ ). By convexity and (10.13),

$$\begin{aligned} f(\alpha x_0 + (1 - \alpha)y_0) &= f\left(2\alpha\left(\frac{1}{2}x_0 + \frac{1}{2}y_0\right) + (1 - 2\alpha)y_0\right) \\ &\leq 2\alpha f\left(\frac{1}{2}x_0 + \frac{1}{2}y_0\right) + (1 - 2\alpha)f(y_0) \\ &\leq 2\alpha\left(\frac{1}{2}f(x_0) + \frac{1}{2}f(y_0) - \psi(\|x_0 - y_0\|)\right) + (1 - 2\alpha)f(y_0) \\ &\leq \alpha f(x_0) + (1 - \alpha)f(y_0) - 2\alpha(1 - \alpha)\psi(t). \end{aligned} \quad (10.14)$$

Hence, Definition 10.9 yields  $2\psi(t) \leq \varphi(t)$ . On the other hand,

$$\varphi(t) \leq \inf_{\substack{x \in \text{dom } f, y \in \text{dom } f \\ \|x - y\| = t}} \frac{\frac{1}{2}f(x) + \frac{1}{2}f(y) - f\left(\frac{1}{2}x + \frac{1}{2}y\right)}{\frac{1}{4}} = 4\psi(t). \quad (10.15)$$

Altogether, (i) is verified.

(ii): In view of (i),  $\varphi$  and  $\psi$  vanish at the same points. Hence (ii) follows from Corollary 10.11.  $\square$

**Proposition 10.13** *Let  $g: \mathcal{H} \rightarrow \mathbb{R}_+$  be uniformly convex, with exact modulus of convexity  $\varphi$ , and let  $p \in [1, +\infty[$ . Then  $g^p$  is uniformly convex, and its exact modulus of convexity  $\chi$  satisfies*

$$\chi \geq 2^{1-2p} \min \{p2^{1-p}, 1 - 2^{-p}\} \varphi^p. \quad (10.16)$$

*Proof.* Let  $t \in \mathbb{R}_+$ , let  $x \in \text{dom } g^p = \text{dom } g = \mathcal{H}$ , and let  $y \in \mathcal{H}$ . We assume that  $t > 0$  and that  $\|x - y\| = t$ . Now set  $\alpha = \frac{1}{2}g(x) + \frac{1}{2}g(y)$ ,  $\beta = g(\frac{1}{2}x + \frac{1}{2}y)$ , and  $\gamma = \varphi(t)/4$ . Since Corollary 10.11 asserts that  $g$  is uniformly convex with modulus  $\varphi$ , we have

$$\alpha \geq \alpha - \beta \geq \gamma > 0. \quad (10.17)$$

If  $\beta \leq \gamma/2$ , then  $\beta^p \leq \gamma^p 2^{-p}$  and  $\alpha^p \geq \gamma^p$ , so that  $\alpha^p - \beta^p \geq \gamma^p(1 - 2^{-p})$ . On the other hand, if  $\beta > \gamma/2$ , the mean value theorem yields  $\alpha^p - \beta^p \geq p\beta^{p-1}(\alpha - \beta) > p(\gamma/2)^{p-1}\gamma = p2^{1-p}\gamma^p$ . Altogether, we always have  $\alpha^p - \beta^p \geq \gamma^p \min\{p2^{1-p}, 1 - 2^{-p}\}$ . Thus, since  $|\cdot|^p$  is convex by Example 8.21,

$$\begin{aligned} \frac{1}{2}g^p(x) + \frac{1}{2}g^p(y) - g^p\left(\frac{1}{2}x + \frac{1}{2}y\right) &\geq \left(\frac{1}{2}g(x) + \frac{1}{2}g(y)\right)^p - g^p\left(\frac{1}{2}x + \frac{1}{2}y\right) \\ &\geq 2^{-2p}\varphi^p(t) \min \{p2^{1-p}, 1 - 2^{-p}\}. \end{aligned} \quad (10.18)$$

Hence, (10.16) follows from Proposition 10.12(i).  $\square$

**Example 10.14** Let  $p \in [2, +\infty[$ . Then  $\|\cdot\|^p$  is uniformly convex.

*Proof.* By Example 10.7(ii),  $\|\cdot\|^2$  is strongly convex with constant 2, hence uniformly convex with modulus  $|\cdot|^2$ . Since  $p/2 \in [1, +\infty[$ , Proposition 10.13 implies that  $\|\cdot\|^p = (\|\cdot\|^2)^{p/2}$  is uniformly convex.  $\square$

**Proposition 10.15** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $C$  be a nonempty compact convex subset of  $\text{dom } f$  such that  $f$  is strictly convex on  $C$  and  $f|_C$  is continuous. Then  $f$  is uniformly convex on  $C$ .*

*Proof.* Set  $g = f + \iota_C$ , define  $\psi$  for  $g$  as in (10.13), and take  $t \in \mathbb{R}_+$  such that  $\psi(t) = 0$ . Then there exist sequences  $(x_n)_{n \in \mathbb{N}}$  and  $(y_n)_{n \in \mathbb{N}}$  in  $\text{dom } g = C$  such that  $\|x_n - y_n\| \equiv t$  and

$$\frac{1}{2}g(x_n) + \frac{1}{2}g(y_n) - g\left(\frac{1}{2}x_n + \frac{1}{2}y_n\right) \rightarrow 0. \quad (10.19)$$

Invoking the compactness of  $C$  and after passing to subsequences if necessary, we assume that there exist  $x \in C$  and  $y \in C$  such that  $x_n \rightarrow x$  and  $y_n \rightarrow y$ . Hence  $\|x - y\| = t$ , and since  $g|_C = f|_C$  is continuous, (10.19) yields  $\frac{1}{2}g(x) + \frac{1}{2}g(y) = g(\frac{1}{2}x + \frac{1}{2}y)$ . In turn, the strict convexity of  $g$  forces  $x = y$ , i.e.,  $t = 0$ . Thus, the result follows from Proposition 10.12(ii).  $\square$

**Corollary 10.16** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be strictly convex, and let  $C$  be a nonempty bounded closed convex subset of  $\mathcal{H}$ . Then  $f$  is uniformly convex on  $C$ .*

*Proof.* Combine Corollary 8.31 and Proposition 10.15.  $\square$

The following example illustrates the importance of the hypotheses in Proposition 10.15 and Corollary 10.16.

**Example 10.17** Suppose that  $\mathcal{H} = \mathbb{R}^2$  and set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi, \eta) \mapsto \begin{cases} 0, & \text{if } \xi = \eta = 0; \\ \frac{\eta^2}{2\xi} + \eta^2, & \text{if } \xi > 0 \text{ and } \eta > 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (10.20)$$

Furthermore, fix  $\rho \in \mathbb{R}_{++}$  and set  $C = B(0; \rho) \cap \text{dom } f$ . Then  $f$  is strictly convex,  $C$  is a nonempty bounded convex subset of  $\text{dom } f$ ,  $f$  is strictly convex on  $C$ , and  $f|_C$  is lower semicontinuous. However,  $f$  is not uniformly convex on  $C$ .

*Proof.* Set  $g = f + \iota_C$ . We verify here only the lack of uniform convexity of  $g$  since the other properties follow from those established in [219, p. 253]. For every  $\eta \in ]0, \rho[$ , if we set  $z_\eta = (\sqrt{\rho^2 - \eta^2}, \eta)$ , then  $\|z_\eta\| = \rho$ ,  $z_\eta \in C$ , and

$$(\forall \alpha \in ]0, 1[) \quad \frac{\alpha g(z_\eta) + (1 - \alpha)g(0) - g(\alpha z_\eta + (1 - \alpha)0)}{\alpha(1 - \alpha)} = \eta^2. \quad (10.21)$$

Denoting the exact modulus of convexity of  $g$  by  $\varphi$ , we deduce that  $0 \leq \varphi(\rho) \leq \inf_{\eta \in ]0, \rho[} \eta^2 = 0$ . Hence  $\varphi(\rho) = 0$ , and in view of Corollary 10.11, we conclude that  $g$  is not uniformly convex.  $\square$

### 10.3 Quasiconvexity

**Definition 10.18** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is *quasiconvex* if its lower level sets  $(\text{lev}_{\leq \xi} f)_{\xi \in \mathbb{R}}$  are convex.

**Example 10.19** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be convex. Then  $f$  is quasiconvex.

*Proof.* A direct consequence of Corollary 8.5.  $\square$

**Example 10.20** Let  $f: \mathbb{R} \rightarrow [-\infty, +\infty]$  be increasing or decreasing. Then  $f$  is quasiconvex.

**Example 10.21** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be quasiconvex and let  $\eta \in \mathbb{R}$ . Then  $\min\{f, \eta\}$  is quasiconvex.

*Proof.* Set  $g = \min\{f, \eta\}$  and let  $\xi \in \mathbb{R}$ . If  $\xi \geq \eta$ , then  $\text{lev}_{\leq \xi} g = \mathcal{H}$  is convex. On the other hand, if  $\xi < \eta$ , then  $\text{lev}_{\leq \xi} g = \text{lev}_{\leq \xi} f$  is also convex.  $\square$

**Proposition 10.22** *Let  $(f_i)_{i \in I}$  be a family of quasiconvex functions from  $\mathcal{H}$  to  $[-\infty, +\infty]$ . Then  $\sup_{i \in I} f_i$  is quasiconvex.*

*Proof.* Let  $\xi \in \mathbb{R}$ . Then  $\text{lev}_{\leq \xi} \sup_{i \in I} f_i = \bigcap_{i \in I} \text{lev}_{\leq \xi} f_i$  is convex as an intersection of convex sets.  $\square$

**Proposition 10.23** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be quasiconvex. Then the following are equivalent:*

- (i)  *$f$  is weakly sequentially lower semicontinuous.*
- (ii)  *$f$  is sequentially lower semicontinuous.*
- (iii)  *$f$  is lower semicontinuous.*
- (iv)  *$f$  is weakly lower semicontinuous.*

*Proof.* Since the sets  $(\text{lev}_{\leq \xi} f)_{\xi \in \mathbb{R}}$  are convex, the equivalences follow from Lemma 1.24, Lemma 1.35, and Theorem 3.32.  $\square$

The following proposition is clear from Definition 10.18.

**Proposition 10.24** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is quasiconvex if and only if*

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) \leq \max \{f(x), f(y)\}. \quad (10.22)$$

We now turn our attention to strict versions of quasiconvexity suggested by (10.22).

**Definition 10.25** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $f$  is

- (i) *strictly quasiconvex* if

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ x \neq y \Rightarrow f(\alpha x + (1 - \alpha)y) < \max \{f(x), f(y)\}; \quad (10.23)$$

- (ii) *uniformly quasiconvex* with modulus  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  if  $\phi$  is increasing,  $\phi$  vanishes only at 0, and

$$(\forall x \in \text{dom } f)(\forall y \in \text{dom } f)(\forall \alpha \in ]0, 1[) \\ f(\alpha x + (1 - \alpha)y) + \alpha(1 - \alpha)\phi(\|x - y\|) \leq \max \{f(x), f(y)\}. \quad (10.24)$$

**Remark 10.26** Each type of quasiconvexity in Definition 10.25 is implied by its convex counterpart, and uniform quasiconvexity implies strict quasiconvexity. As examples in the remainder of this chapter show, these notions are all distinct.

**Example 10.27** Let  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be proper, and strictly increasing or strictly decreasing on  $\text{dom } f$ . Then it follows from (10.23) that  $f$  is strictly quasiconvex.

**Example 10.28** Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  and let  $\phi: \text{ran } f \rightarrow \mathbb{R}$  be increasing. Then the following hold:

- (i) Suppose that  $f$  is strictly quasiconvex and that  $\phi$  is strictly increasing. Then  $\phi \circ f$  is strictly quasiconvex.
- (ii) Suppose that  $\phi \circ f$  is strictly convex. Then  $f$  is strictly quasiconvex.

*Proof.* Assume that  $x$  and  $y$  are distinct points in  $\mathcal{H} = \text{dom } f = \text{dom}(\phi \circ f)$ , and let  $\alpha \in ]0, 1[$ .

- (i): In view of (10.23),  $f(\alpha x + (1 - \alpha)y) < \max\{f(x), f(y)\}$ . Therefore

$$\begin{aligned} (\phi \circ f)(\alpha x + (1 - \alpha)y) &= \phi(f(\alpha x + (1 - \alpha)y)) \\ &< \phi(\max\{f(x), f(y)\}) \\ &= \max\{(\phi \circ f)(x), (\phi \circ f)(y)\}. \end{aligned} \quad (10.25)$$

- (ii): We derive from (8.3) that

$$\begin{aligned} \phi(f(\alpha x + (1 - \alpha)y)) &= (\phi \circ f)(\alpha x + (1 - \alpha)y) \\ &< \alpha(\phi \circ f)(x) + (1 - \alpha)(\phi \circ f)(y) \\ &\leq \max\{(\phi \circ f)(x), (\phi \circ f)(y)\} \\ &= \phi(\max\{f(x), f(y)\}). \end{aligned} \quad (10.26)$$

Hence,  $f(\alpha x + (1 - \alpha)y) < \max\{f(x), f(y)\}$ .  $\square$

**Example 10.29** Let  $p \in \mathbb{R}_{++}$ . Then the following hold:

- (i)  $\|\cdot\|^p$  is strictly quasiconvex.
- (ii) Suppose that  $\mathcal{H} \neq \{0\}$  and that  $p < 1$ . Then  $\|\cdot\|^p$  is not convex and not uniformly quasiconvex.

*Proof.* (i): Take  $f = \|\cdot\|^2$  (which is strictly quasiconvex by Example 8.8) and  $\phi: \mathbb{R}_+ \rightarrow \mathbb{R}: t \mapsto t^{p/2}$  in Example 10.28(i).

(ii): Let  $z \in \mathcal{H}$  be such that  $\|z\| = 1$ . If we set  $x = z$  and  $y = 0$ , and let  $\alpha \in ]0, 1[$ , then (8.1) turns into the false inequality  $\alpha^p \leq \alpha$ . Thus, by Proposition 8.4,  $f$  is not convex. Now assume that  $\|\cdot\|^p$  is uniformly quasiconvex with modulus  $\phi$ , let  $t \in \mathbb{R}_{++}$ , and let  $s \in ]t, +\infty[$ . Setting  $x = (s - t)z$ ,  $y = (s + t)z$ , and  $\alpha = 1/2$ , we estimate

$$\begin{aligned} \frac{\phi(2t)}{4} &\leq \max\{\|(s - t)z\|^p, \|(s + t)z\|^p\} - \left\|\frac{1}{2}(s - t)z + \frac{1}{2}(s + t)z\right\|^p \\ &= (s + t)^p - s^p. \end{aligned} \quad (10.27)$$

Letting  $s \uparrow +\infty$ , we deduce that  $\phi(2t) = 0$ , which is impossible.  $\square$

**Example 10.30** Suppose that  $\mathcal{H} = \mathbb{R}$  and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \nu x$ , where  $\nu \in \mathbb{R}_{++}$ . Then  $f$  is convex, not strictly convex, but uniformly quasiconvex with modulus  $\phi: t \mapsto \nu t$ .



*Proof.* Take  $x$  and  $y$  in  $\mathbb{R}$  such that  $x < y$ , and take  $\alpha \in ]0, 1[$ . Since  $f(\alpha x + (1 - \alpha)y) = \alpha f(x) + (1 - \alpha)f(y)$ ,  $f$  is convex but not strictly so. The uniform quasiconvexity follows from the identity  $f(\alpha x + (1 - \alpha)y) = \nu y + \alpha\nu(x - y) = \max\{f(x), f(y)\} - \alpha\nu|x - y| \leq \max\{f(x), f(y)\} - \alpha(1 - \alpha)\nu|x - y|$ .  $\square$

## Exercises

**Exercise 10.1** Prove Proposition 10.2.

**Exercise 10.2** Prove Proposition 10.3.

**Exercise 10.3** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Show that  $f$  is sublinear if and only if  $\text{epi } f$  is a convex cone.

**Exercise 10.4** Check Example 10.7.

**Exercise 10.5** Suppose that  $\mathcal{H} \neq \{0\}$  and let  $p \in ]1, 2[$ . Show that  $\|\cdot\|^p$  is not uniformly convex.

**Exercise 10.6** Show that

$$f: \mathbb{R} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi, x) \mapsto \begin{cases} \frac{\xi^4 + \xi^2\|x\|^2 + \|x\|^3}{\xi^2}, & \text{if } \xi > 0; \\ +\infty, & \text{otherwise,} \end{cases} \quad (10.28)$$

is strongly convex.

**Exercise 10.7** Suppose that  $\mathcal{H} = \mathbb{R}$  and set  $f = |\cdot|^4$ . Show that  $f$  is not strongly convex. Determine the function  $\psi$  defined in (10.13) explicitly and conclude from the same result that  $f$  is uniformly convex.

**Exercise 10.8** Use Exercise 8.9 to show that the function  $\psi$  defined in (10.13) is increasing.

**Exercise 10.9** Let  $f \in \Gamma_0(\mathbb{R})$ , and let  $C$  be a nonempty bounded closed interval in  $\text{dom } f$  such that  $f$  is strictly convex on  $C$ . Show that  $f$  is uniformly convex on  $C$ .

**Exercise 10.10** Show that none of the following functions from  $\Gamma_0(\mathbb{R})$  is uniformly convex:

(i) The negative Boltzmann–Shannon entropy function (Example 9.29).

(ii)  $x \mapsto \begin{cases} 1/x, & \text{if } x > 0; \\ +\infty, & \text{if } x \leq 0. \end{cases}$

(iii) The negative Burg entropy function (Example 9.30(viii)).

**Exercise 10.11** Set

$$f: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto \begin{cases} 2x, & \text{if } x \leq 0; \\ x, & \text{if } x > 0. \end{cases} \quad (10.29)$$

Show that  $f$  is not convex but that it is uniformly quasiconvex.

**Exercise 10.12** Suppose that  $\mathcal{H} \neq \{0\}$  and set

$$f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \frac{\|x\|}{\|x\| + 1}. \quad (10.30)$$

Show that  $f$  is not convex, not uniformly quasiconvex, but strictly quasiconvex.

**Exercise 10.13** Set

$$\begin{aligned} f: \mathbb{R} &\rightarrow ]-\infty, +\infty] \\ x &\mapsto \begin{cases} \frac{x}{n+1} + \sum_{k=2}^{n+1} \frac{1}{k}, & \text{if } (\exists n \in \mathbb{N}) \ x \in [n, n+1[; \\ +\infty, & \text{if } x < 0. \end{cases} \end{aligned} \quad (10.31)$$

Then  $f$  is not convex, not uniformly quasiconvex, but strictly quasiconvex.

**Exercise 10.14** Show that the sum of a quasiconvex function and a convex function may not be quasiconvex.

**Exercise 10.15** Let  $\mathcal{K}$  be a real Hilbert space and let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$  be quasiconvex. Show that the marginal function  $f: \mathcal{H} \rightarrow [-\infty, +\infty]: x \mapsto \inf F(x, \mathcal{K})$  is quasiconvex.

**Exercise 10.16** Suppose that  $\mathcal{H} \neq \{0\}$ . Provide a function that is convex, but not strictly quasiconvex.

**Exercise 10.17** Show that the converse of Example 10.28(ii) is false by providing a function  $f: \mathbb{R} \rightarrow \mathbb{R}_+$  that is strictly quasiconvex, but such that  $f^2$  is not strictly convex.

**Exercise 10.18** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $x \in \text{dom } f$ . The *recession function* of  $f$  is defined by

$$(\forall y \in \mathcal{H}) \quad (\text{rec } f)(y) = \lim_{\alpha \rightarrow +\infty} \frac{f(x + \alpha y) - f(x)}{\alpha}. \quad (10.32)$$

Prove the following:

- (i) For every  $y \in \mathcal{H}$ , the limit in (10.32) exists and is independent of  $x$ .
- (ii)  $(\forall y \in \mathcal{H}) \quad (\text{rec } f)(y) = \sup_{\alpha \in \mathbb{R}_{++}} (f(x + \alpha y) - f(x))/\alpha$ .

- (iii)  $(\forall y \in \mathcal{H}) \text{ (rec } f)(y) = \sup_{z \in \text{dom } f} (f(z + y) - f(z)).$
- (iv)  $\text{epi rec } f = \text{rec epi } f.$
- (v)  $\text{rec } f$  is proper, lower semicontinuous, and sublinear.
- (vi) Suppose that  $\inf f(\mathcal{H}) > -\infty.$  Then  $(\forall y \in \mathcal{H}) \text{ (rec } f)(y) \geq 0.$



# Chapter 11

## Convex Variational Problems

Convex optimization is one of the main areas of application of convex analysis. This chapter deals with the issues of existence and uniqueness in minimization problems, and investigates properties of minimizing sequences.

### 11.1 Infima and Suprema

**Proposition 11.1** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  and let  $C$  be a nonempty subset of  $\mathcal{H}$ . Then the following hold:*

- (i) *Suppose that  $f$  is lower semicontinuous. Then  $\sup f(\overline{C}) = \sup f(C)$ .*
- (ii) *Suppose that  $f$  is convex. Then  $\sup f(\operatorname{conv} C) = \sup f(C)$ .*
- (iii) *Let  $u \in \mathcal{H}$ . Then  $\sup \langle \overline{\operatorname{conv} C} \mid u \rangle = \sup \langle C \mid u \rangle$  and  $\inf \langle \overline{\operatorname{conv} C} \mid u \rangle = \inf \langle C \mid u \rangle$ .*
- (iv) *Suppose that  $f \in \Gamma_0(\mathcal{H})$ , that  $C$  is convex, and that  $\operatorname{dom} f \cap \operatorname{int} C \neq \emptyset$ . Then  $\inf f(\overline{C}) = \inf f(C)$ .*

*Proof.* (i): Since  $C \subset \overline{C}$ , we have  $\sup f(C) \leq \sup f(\overline{C})$ . Now take  $x \in \overline{C}$ . Then there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $C$  such that  $x_n \rightarrow x$ . Thus  $f(x) \leq \liminf f(x_n) \leq \sup f(C)$ , and we conclude that  $\sup f(\overline{C}) \leq \sup f(C)$ .

(ii): Since  $C \subset \operatorname{conv} C$ , we have  $\sup f(C) \leq \sup f(\operatorname{conv} C)$ . Now take  $x \in \operatorname{conv} C$ , say  $x = \sum_{i \in I} \alpha_i x_i$ , where  $(\alpha_i)_{i \in I}$  is a finite family in  $]0, 1[$  that satisfies  $\sum_{i \in I} \alpha_i = 1$ , and where  $(x_i)_{i \in I}$  is a finite family in  $C$ . Then, since  $f$  is convex, Corollary 8.10 yields  $f(x) = f(\sum_{i \in I} \alpha_i x_i) \leq \sum_{i \in I} \alpha_i f(x_i) \leq \sum_{i \in I} \alpha_i \sup f(C) = \sup f(C)$ . Therefore,  $\sup f(\operatorname{conv} C) \leq \sup f(C)$ .

(iii): This follows from (i) and (ii), since  $\langle \cdot \mid u \rangle$  and  $-\langle \cdot \mid u \rangle$  are continuous and convex.

(iv): It is clear that  $\inf f(\overline{C}) \leq \inf f(C)$ . Now take  $x_0 \in \overline{C}$ ,  $x_1 \in \operatorname{dom} f \cap \operatorname{int} C$ , and set  $(\forall \alpha \in ]0, 1[) x_\alpha = (1 - \alpha)x_0 + \alpha x_1$ . Using Proposition 9.14 and Proposition 3.35, we deduce that  $f(x_0) = \lim_{\alpha \downarrow 0} f(x_\alpha) \geq \inf f(C)$ . Therefore,  $\inf f(\overline{C}) \geq \inf f(C)$ .  $\square$

**Example 11.2** Let  $C$  be a nonempty subset of  $\mathcal{H}$ . Then the support function  $\sigma_C$  is a sublinear function in  $\Gamma_0(\mathcal{H})$ ,  $\sigma_C = \sigma_{\overline{\text{conv}} C}$ , and  $\text{dom } \sigma_C = \text{bar } \overline{\text{conv}} C$ . If  $C$  is bounded, then  $\sigma_C$  is real-valued and continuous on  $\mathcal{H}$ .

*Proof.* Definition 7.8 implies that  $\sigma_C(0) = 0$  and that  $\sigma_C$  is the supremum of the family of (continuous, hence) lower semicontinuous and (linear, hence) convex functions  $(\langle x | \cdot \rangle)_{x \in C}$ . Therefore,  $\sigma_C$  is lower semicontinuous and convex by Proposition 9.3. On the other hand, it is clear from Definition 7.8 that  $\sigma_C$  is positively homogeneous. Altogether,  $\sigma_C$  is sublinear. Furthermore, Proposition 11.1(iii) (alternatively, Proposition 7.11) implies that  $\sigma_C = \sigma_{\overline{\text{conv}} C}$  and hence that  $\text{dom } \sigma_C = \text{bar } \overline{\text{conv}} C$  by (6.39). If  $C$  is bounded, then  $\text{bar } \overline{\text{conv}} C = \mathcal{H}$  by Proposition 6.48(iii), and the continuity of  $\sigma_C$  is a consequence of Corollary 8.30(ii).  $\square$

## 11.2 Minimizers

**Definition 11.3** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, and let  $x \in \mathcal{H}$ . Then  $x$  is a *minimizer* of  $f$  if  $f(x) = \inf f(\mathcal{H})$ , i.e. (see Section 1.5),  $f(x) = \min f(\mathcal{H}) \in \mathbb{R}$ . The set of minimizers of  $f$  is denoted by  $\text{Argmin } f$ . Now let  $C$  be a subset of  $\mathcal{H}$  such that  $C \cap \text{dom } f \neq \emptyset$ . A *minimizer of  $f$  over  $C$*  is a minimizer of  $f + \iota_C$ . The set of minimizers of  $f$  over  $C$  is denoted by  $\text{Argmin}_C f$ . If there exists  $\rho \in \mathbb{R}_{++}$  such that  $x$  is a minimizer of  $f$  over  $B(x; \rho)$ , then  $x$  is a *local minimizer* of  $f$ .

The following result underlines the fundamental importance of convexity in minimization problems.

**Proposition 11.4** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Then every local minimizer of  $f$  is a minimizer.

*Proof.* Take  $x \in \mathcal{H}$  and  $\rho \in \mathbb{R}_{++}$  such that  $f(x) = \min f(B(x; \rho))$ . Fix  $y \in \text{dom } f \setminus B(x; \rho)$ , and set  $\alpha = 1 - \rho/\|x - y\|$  and  $z = \alpha x + (1 - \alpha)y$ . Then  $\alpha \in ]0, 1[$  and  $z \in B(x; \rho)$ . In view of the convexity of  $f$ , we deduce that

$$f(x) \leq f(z) = f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y). \quad (11.1)$$

Therefore  $f(x) \leq f(y)$ .  $\square$

**Proposition 11.5** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $C$  be a subset of  $\mathcal{H}$ . Suppose that  $x$  is a minimizer of  $f$  over  $C$  such that  $x \in \text{int } C$ . Then  $x$  is a minimizer of  $f$ .

*Proof.* There exists  $\rho \in \mathbb{R}_{++}$  such that  $B(x; \rho) \subset C$ . Therefore,  $f(x) = \inf f(B(x; \rho))$ , and the conclusion follows from Proposition 11.4.  $\square$

**Proposition 11.6** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and quasiconvex. Then  $\text{Argmin } f$  is convex.*

*Proof.* This follows from Definition 10.18.  $\square$

### 11.3 Uniqueness of Minimizers

**Proposition 11.7** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be quasiconvex and let  $C$  be a convex subset of  $\mathcal{H}$  such that  $C \cap \text{dom } f \neq \emptyset$ . Suppose that one of the following holds:*

- (i)  $f + \iota_C$  is strictly quasiconvex.
- (ii)  $f$  is convex,  $C \cap \text{Argmin } f = \emptyset$ , and  $C$  is strictly convex, i.e.,

$$(\forall x \in C)(\forall y \in C) \quad x \neq y \Rightarrow \frac{x+y}{2} \in \text{int } C. \quad (11.2)$$

*Then  $f$  has at most one minimizer over  $C$ .*

*Proof.* We assume that  $C$  is not a singleton. Set  $\mu = \inf f(C)$  and suppose that there exist two distinct points  $x$  and  $y$  in  $C \cap \text{dom } f$  such that  $f(x) = f(y) = \mu$ . Since  $x$  and  $y$  lie in the convex set  $C \cap \text{lev}_{\leq \mu} f$ , so does  $z = (x+y)/2$ . Therefore  $f(z) = \mu$ .

(i): It follows from the strict quasiconvexity of  $f + \iota_C$  that  $\mu = f(z) < \max\{f(x), f(y)\} = \mu$ , which is impossible.

(ii): We have  $z \in \text{int } C$  and  $f(z) = \inf f(C)$ . Since  $f$  is convex, it follows from Proposition 11.5 that  $f(z) = \inf f(\mathcal{H})$ . Therefore,  $z \in C \cap \text{Argmin } f = \emptyset$ , which is absurd.  $\square$

**Corollary 11.8** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and strictly convex. Then  $f$  has at most one minimizer.*

*Proof.* Since  $f$  is strictly quasiconvex, the result follows from Proposition 11.7(i) with  $C = \mathcal{H}$ .  $\square$

### 11.4 Existence of Minimizers

**Theorem 11.9** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be lower semicontinuous and quasiconvex, and let  $C$  be a closed convex subset of  $\mathcal{H}$  such that, for some  $\xi \in \mathbb{R}$ ,  $C \cap \text{lev}_{\leq \xi} f$  is nonempty and bounded. Then  $f$  has a minimizer over  $C$ .*

*Proof.* Since  $f$  is lower semicontinuous and quasiconvex, it follows from Proposition 10.23 that  $f$  is weakly lower semicontinuous. On the other hand, since  $C$  and  $\text{lev}_{\leq \xi} f$  are closed and convex, the set  $D = C \cap \text{lev}_{\leq \xi} f$  is closed and

convex, and, by assumption, bounded. Thus,  $D$  is nonempty and weakly compact by Theorem 3.33. Consequently, since minimizing  $f$  over  $C$  is equivalent to minimizing  $f$  over  $D$ , the claim follows from Lemma 2.23 and Theorem 1.28 in  $\mathcal{H}^{\text{weak}}$ .  $\square$

**Definition 11.10** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is *coercive* if

$$\lim_{\|x\| \rightarrow +\infty} f(x) = +\infty, \quad (11.3)$$

and *supercoercive* if

$$\lim_{\|x\| \rightarrow +\infty} \frac{f(x)}{\|x\|} = +\infty. \quad (11.4)$$

By convention,  $f$  is coercive and supercoercive if  $\mathcal{H} = \{0\}$ .

**Proposition 11.11** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f$  is coercive if and only if its lower level sets  $(\text{lev}_{\leq \xi} f)_{\xi \in \mathbb{R}}$  are bounded.

*Proof.* Suppose that, for some  $\xi \in \mathbb{R}$ ,  $\text{lev}_{\leq \xi} f$  is unbounded. Then we can find a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\text{lev}_{\leq \xi} f$  such that  $\|x_n\| \rightarrow +\infty$ . As a result,  $f$  is not coercive. Conversely, suppose that the lower level sets of  $f$  are bounded and take a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{H}$  such that  $\|x_n\| \rightarrow +\infty$ . Then, for every  $\xi \in \mathbb{R}_{++}$ , we can find  $N \in \mathbb{N}$  such that  $\inf_{n \geq N} f(x_n) \geq \xi$ . Therefore  $f(x_n) \rightarrow +\infty$ .  $\square$

**Proposition 11.12** Suppose that  $\mathcal{H}$  is finite-dimensional and let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f$  is coercive if and only if there exists  $\xi \in \mathbb{R}$  such that  $\text{lev}_{\leq \xi} f$  is nonempty and bounded.

*Proof.* If  $f$  is coercive, then all level sets  $(\text{lev}_{\leq \xi} f)_{\xi \in \mathbb{R}}$  are bounded by Proposition 11.11. Now suppose that  $\text{lev}_{\leq \xi} f$  is nonempty and bounded, and take  $x \in \text{lev}_{\leq \xi} f$ . It is clear that all lower level sets at a lower height are bounded. Take  $\eta \in ]\xi, +\infty[$  and suppose that  $\text{lev}_{\leq \eta} f$  is unbounded. By Corollary 6.51,  $\text{rec lev}_{\leq \eta} f \neq \{0\}$ . Take  $y \in \text{rec lev}_{\leq \eta} f$ . Since  $x \in \text{lev}_{\leq \eta} f$ , it follows that  $(\forall \lambda \in \mathbb{R}_{++}) x + \lambda y \in \text{lev}_{\leq \eta} f$ . For every  $\lambda \in ]1, +\infty[$ , we have  $x + y = (1 - \lambda^{-1})x + \lambda^{-1}(x + \lambda y)$  and hence  $f(x + y) \leq (1 - \lambda^{-1})f(x) + \lambda^{-1}f(x + \lambda y) \leq (1 - \lambda^{-1})f(x) + \lambda^{-1}\eta$ , which implies that  $\lambda(f(x + y) - f(x)) \leq \eta - f(x)$  and, in turn, that  $f(x + y) \leq f(x) \leq \xi$ . We conclude that  $x + \text{rec lev}_{\leq \eta} f \subset \text{lev}_{\leq \xi} f$ , which is impossible, since  $\text{lev}_{\leq \xi} f$  is bounded and  $\text{rec lev}_{\leq \eta} f$  is unbounded. Therefore, all lower level sets  $(\text{lev}_{\leq \eta} f)_{\eta \in \mathbb{R}}$  are bounded and Proposition 11.11 implies that  $f$  is coercive.  $\square$

**Proposition 11.13** Let  $f$  be in  $\Gamma_0(\mathcal{H})$ , and let  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be supercoercive. Then  $f + g$  is supercoercive.

*Proof.* According to Theorem 9.19,  $f$  is minorized by a continuous affine functional, say  $x \mapsto \langle x | u \rangle + \eta$ , where  $u \in \mathcal{H}$  and  $\eta \in \mathbb{R}$ . Thus,  $(\forall x \in \mathcal{H}) f(x) + g(x) \geq \langle x | u \rangle + \eta + g(x) \geq -\|x\| \|u\| + \eta + g(x)$ . We conclude that  $(f(x) + g(x))/\|x\| \geq -\|u\| + (\eta + g(x))/\|x\| \rightarrow +\infty$  as  $\|x\| \rightarrow +\infty$ .  $\square$



**Proposition 11.14** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $C$  be a closed convex subset of  $\mathcal{H}$  such that  $C \cap \text{dom } f \neq \emptyset$ . Suppose that one of the following holds:*

- (i)  $f$  is coercive.
- (ii)  $C$  is bounded.

*Then  $f$  has a minimizer over  $C$ .*

*Proof.* Since  $C \cap \text{dom } f \neq \emptyset$ , there exists  $x \in \text{dom } f$  such that  $D = C \cap \text{lev}_{\leq f(x)} f$  is nonempty, closed, and convex. Moreover,  $D$  is bounded since  $C$  or, by Proposition 11.11,  $\text{lev}_{\leq f(x)} f$  is. The result therefore follows from Theorem 11.9.  $\square$

**Corollary 11.15** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Suppose that  $\text{dom } f \cap \text{dom } g \neq \emptyset$  and that one of the following holds:*

- (i)  $f$  is supercoercive.
- (ii)  $f$  is coercive and  $g$  is bounded below.

*Then  $f + g$  is coercive and it has a minimizer over  $\mathcal{H}$ . If  $f$  or  $g$  is strictly convex, then  $f + g$  has exactly one minimizer over  $\mathcal{H}$ .*

*Proof.* It follows from Corollary 9.4 that  $f + g \in \Gamma_0(\mathcal{H})$ . Hence, in view of Proposition 11.14(i), it suffices to show that  $f + g$  is coercive in both cases. The uniqueness of the minimizer assertion will then follow from Corollary 11.8 by observing that  $f + g$  is strictly convex.

(i): By Proposition 11.13,  $f + g$  is supercoercive, hence coercive.

(ii): Set  $\mu = \inf g(\mathcal{H}) > -\infty$ . Then  $(f + g)(x) \geq f(x) + \mu \rightarrow +\infty$  as  $\|x\| \rightarrow +\infty$  by coercivity of  $f$ .  $\square$

**Corollary 11.16** *Let  $f \in \Gamma_0(\mathcal{H})$  be strongly convex. Then  $f$  is supercoercive and it has exactly one minimizer over  $\mathcal{H}$ .*

*Proof.* Set  $q = (1/2)\|\cdot\|^2$ . By Proposition 10.6, there exists  $\beta \in \mathbb{R}_{++}$  such that  $f - \beta q$  is convex. Hence,  $f = \beta q + (f - \beta q)$  is the sum of the supercoercive function  $\beta q$  and  $f - \beta q \in \Gamma_0(\mathcal{H})$ . Therefore,  $f$  is supercoercive by Proposition 11.13. In view of Corollary 11.15,  $f$  has exactly one minimizer.  $\square$

**Proposition 11.17 (asymptotic center)** *Let  $(z_n)_{n \in \mathbb{N}}$  be a bounded sequence in  $\mathcal{H}$ , let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $T: C \rightarrow C$  be nonexpansive, and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \overline{\lim} \|x - z_n\|^2$ . Then the following hold:*

- (i)  $f$  is strongly convex with constant 2.
- (ii)  $f$  is supercoercive.
- (iii)  $f + \iota_C$  is strongly convex and supercoercive; its unique minimizer, denoted by  $z_C$ , is called the asymptotic center of  $(z_n)_{n \in \mathbb{N}}$  relative to  $C$ .
- (iv) Suppose that  $z \in \mathcal{H}$  and that  $z_n \rightharpoonup z$ . Then  $(\forall x \in \mathcal{H}) f(x) = \|x - z\|^2 + f(z)$  and  $z_C = P_C z$ .

- (v) Suppose that  $(z_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $C$ . Then  $P_C z_n \rightarrow z_C$ .
- (vi) Suppose that  $(\forall n \in \mathbb{N}) \ z_{n+1} = Tz_n$ . Then  $z_C \in \text{Fix } T$ .
- (vii) Suppose that  $z_n - Tz_n \rightarrow 0$ . Then  $z_C \in \text{Fix } T$ .

*Proof.* (i): Let  $x \in \mathcal{H}$ , let  $y \in \mathcal{H}$ , and let  $\alpha \in ]0, 1[$ . Corollary 2.14 yields  $(\forall n \in \mathbb{N}) \ \|\alpha x + (1 - \alpha)y - z_n\|^2 = \alpha\|x - z_n\|^2 + (1 - \alpha)\|y - z_n\|^2 - \alpha(1 - \alpha)\|x - y\|^2$ . Now take the limit superior.

(ii)&(iii): By (i),  $f$  is strongly convex with constant 2, as is  $f + \iota_C$ . Hence, the claim follows from Corollary 11.16.

(iv): Let  $x \in \mathcal{H}$ . Then  $(\forall n \in \mathbb{N}) \ \|x - z_n\|^2 = \|x - z\|^2 + \|z - z_n\|^2 + 2\langle x - z | z - z_n \rangle$ . Consequently,  $f(x) = \overline{\lim} \|x - z_n\|^2 = \|x - z\|^2 + f(z)$ , and thus  $P_C z$  minimizes  $f + \iota_C$ . Thus (iii) implies that  $P_C z = z_C$ .

(v): By Proposition 5.7,  $\bar{y} = \lim P_C z_n$  is well defined. For every  $n \in \mathbb{N}$  and every  $y \in C$ ,  $\|\bar{y} - z_n\| \leq \|\bar{y} - P_C z_n\| + \|P_C z_n - z_n\| \leq \|\bar{y} - P_C z_n\| + \|y - z_n\|$ . Hence,  $\overline{\lim} \|\bar{y} - z_n\| \leq \inf_{y \in C} \overline{\lim} \|y - z_n\|$  and  $\bar{y}$  is thus a minimizer of  $f + \iota_C$ . By (iii),  $\bar{y} = z_C$ .

(vi): Observe that  $z_C \in C$  and that  $Tz_C \in C$ . For every  $n \in \mathbb{N}$ ,  $\|Tz_C - z_{n+1}\| = \|Tz_C - Tz_n\| \leq \|z_C - z_n\|$ . Thus, taking the limit superior, we obtain  $(f + \iota_C)(Tz_C) \leq (f + \iota_C)(z_C)$ . By (iii),  $Tz_C = z_C$ .

(vii): For every  $n \in \mathbb{N}$ ,  $\|Tz_C - z_n\| \leq \|Tz_C - Tz_n\| + \|Tz_n - z_n\| \leq \|z_C - z_n\| + \|Tz_n - z_n\|$ . Hence  $\overline{\lim} \|Tz_C - z_n\| \leq \overline{\lim} \|z_C - z_n\|$ , and thus  $(f + \iota_C)(Tz_C) \leq (f + \iota_C)(z_C)$ . Again by (iii),  $Tz_C = z_C$ .  $\square$

**Corollary 11.18** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $T: C \rightarrow C$  be nonexpansive. For every  $z_0 \in C$ , set  $(\forall n \in \mathbb{N}) \ z_{n+1} = Tz_n$ . Then the following are equivalent:*

- (i)  $\text{Fix } T \neq \emptyset$ .
- (ii) For every  $z_0 \in C$ ,  $(z_n)_{n \in \mathbb{N}}$  is bounded.
- (iii) For some  $z_0 \in C$ ,  $(z_n)_{n \in \mathbb{N}}$  is bounded.

*Proof.* (i) $\Rightarrow$ (ii): Combine Example 5.3 with Proposition 5.4(i).

(ii) $\Rightarrow$ (iii): Clear.

(iii) $\Rightarrow$ (i): This follows from Proposition 11.17(vi).  $\square$

## 11.5 Minimizing Sequences

Minimizing sequences were introduced in Definition 1.8. In this section we investigate some of their properties.

**Proposition 11.19** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a coercive proper function. Then every minimizing sequence of  $f$  is bounded.*

*Proof.* This follows at once from Definition 1.8 and Proposition 11.11.  $\square$

**Proposition 11.20** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper lower semicontinuous quasiconvex function and let  $(x_n)_{n \in \mathbb{N}}$  be a minimizing sequence of  $f$  that converges weakly to some point  $x \in \mathcal{H}$ . Then  $f(x) = \inf f(\mathcal{H})$ .*

*Proof.* It follows from Proposition 10.23 that  $f$  is weakly sequentially lower semicontinuous. Hence  $\inf f(\mathcal{H}) \leq f(x) \leq \liminf f(x_n) = \inf f(\mathcal{H})$ .  $\square$

**Remark 11.21** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{dom } f$ .

- (i) Suppose that  $(x_n)_{n \in \mathbb{N}}$  converges strongly to a minimizer  $x$  of  $f$  and that  $\mathcal{H} = \mathbb{R}$  or  $x \in \text{int dom } f$ . Then  $(x_n)_{n \in \mathbb{N}}$  is a minimizing sequence of  $f$ . Indeed, it follows from Corollary 9.15 in the former case, and from Corollary 8.30 in the latter, that  $f(x_n) \rightarrow f(x)$ .
- (ii) Suppose that  $(x_n)_{n \in \mathbb{N}}$  converges strongly to a minimizer  $x$  of  $f$ . Then  $(x_n)_{n \in \mathbb{N}}$  may not be a minimizing sequence of  $f$ , even if  $\mathcal{H}$  is the Euclidean plane (see the construction in the proof of Example 9.27).
- (iii) Suppose that  $\mathcal{H}$  is infinite-dimensional and that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a minimizer  $x$  of  $f$ . Then  $(x_n)_{n \in \mathbb{N}}$  may not be a minimizing sequence of  $f$ , even if  $x \in \text{int dom } f$ . For instance, suppose that  $(x_n)_{n \in \mathbb{N}}$  is an orthonormal sequence in  $\mathcal{H}$  and set  $f = \|\cdot\|$ . Then, as seen in Example 2.25,  $x_n \rightarrow 0$  and  $f(0) = 0 = \inf f(\mathcal{H})$ , while  $f(x_n) \equiv 1$ .

The next examples illustrate various behaviors of minimizing sequences.

**Example 11.22** Suppose that  $x \in \mathcal{H} \setminus \{0\}$ , and set  $f = \iota_{[-x, x]}$  and  $(\forall n \in \mathbb{N}) x_n = (-1)^n x$ . Then  $(x_n)_{n \in \mathbb{N}}$  is a bounded minimizing sequence for  $f$  that is not weakly convergent.

**Example 11.23** Suppose that  $\mathcal{H}$  is separable. Let  $(\omega_k)_{k \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  such that  $\omega_k \rightarrow 0$ , let  $(x_k)_{k \in \mathbb{N}}$  be an orthonormal basis of  $\mathcal{H}$ , and set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty] : x \mapsto \sum_{k \in \mathbb{N}} \omega_k |\langle x | x_k \rangle|^2. \quad (11.5)$$

Then  $f$  is real-valued, continuous, and strictly convex, and 0 is its unique minimizer. However,  $f$  is not coercive. Moreover,  $(x_n)_{n \in \mathbb{N}}$  is a minimizing sequence of  $f$  that converges weakly but not strongly.

*Proof.* Note that  $f$  vanishes only at 0. Moreover, by Parseval,  $(\forall x \in \mathcal{H}) \|x\|^2 = \sum_{k \in \mathbb{N}} |\langle x | x_k \rangle|^2 \geq f(x) / \sup_{k \in \mathbb{N}} \omega_k$ . Hence,  $f$  is real-valued. Furthermore, since the functions  $(\omega_k |\langle \cdot | x_k \rangle|^2)_{k \in \mathbb{N}}$  are positive, convex, and continuous, it follows from Corollary 9.4 that  $f$  is convex and lower semicontinuous. Thus, Corollary 8.30(ii) implies that  $f$  is continuous.

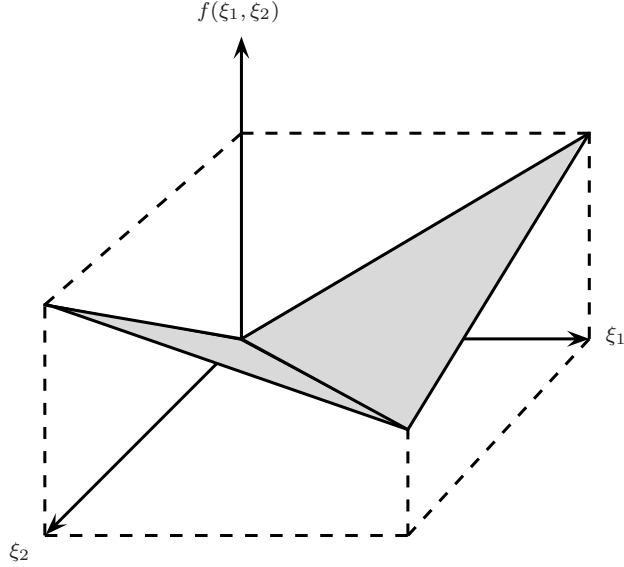
To show that  $f$  is strictly convex, take  $x$  and  $y$  in  $\mathcal{H}$  such that  $x \neq y$ , and fix  $\alpha \in ]0, 1[$ . Then there exists  $m \in \mathbb{N}$  such that  $\sqrt{\omega_m} \langle x | x_m \rangle \neq \sqrt{\omega_m} \langle y | x_m \rangle$ . Since  $|\cdot|^2$  is strictly convex (Example 8.8), we get

$$\omega_m |\langle \alpha x + (1 - \alpha)y | x_m \rangle|^2 < \alpha \omega_m |\langle x | x_m \rangle|^2 + (1 - \alpha) \omega_m |\langle y | x_m \rangle|^2 \quad (11.6)$$

and, for every  $k \in \mathbb{N} \setminus \{m\}$ ,

$$\omega_k |\langle \alpha x + (1 - \alpha)y \mid x_k \rangle|^2 \leq \alpha \omega_k |\langle x \mid x_k \rangle|^2 + (1 - \alpha) \omega_k |\langle y \mid x_k \rangle|^2. \quad (11.7)$$

Thus,  $f(\alpha x + (1 - \alpha)y) < \alpha f(x) + (1 - \alpha)f(y)$ . Now set  $(\forall n \in \mathbb{N}) y_n = x_n / \sqrt{\omega_n}$ . Then  $\|y_n\| = 1/\sqrt{\omega_n} \rightarrow +\infty$ , but  $f(y_n) \equiv 1$ . Therefore,  $f$  is not coercive. Finally,  $f(x_n) = \omega_n \rightarrow 0 = f(0) = \inf f(\mathcal{H})$  and, by Example 2.25,  $x_n \rightharpoonup 0$  but  $x_n \not\rightarrow 0$ .  $\square$



**Fig. 11.1** Graph of the function in Example 11.24.

**Example 11.24** Suppose that  $\mathcal{H} = \mathbb{R}^2$ . This example provides a coercive function  $f \in \Gamma_0(\mathcal{H})$  with a unique minimizer  $\bar{x}$  for which alternating minimizations produce a convergent sequence that is not a minimizing sequence and the limit of which is not  $\bar{x}$ . Suppose that  $\mathcal{H} = \mathbb{R}^2$  and set (see [Figure 11.1](#))

$$\begin{aligned} f: \mathcal{H} &\rightarrow ]-\infty, +\infty] \\ (\xi_1, \xi_2) &\mapsto \max\{2\xi_1 - \xi_2, 2\xi_2 - \xi_1\} + \iota_{\mathbb{R}_+^2}(\xi_1, \xi_2) \\ &= \begin{cases} 2\xi_1 - \xi_2, & \text{if } \xi_1 \geq \xi_2 \geq 0; \\ 2\xi_2 - \xi_1, & \text{if } \xi_2 \geq \xi_1 \geq 0; \\ +\infty, & \text{otherwise.} \end{cases} \end{aligned} \quad (11.8)$$

It is clear that  $f$  is lower semicontinuous and convex as the sum of two such functions, and that it is coercive. Moreover,  $f$  admits  $\bar{x} = (0, 0)$  as its unique minimizer and  $\inf f(\mathcal{H}) = 0$ . Given an initial point  $x_0 \in \mathbb{R}_+ \times \mathbb{R}_{++}$ , we define iteratively an alternating minimization sequence  $(x_n)_{n \in \mathbb{N}}$  as follows: at iteration  $n$ ,  $x_n = (\xi_{1,n}, \xi_{2,n})$  is known and we construct  $x_{n+1} = (\xi_{1,n+1}, \xi_{2,n+1})$  by first letting  $\xi_{1,n+1}$  be the minimizer of  $f(\cdot, \xi_{2,n})$  over  $\mathbb{R}$  and then letting  $\xi_{2,n+1}$  be the minimizer of  $f(\xi_{1,n+1}, \cdot)$  over  $\mathbb{R}$ . In view of (11.8), for every integer  $n \geq 1$ , we obtain  $x_n = (\xi_{0,2}, \xi_{0,2}) \neq \bar{x}$  and  $f(x_n) = \xi_{0,2} \neq \inf f(\mathcal{H})$ .

We now provide weak and strong convergence conditions for minimizing sequences.

**Proposition 11.25** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, lower semicontinuous, and quasiconvex, and let  $(x_n)_{n \in \mathbb{N}}$  be a minimizing sequence of  $f$ . Suppose that there exists  $\xi \in ]\inf f(\mathcal{H}), +\infty[$  such that  $C = \text{lev}_{\leq \xi} f$  is bounded. Then the following hold:*

- (i) *The sequence  $(x_n)_{n \in \mathbb{N}}$  has a weak sequential cluster point and every such point is a minimizer of  $f$ .*
- (ii) *Suppose that  $f + \iota_C$  is strictly quasiconvex. Then  $f$  possesses a unique minimizer  $x$  and  $x_n \rightarrow x$ .*
- (iii) *Suppose that  $f + \iota_C$  is uniformly quasiconvex. Then  $f$  possesses a unique minimizer  $x$  and  $x_n \rightarrow x$ .*

*Proof.* Without loss of generality, we assume that  $(x_n)_{n \in \mathbb{N}}$  lies entirely in the bounded closed convex set  $C$ .

(i): The existence of a weak sequential cluster point is guaranteed by Lemma 2.37. The second assertion follows from Proposition 11.20.

(ii): Uniqueness follows from Proposition 11.7(i). In turn, the second assertion follows from (i) and Lemma 2.38.

(iii): Since  $f + \iota_C$  is strictly quasiconvex, we derive from (ii) that  $f$  possesses a unique minimizer  $x \in C$ . Now fix  $\alpha \in ]0, 1[$  and let  $\phi$  be the modulus of quasiconvexity of  $f + \iota_C$ . Then it follows from (10.24) that, for every  $n \in \mathbb{N}$ ,

$$\begin{aligned} f(x) + \alpha(1 - \alpha)\phi(\|x_n - x\|) &= \inf f(\mathcal{H}) + \alpha(1 - \alpha)\phi(\|x_n - x\|) \\ &\leq f(\alpha x_n + (1 - \alpha)x) + \alpha(1 - \alpha)\phi(\|x_n - x\|) \\ &\leq \max\{f(x_n), f(x)\} \\ &= f(x_n). \end{aligned} \tag{11.9}$$

Consequently, since  $f(x_n) \rightarrow f(x)$ , we obtain  $\phi(\|x_n - x\|) \rightarrow 0$  and, in turn,  $\|x_n - x\| \rightarrow 0$ .  $\square$

**Corollary 11.26** *Let  $f \in \Gamma_0(\mathcal{H})$  be coercive and let  $(x_n)_{n \in \mathbb{N}}$  be a minimizing sequence of  $f$ . Then the following hold:*

- (i) *The sequence  $(x_n)_{n \in \mathbb{N}}$  has a weak sequential cluster point and every such point is a minimizer of  $f$ .*

- (ii) Suppose that  $f$  is strictly convex. Then  $f$  possesses a unique minimizer  $x$  and  $x_n \rightarrow x$ .
- (iii) Suppose that  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } f$ . Then  $f$  possesses a unique minimizer  $x$  and  $x_n \rightarrow x$ .

Another instance of strong convergence of minimizing sequences in convex variational problems is the following.

**Proposition 11.27** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $C$  be a bounded closed convex subset of  $\mathcal{H}$  such that  $C \cap \text{dom } f \neq \emptyset$ . Suppose that  $C \cap \text{Argmin } f = \emptyset$  and that  $C$  is uniformly convex, i.e., there exists an increasing function  $\phi: [0, \text{diam } C] \rightarrow \mathbb{R}_+$  that vanishes only at 0 such that*

$$(\forall x \in C)(\forall y \in C) \quad B\left(\frac{x+y}{2}; \phi(\|x-y\|)\right) \subset C, \quad (11.10)$$

*and let  $(x_n)_{n \in \mathbb{N}}$  be a minimizing sequence of  $f + \iota_C$ . Then  $f$  admits a unique minimizer  $x$  over  $C$  and  $x_n \rightarrow x$ .*

*Proof.* We assume that  $C$  is not a singleton. The existence of  $x$  follows from Proposition 11.14(ii) and its uniqueness from Proposition 11.7(ii). In turn, we deduce from Proposition 11.25(i) and Lemma 2.38 that  $x_n \rightarrow x$ . Moreover, since  $C \cap \text{Argmin } f = \emptyset$ , it follows from Proposition 11.5 that  $x \in \text{bdry } C$ . Hence, since (11.10) asserts that  $\text{int } C \neq \emptyset$ , we derive from Corollary 7.6(i) that  $x$  is a support point of  $C$ . Denote by  $u$  an associated normal vector such that  $\|u\| = 1$ . Now let  $n \in \mathbb{N}$  and set  $z_n = (x_n + x)/2 + \phi(\|x_n - x\|)u$ . Then it follows from (11.10) that  $z_n \in C$  and from (7.1) that  $\langle x_n - x \mid u \rangle / 2 + \phi(\|x_n - x\|) = \langle z_n - x \mid u \rangle \leq 0$ . Hence,  $\phi(\|x_n - x\|) \leq \langle x - x_n \mid u \rangle / 2 \rightarrow 0$  and therefore  $x_n \rightarrow x$ .  $\square$

## Exercises

**Exercise 11.1** Provide a function  $f \in \Gamma_0(\mathcal{H})$  and a nonempty set  $C \subset \text{int dom } f$  such that  $\inf f(\overline{C}) < \inf f(C)$ . Compare with Proposition 11.1(iv).

**Exercise 11.2** Provide a function  $f \in \Gamma_0(\mathcal{H})$  and a nonempty convex set  $C \subset \text{dom } f$  such that  $\inf f(\overline{C}) < \inf f(C)$ . Compare with Proposition 11.1(iv).

**Exercise 11.3** Provide a convex function  $f: \mathcal{H} \rightarrow \mathbb{R}$  and a nonempty convex set  $C \subset \mathcal{H}$  such that  $\inf f(\overline{C}) < \inf f(C)$ . Compare with Proposition 11.1(iv).

**Exercise 11.4** Show that Proposition 11.4 is false if  $f$  is merely quasiconvex.

**Exercise 11.5** Find a function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  that is proper and convex, a set  $C \subset \mathcal{H}$ , and a minimizer  $x$  of  $f$  over  $C$  such that  $x \in \text{bdry } C$  and  $\text{Argmin}(f + \iota_C) \cap \text{Argmin } f = \emptyset$ . Compare with Proposition 11.5.

**Exercise 11.6** Find a convex function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and a convex set  $C \subset \mathcal{H}$  such that  $C \cap \text{dom } f \neq \emptyset$ ,  $C \cap \text{Argmin } f = \emptyset$ , and  $f$  has at least two minimizers over  $C$ . Compare with Proposition 11.7(ii).

**Exercise 11.7** Let  $C$  be a convex subset of  $\mathcal{H}$ . Show that  $C$  is strictly convex if and only if  $(\forall x \in C)(\forall y \in C) \ x \neq y \Rightarrow ]x, y[ \subset \text{int } C$ .

**Exercise 11.8** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper lower semicontinuous coercive quasiconvex function. Without using Theorem 11.9, derive from Proposition 3.18 and Proposition 11.11 that  $f$  has a minimizer.

**Exercise 11.9** Show that the conclusion of Proposition 11.12 is false if  $f$  is not lower semicontinuous, even when  $\mathcal{H} = \mathbb{R}^2$ .

**Exercise 11.10** Show that the conclusion of Proposition 11.12 is false if  $\mathcal{H}$  is infinite-dimensional.

**Exercise 11.11** Show that Proposition 11.13 is false if “supercoercive” is replaced by “coercive,” even if  $g \in \Gamma_0(\mathcal{H})$ .

**Exercise 11.12** Provide an alternative proof for Theorem 4.19 using Corollary 11.18.

**Exercise 11.13** In Example 11.23, suppose that  $(x_n)_{n \in \mathbb{N}}$  is an orthonormal sequence that is not an orthonormal basis. What is  $\text{Argmin } f$ ? Is  $f$  strictly convex?

**Exercise 11.14** Let  $f \in \Gamma_0(\mathcal{H})$  be bounded below, let  $\beta \in \mathbb{R}_{++}$ , let  $y \in \text{dom } f$ , let  $p \in ]1, +\infty[$ , and set  $\alpha = f(y) - \inf f(\mathcal{H})$ . Prove that there exists  $z \in \mathcal{H}$  such that  $\|z - y\| \leq \beta$  and

$$(\forall x \in \mathcal{H}) \quad f(z) + \frac{\alpha}{\beta^p} \|z - y\|^p \leq f(x) + \frac{\alpha}{\beta^p} \|x - y\|^p. \quad (11.11)$$

**Exercise 11.15** Find a function  $f \in \Gamma_0(\mathcal{H})$  such that  $\text{Argmin } f$  is a singleton and such that  $f$  admits an unbounded minimizing sequence.

**Exercise 11.16** Show that  $B(0; \rho)$  is uniformly convex and that (11.10) holds with  $\phi: [0, 2\rho] \rightarrow \mathbb{R}_+: t \mapsto \rho - \sqrt{\rho^2 - (t/2)^2}$ .





# Chapter 12

## Infimal Convolution

This chapter is devoted to a fundamental convexity-preserving operation for functions: the infimal convolution. Its properties are investigated, with special emphasis on the Moreau envelope, which is obtained by convolving a function with the halved squared norm.

### 12.1 Definition and Basic Facts

**Definition 12.1** Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . The *infimal convolution* (or *epi-sum*) of  $f$  and  $g$  is

$$f \square g: \mathcal{H} \rightarrow [-\infty, +\infty]: x \mapsto \inf_{y \in \mathcal{H}} (f(y) + g(x - y)), \quad (12.1)$$

and it is *exact at a point*  $x \in \mathcal{H}$  if  $(f \square g)(x) = \min_{y \in \mathcal{H}} f(y) + g(x - y)$ , i.e. (see Definition 1.7),

$$(\exists y \in \mathcal{H}) \quad (f \square g)(x) = f(y) + g(x - y) \in ]-\infty, +\infty]; \quad (12.2)$$

$f \square g$  is *exact* if it is exact at every point of its domain, in which case it is denoted by  $f \square g$ .

**Example 12.2** Let  $C$  be a subset of  $\mathcal{H}$ . Then it follows from (1.39) and (1.45) that  $d_C = \iota_C \square \|\cdot\|$ . Moreover, Remark 3.9(i) asserts that, if  $C$  is nonempty and open, this infimal convolution is never exact on  $\mathcal{H} \setminus C$ , though always real-valued.

**Example 12.3** Let  $C$  and  $D$  be nonempty subsets of  $\mathcal{H}$ . Then  $\iota_C \square \iota_D = \iota_{C+D}$ .

**Example 12.4** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $y \in \mathcal{H}$ . Then  $\iota_{\{y\}} \square f = \tau_y f$ .

**Definition 12.5** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $u \in \mathcal{H}$ . Then  $f$  possesses a *continuous affine minorant with slope  $u$*  if  $f - \langle \cdot | u \rangle$  is bounded below.

**Proposition 12.6** Let  $f, g$ , and  $h$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Then the following hold:

- (i) Suppose that  $f$  and  $g$  possess continuous affine minorants with slope  $u \in \mathcal{H}$ . Then  $f \square g$  possesses a continuous affine minorant with slope  $u$  and  $-\infty \notin (f \square g)(\mathcal{H})$ .
- (ii)  $\text{dom}(f \square g) = \text{dom } f + \text{dom } g$ .
- (iii)  $f \square g = g \square f$ .
- (iv) Suppose that  $f, g$ , and  $h$  possess continuous affine minorants with the same slope. Then  $f \square (g \square h) = (f \square g) \square h$ .

*Proof.* (i): By assumption, there exist  $\eta \in \mathbb{R}$  and  $\mu \in \mathbb{R}$  such that  $f \geq \langle \cdot | u \rangle + \eta$  and  $g \geq \langle \cdot | u \rangle + \mu$ . Now fix  $x \in \mathcal{H}$ . Then, for every  $y \in \mathcal{H}$ ,  $f(y) + g(x - y) \geq \langle x | u \rangle + \eta + \mu$  and, therefore,  $f \square g \geq \langle \cdot | u \rangle + \eta + \mu > -\infty$ .

(ii)&(iii): Observe that (12.1) can be rewritten as

$$f \square g: \mathcal{H} \rightarrow [-\infty, +\infty]: x \mapsto \inf_{\substack{(u,v) \in \mathcal{H} \times \mathcal{H} \\ u+v=x}} f(u) + g(v). \quad (12.3)$$

(iv): It follows from (i) that  $g \square h$  and  $f \square g$  are functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Therefore, the infimal convolutions  $f \square (g \square h)$  and  $(f \square g) \square h$  are well defined. Furthermore, using (12.3), we obtain

$$\begin{aligned} (\forall x \in \mathcal{H}) \quad (f \square (g \square h))(x) &= \inf_{\substack{(u,v,w) \in \mathcal{H} \times \mathcal{H} \times \mathcal{H} \\ u+v+w=x}} (f(u) + g(v) + h(w)) \\ &= ((f \square g) \square h)(x), \end{aligned} \quad (12.4)$$

as desired.  $\square$

**Example 12.7** Set  $f: \mathbb{R} \rightarrow \mathbb{R}: \xi \mapsto \xi$  and  $g = -f$ . Then  $f \square g \equiv -\infty$ . This shows that Proposition 12.6(i) fails if the assumption on the minorants is not satisfied.

**Proposition 12.8** Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Then the following hold:

- (i)  $\text{epi } f + \text{epi } g \subset \text{epi}(f \square g)$ .
- (ii) Suppose that  $f \square g = f \square g$ . Then  $\text{epi}(f \square g) = \text{epi } f + \text{epi } g$ .

*Proof.* (i): Take  $(x, \xi) \in \text{epi } f$  and  $(y, \eta) \in \text{epi } g$ . Then (12.3) yields

$$(f \square g)(x + y) \leq f(x) + g(y) \leq \xi + \eta. \quad (12.5)$$

Therefore  $(x + y, \xi + \eta) \in \text{epi}(f \square g)$ .

(ii): Take  $(x, \xi) \in \text{epi}(f \square g)$ . In view of (i), it suffices to show that  $(x, \xi) \in \text{epi } f + \text{epi } g$ . By assumption, there exists  $y \in \mathcal{H}$  such that  $(f \square g)(x) = f(y) +$

$g(x - y) \leq \xi$ . Therefore  $(x - y, \xi - f(y)) \in \text{epi } g$  and, in turn,  $(x, \xi) = (y, f(y)) + (x - y, \xi - f(y)) \in \text{epi } f + \text{epi } g$ .  $\square$

A central example of infimal convolution is obtained by using a power of the norm.

**Proposition 12.9** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $p \in [1, +\infty[$ , and set*

$$(\forall \gamma \in \mathbb{R}_{++}) \quad g_\gamma = f \square \left( \frac{1}{\gamma p} \|\cdot\|^p \right). \quad (12.6)$$

*Then the following hold for every  $\gamma \in \mathbb{R}_{++}$  and every  $x \in \mathcal{H}$ :*

- (i)  $\text{dom } g_\gamma = \mathcal{H}$ .
- (ii) Take  $\mu \in ]\gamma, +\infty[$ . Then  $\inf f(\mathcal{H}) \leq g_\mu(x) \leq g_\gamma(x) \leq f(x)$ .
- (iii)  $\inf g_\gamma(\mathcal{H}) = \inf f(\mathcal{H})$ .
- (iv)  $g_\mu(x) \downarrow \inf f(\mathcal{H})$  as  $\mu \uparrow +\infty$ .
- (v)  $g_\gamma$  is bounded above on every ball in  $\mathcal{H}$ .

*Proof.* Let  $x \in \mathcal{H}$  and  $\gamma \in \mathbb{R}_{++}$ .

(i): By Proposition 12.6(ii),  $\text{dom } g_\gamma = \text{dom } f + \text{dom}(\|\cdot\|^p/(\gamma p)) = \text{dom } f + \mathcal{H} = \mathcal{H}$ .

(ii): It is clear that  $g_\mu(x) \leq g_\gamma(x)$ . On the other hand,

$$\inf f(\mathcal{H}) \leq g_\gamma(x) = \inf_{y \in \mathcal{H}} \left( f(y) + \frac{1}{\gamma p} \|x - y\|^p \right) \leq f(x). \quad (12.7)$$

(ii)  $\Rightarrow$  (iii): Clear.

(iv): Let  $y \in \mathcal{H}$ . Then, for every  $\mu \in \mathbb{R}_{++}$ ,  $g_\mu(x) \leq f(y) + (\mu p)^{-1} \|x - y\|^p$  and therefore  $\overline{\lim}_{\mu \uparrow +\infty} g_\mu(x) \leq f(y)$ . Appealing to (ii) and taking the infimum over  $y \in \mathcal{H}$ , we obtain

$$\inf f(\mathcal{H}) \leq \underline{\lim}_{\mu \uparrow +\infty} g_\mu(x) \leq \overline{\lim}_{\mu \uparrow +\infty} g_\mu(x) \leq \inf f(\mathcal{H}). \quad (12.8)$$

(v): Fix  $z \in \text{dom } f$  and  $\rho \in \mathbb{R}_{++}$ . Then

$$\begin{aligned} (\forall y \in B(x; \rho)) \quad g_\gamma(y) &\leq f(z) + \|y - z\|^p / (\gamma p) \\ &\leq f(z) + 2^{p-1} (\|y - x\|^p + \|x - z\|^p) / (\gamma p) \end{aligned} \quad (12.9)$$

$$\begin{aligned} &\leq f(z) + 2^{p-1} (\rho^p + \|x - z\|^p) / (\gamma p) \\ &< +\infty, \end{aligned} \quad (12.10)$$

where (12.9) follows either from the triangle inequality or from (8.15), depending on whether  $p = 1$  or  $p > 1$ .  $\square$

**Remark 12.10** In (12.6), we may have  $g_\gamma \equiv -\infty$ , even if  $f$  is real-valued. For instance, suppose that  $\mathcal{H}$  is infinite-dimensional and let  $f$  be a discontinuous linear functional (see Example 2.20 for a construction and Example 8.33 for

properties). Since, for every  $\rho \in \mathbb{R}_{++}$ ,  $f$  is unbounded below on  $B(0; \rho)$ , there exists a sequence  $(y_n)_{n \in \mathbb{N}}$  in  $\mathcal{H}$  such that  $y_n \rightarrow 0$  and  $f(y_n) \rightarrow -\infty$ . Thus, for every  $x \in \mathcal{H}$ ,  $g_\gamma(x) \leq f(y_n) + (\gamma p)^{-1} \|x - y_n\|^p \rightarrow -\infty$ . We conclude that  $g_\gamma \equiv -\infty$ .

## 12.2 Infimal Convolution of Convex Functions

**Proposition 12.11** *Let  $f$  and  $g$  be convex functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$ . Then  $f \square g$  is convex.*

*Proof.* Take  $F: \mathcal{H} \times \mathcal{H} \rightarrow ] -\infty, +\infty]: (x, y) \mapsto f(y) + g(x - y)$  in Proposition 8.26.  $\square$

**Corollary 12.12** *Let  $C$  be a convex subset of  $\mathcal{H}$ . Then  $d_C$  is convex.*

*Proof.* As seen in Example 12.2,  $d_C = \iota_C \square \|\cdot\|$ . Since  $\iota_C$  and  $\|\cdot\|$  are convex (see Example 8.3 and Example 8.7), so is  $d_C$  by Proposition 12.11.  $\square$

The following examples show that the infimal convolution of two functions in  $\Gamma_0(\mathcal{H})$  need not be exact or lower semicontinuous.

**Example 12.13** Set  $f: \mathbb{R} \rightarrow ] -\infty, +\infty]: x \mapsto 1/x$  if  $x > 0$ ;  $+\infty$  otherwise, and set  $g = f^\vee$ . Then the following hold:

- (i)  $f \in \Gamma_0(\mathbb{R})$  and  $g \in \Gamma_0(\mathbb{R})$ .
- (ii)  $f \square g \equiv 0$  and  $f \square g$  is nowhere exact.
- (iii) Set  $\varphi = \iota_C$  and  $\psi = \iota_D$ , where  $C = \text{epi } f$  and  $D = \text{epi } g$ . It follows from (i) that  $C$  and  $D$  are nonempty closed convex subsets of  $\mathbb{R}^2$ . Therefore  $\varphi$  and  $\psi$  are in  $\Gamma_0(\mathbb{R}^2)$ . However,  $C + D$  is the open upper half-plane in  $\mathbb{R}^2$  and thus  $\varphi \square \psi = \iota_{C+D}$  is not lower semicontinuous.

We now present conditions under which the infimal convolution of two functions in  $\Gamma_0(\mathcal{H})$  is exact and in  $\Gamma_0(\mathcal{H})$  (see also Proposition 15.7).

**Proposition 12.14** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and suppose that one of the following holds:*

- (i)  $f$  is supercoercive.
- (ii)  $f$  is coercive and  $g$  is bounded below.

*Then  $f \square g = f \boxplus g \in \Gamma_0(\mathcal{H})$ .*

*Proof.* By Proposition 12.6(ii),  $\text{dom } f \square g = \text{dom } f + \text{dom } g \neq \emptyset$ . Now let  $x \in \text{dom } f \square g$ . Then  $\text{dom } f \cap \text{dom } g(x - \cdot) \neq \emptyset$  and hence Corollary 11.15 implies that  $f + g(x - \cdot)$  has a minimizer over  $\mathcal{H}$ . Thus,  $(\forall x \in \text{dom}(f \square g))$   $(f \square g)(x) = \min_{y \in \mathcal{H}} f(y) + g(x - y) \in \mathbb{R}$ . Therefore,  $f \square g = f \boxplus g$  and  $f \square g$  is proper. In view of Proposition 12.11 and Theorem 9.1, to complete the

proof it suffices to show that  $f \square g$  is sequentially lower semicontinuous. To this end, let  $x \in \mathcal{H}$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  such that  $x_n \rightarrow x$ . We need to show that

$$(f \square g)(x) \leq \liminf (f \square g)(x_n). \quad (12.11)$$

After passing to a subsequence and relabeling, we assume that the sequence  $((f \square g)(x_n))_{n \in \mathbb{N}}$  converges, say  $(f \square g)(x_n) \rightarrow \mu \in [-\infty, +\infty[$ . Let  $(y_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) (f \square g)(x_n) = f(y_n) + g(x_n - y_n)$ . We claim that

$$(y_n)_{n \in \mathbb{N}} \text{ is bounded.} \quad (12.12)$$

Assume that (12.12) is false. After passing to a subsequence and relabeling, we obtain  $0 \neq \|y_n\| \rightarrow +\infty$ . We now show that a contradiction ensues from each hypothesis.

(i): By Theorem 9.19,  $g$  possesses a continuous affine minorant, say  $\langle \cdot | u \rangle + \eta$ , where  $u \in \mathcal{H}$  and  $\eta \in \mathbb{R}$ . Using the supercoercivity of  $f$ , we get

$$\begin{aligned} +\infty &> \mu \\ &\leftarrow (f \square g)(x_n) \\ &= f(y_n) + g(x_n - y_n) \\ &\geq f(y_n) + \langle x_n - y_n | u \rangle + \eta \\ &\geq \|y_n\| \left( \frac{f(y_n)}{\|y_n\|} - \|u\| \right) + \langle x_n | u \rangle + \eta \\ &\rightarrow +\infty, \end{aligned} \quad (12.13)$$

which is impossible.

(ii): Since  $f$  is coercive, we have  $f(y_n) \rightarrow +\infty$ . Hence,  $g(x_n - y_n) \rightarrow -\infty$  since  $f(y_n) + g(x_n - y_n) \rightarrow \mu < +\infty$ . However, this is impossible since  $g$  is bounded below.

Hence, (12.12) holds in both cases. After passing to a subsequence and relabeling, we assume that  $(y_n)_{n \in \mathbb{N}}$  converges weakly to some point  $y \in \mathcal{H}$ . Then  $x_n - y_n \rightarrow x - y$  and thus  $\mu = \lim (f \square g)(x_n) = \lim f(y_n) + g(x_n - y_n) \geq \liminf f(y_n) + \liminf g(x_n - y_n) \geq f(y) + g(x - y) \geq (f \square g)(x)$ . Therefore, (12.11) is verified and the proof is complete.  $\square$

The next proposition examines the properties of the infimal convolution of a convex function with a power of the norm.

**Proposition 12.15** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $p \in ]1, +\infty[$ . Then the infimal convolution*

$$f \square \left( \frac{1}{\gamma^p} \|\cdot\|^p \right) : \mathcal{H} \rightarrow ]-\infty, +\infty] : x \mapsto \inf_{y \in \mathcal{H}} \left( f(y) + \frac{1}{\gamma^p} \|x - y\|^p \right) \quad (12.14)$$

is convex, real-valued, continuous, and exact. Moreover, for every  $x \in \mathcal{H}$ , the infimum in (12.14) is uniquely attained.

*Proof.* Let us define  $g_\gamma$  as in (12.6). We observe that  $(\gamma p)^{-1} \|\cdot\|^p$  is supercoercive and, by Example 8.21, strictly convex. Proposition 12.14(i) implies that  $g_\gamma$  is proper, lower semicontinuous, convex, and exact. Combining this, Proposition 12.9(v), and Corollary 8.30(i), we obtain that  $g_\gamma$  is real-valued and continuous. The statement concerning the unique minimizer follows from Corollary 11.15(i).  $\square$

In the next two sections, we examine the cases  $p = 1$  and  $p = 2$  in (12.14) in more detail.

### 12.3 Pasch–Hausdorff Envelope

**Definition 12.16** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $\beta \in \mathbb{R}_+$ . The  $\beta$ -Pasch–Hausdorff envelope of  $f$  is  $f \square (\beta \|\cdot\|)$ .

Observe that the Pasch–Hausdorff envelope enjoys all the properties listed in Proposition 12.9.

**Proposition 12.17** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $\beta \in \mathbb{R}_+$ , and let  $g$  be the  $\beta$ -Pasch–Hausdorff envelope of  $f$ . Then exactly one of the following holds:

- (i)  $f$  possesses a  $\beta$ -Lipschitz continuous minorant, and  $g$  is the largest  $\beta$ -Lipschitz continuous minorant of  $f$ .
- (ii)  $f$  possesses no  $\beta$ -Lipschitz continuous minorant, and  $g \equiv -\infty$ .

*Proof.* We first note that Proposition 12.9(i) implies that

$$\begin{aligned} (\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad g(x) &= \inf_{z \in \mathcal{H}} (f(z) + \beta \|x - z\|) \\ &\leq \beta \|x - y\| + \inf_{z \in \mathcal{H}} (f(z) + \beta \|y - z\|) \\ &= \beta \|x - y\| + g(y) \\ &< +\infty. \end{aligned} \tag{12.15}$$

(i): Suppose that  $f$  possesses a  $\beta$ -Lipschitz continuous minorant  $h: \mathcal{H} \rightarrow \mathbb{R}$ . Then

$$\begin{aligned} (\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad h(x) &\leq h(y) + \beta \|x - y\| \\ &\leq f(y) + \beta \|x - y\|. \end{aligned} \tag{12.16}$$

Taking the infimum over  $y \in \mathcal{H}$ , we obtain

$$(\forall x \in \mathcal{H}) \quad h(x) \leq g(x). \tag{12.17}$$

Consequently,  $-\infty \notin g(\mathcal{H})$  and (12.15) implies that  $g$  is real-valued and  $\beta$ -Lipschitz continuous. On the other hand, it follows from Proposition 12.9(ii) that  $g \leq f$ .

(ii): Suppose that  $f$  possesses no  $\beta$ -Lipschitz continuous minorant and that  $g$  is real-valued at some point in  $\mathcal{H}$ . We derive from (12.15) that  $g$  is everywhere real-valued and  $\beta$ -Lipschitz continuous. If  $\beta = 0$ , then  $g \equiv \inf f(\mathcal{H})$ , which contradicts the assumption that  $f$  has no 0-Lipschitz continuous minorant. On the other hand, if  $\beta > 0$ , then Proposition 12.9(ii) implies that  $g$  is a minorant of  $f$ , and we once again reach a contradiction.  $\square$

We deduce at once the following involution property from Proposition 12.17(i).

**Corollary 12.18** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be  $\beta$ -Lipschitz continuous for some  $\beta \in \mathbb{R}_{++}$ . Then  $f$  is its own  $\beta$ -Pasch-Hausdorff envelope.*

**Corollary 12.19** *Let  $C$  be a nonempty subset of  $\mathcal{H}$  and let  $h: C \rightarrow \mathbb{R}$  be  $\beta$ -Lipschitz continuous for some  $\beta \in \mathbb{R}_+$ . Set*

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} h(x), & \text{if } x \in C; \\ +\infty, & \text{otherwise.} \end{cases} \quad (12.18)$$

*Then  $f \square (\beta \|\cdot\|)$  is a  $\beta$ -Lipschitz continuous extension of  $h$ .*

*Proof.* Set  $g = f \square (\beta \|\cdot\|)$ . Then, for every  $x \in C$ ,

$$g(x) = \inf_{y \in \mathcal{H}} (f(y) + \beta \|x - y\|) = \inf_{y \in C} (h(y) + \beta \|x - y\|) \geq h(x) > -\infty. \quad (12.19)$$

Hence,  $g \not\equiv -\infty$  and Proposition 12.17 implies that  $g$  is the largest  $\beta$ -Lipschitz continuous minorant of  $f$ . In particular,  $g|_C \leq f|_C = h$ . On the other hand, (12.19) yields  $g|_C \geq h$ . Altogether,  $g|_C = h$  and the proof is complete.  $\square$

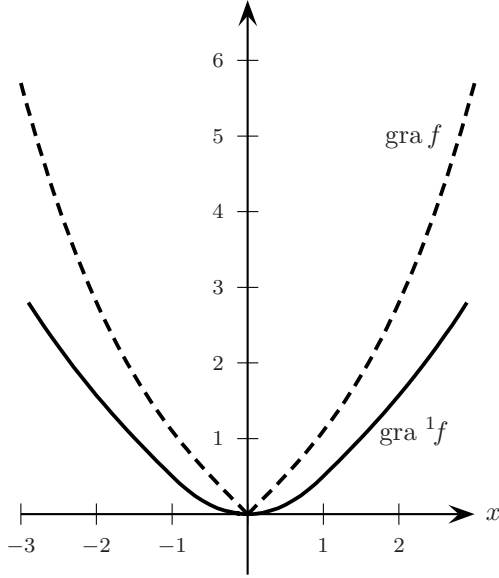
## 12.4 Moreau Envelope

The most important instance of (12.6) is obtained when  $p = 2$ .

**Definition 12.20** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $\gamma \in \mathbb{R}_{++}$ . The *Moreau envelope* of  $f$  of parameter  $\gamma$  is

$$\gamma f = f \square \left( \frac{1}{2\gamma} \|\cdot\|^2 \right). \quad (12.20)$$

**Example 12.21** Let  $C \subset \mathcal{H}$  and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\gamma \iota_C = (2\gamma)^{-1} d_C^2$ .



**Fig. 12.1** Graphs of  $f: x \mapsto |x| + 0.1|x|^3$  and of its Moreau envelope  ${}^1f$ .

As a special case of (12.6) with  $p > 1$ , the Moreau envelope inherits all the properties recorded in Proposition 12.9 and Proposition 12.15. In addition, it possesses specific properties that we now examine.

**Proposition 12.22** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ ,  $\gamma \in \mathbb{R}_{++}$ , and  $\mu \in \mathbb{R}_{++}$ . Then the following hold:*

- (i)  ${}^\mu(\gamma f) = \gamma({}^\gamma{}^\mu f)$ .
- (ii)  $\gamma({}^\mu f) = {}^{(\gamma+\mu)}f$ .

*Proof.* Fix  $x \in \mathcal{H}$ .

(i): We derive from (12.20) that

$${}^\mu(\gamma f)(x) = \gamma \left( \inf_{y \in \mathcal{H}} f(y) + \frac{1}{2\gamma\mu} \|x - y\|^2 \right) = \gamma({}^\gamma{}^\mu f)(x). \quad (12.21)$$

(ii): Set  $\alpha = \mu/(\mu + \gamma)$ . Then it follows from Corollary 2.14 that

$$\begin{aligned} \gamma({}^\mu f)(x) &= \inf_{z \in \mathcal{H}} \left( \inf_{y \in \mathcal{H}} \left( f(y) + \frac{1}{2\mu} \|z - y\|^2 \right) + \frac{1}{2\gamma} \|z - x\|^2 \right) \\ &= \inf_{y \in \mathcal{H}} \left( f(y) + \frac{1}{2\alpha\gamma} \inf_{z \in \mathcal{H}} (\alpha \|z - x\|^2 + (1 - \alpha) \|z - y\|^2) \right) \end{aligned}$$



$$\begin{aligned}
&= \inf_{y \in \mathcal{H}} \left( f(y) \right. \\
&\quad \left. + \frac{1}{2\alpha\gamma} \inf_{z \in \mathcal{H}} (\|z - (\alpha x + (1 - \alpha)y\|^2 + \alpha(1 - \alpha)\|x - y\|^2) \right) \\
&= \inf_{y \in \mathcal{H}} \left( f(y) + \frac{1}{2(\gamma + \mu)} \|x - y\|^2 \right), \tag{12.22}
\end{aligned}$$

which yields  $\gamma(\mu f)(x) = {}^{(\gamma+\mu)}f(x)$ .  $\square$

In the case  $p = 2$ , Proposition 12.15 motivates the following definition.

**Definition 12.23** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $x \in \mathcal{H}$ . Then  $\text{Prox}_f x$  is the unique point in  $\mathcal{H}$  that satisfies

$${}^1f(x) = \min_{y \in \mathcal{H}} \left( f(y) + \frac{1}{2} \|x - y\|^2 \right) = f(\text{Prox}_f x) + \frac{1}{2} \|x - \text{Prox}_f x\|^2. \tag{12.23}$$

The operator  $\text{Prox}_f: \mathcal{H} \rightarrow \mathcal{H}$  is the *proximity operator*—or *proximal mapping*—of  $f$ .

**Remark 12.24** Let  $f \in \Gamma_0(\mathcal{H})$ ,  $\gamma \in \mathbb{R}_{++}$ , and  $x \in \mathcal{H}$ . Proposition 12.22(i) with  $\mu = 1$  yields  ${}^1(\gamma f) = \gamma({}^\gamma f)$ . Hence we derive from (12.23) that

$${}^\gamma f(x) = f(\text{Prox}_{\gamma f} x) + \frac{1}{2\gamma} \|x - \text{Prox}_{\gamma f} x\|^2. \tag{12.24}$$

**Example 12.25** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $\text{Prox}_{\iota_C} = P_C$ .

The next two results generalize (3.6) and Proposition 4.8, respectively.

**Proposition 12.26** Let  $f \in \Gamma_0(\mathcal{H})$ , and let  $x$  and  $p$  be in  $\mathcal{H}$ . Then

$$p = \text{Prox}_f x \iff (\forall y \in \mathcal{H}) \quad \langle y - p \mid x - p \rangle + f(p) \leq f(y). \tag{12.25}$$

*Proof.* Let  $y \in \mathcal{H}$ . First, suppose that  $p = \text{Prox}_f x$  and set  $(\forall \alpha \in ]0, 1[)$   $p_\alpha = \alpha y + (1 - \alpha)p$ . Then, for every  $\alpha \in ]0, 1[$ , (12.23) and the convexity of  $f$  yield

$$\begin{aligned}
f(p) &\leq f(p_\alpha) + \frac{1}{2} \|x - p_\alpha\|^2 - \frac{1}{2} \|x - p\|^2 \\
&\leq \alpha f(y) + (1 - \alpha)f(p) - \alpha \langle x - p \mid y - p \rangle + \frac{\alpha^2}{2} \|y - p\|^2 \tag{12.26}
\end{aligned}$$

and hence  $\langle y - p \mid x - p \rangle + f(p) \leq f(y) + (\alpha/2)\|y - p\|^2$ . Letting  $\alpha \downarrow 0$ , we obtain the desired inequality. Conversely, suppose that  $\langle y - p \mid x - p \rangle + f(p) \leq f(y)$ . Then certainly  $f(p) + (1/2)\|x - p\|^2 \leq f(y) + (1/2)\|x - p\|^2 + \langle x - p \mid p - y \rangle + (1/2)\|p - y\|^2 = f(y) + (1/2)\|x - y\|^2$  and we conclude that  $p = \text{Prox}_f x$ .  $\square$

**Proposition 12.27** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{Prox}_f$  and  $\text{Id} - \text{Prox}_f$  are firmly nonexpansive.*

*Proof.* Take  $x$  and  $y$  in  $\mathcal{H}$ , and set  $p = \text{Prox}_f x$  and  $q = \text{Prox}_f y$ . Then Proposition 12.26 yields  $\langle q - p \mid x - p \rangle + f(p) \leq f(q)$  and  $\langle p - q \mid y - q \rangle + f(q) \leq f(p)$ . Since  $p$  and  $q$  lie in  $\text{dom } f$ , upon adding these two inequalities, we get  $0 \leq \langle p - q \mid (x - p) - (y - q) \rangle$  and conclude via Proposition 4.2.  $\square$

**Proposition 12.28** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then*

$$\text{Fix } \text{Prox}_f = \text{Argmin } f. \quad (12.27)$$

*Proof.* Let  $x \in \mathcal{H}$ . Then it follows from Proposition 12.26 that  $x = \text{Prox}_f x \Leftrightarrow (\forall y \in \mathcal{H}) \langle y - x \mid x - x \rangle + f(x) \leq f(y) \Leftrightarrow (\forall y \in \mathcal{H}) f(x) \leq f(y) \Leftrightarrow x \in \text{Argmin } f$ .  $\square$

It follows from Proposition 12.15 that the Moreau envelope of  $f \in \Gamma_0(\mathcal{H})$  is convex, real-valued, and continuous. The next result states that it is actually Fréchet differentiable on  $\mathcal{H}$ .

**Proposition 12.29** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\gamma f: \mathcal{H} \rightarrow \mathbb{R}$  is Fréchet differentiable on  $\mathcal{H}$ , and its gradient*

$$\nabla(\gamma f) = \gamma^{-1}(\text{Id} - \text{Prox}_{\gamma f}) \quad (12.28)$$

*is  $\gamma^{-1}$ -Lipschitz continuous.*

*Proof.* Assume that  $x$  and  $y$  are distinct points in  $\mathcal{H}$ , and set  $p = \text{Prox}_{\gamma f} x$  and  $q = \text{Prox}_{\gamma f} y$ . Using (12.24) and Proposition 12.26, we obtain

$$\begin{aligned} \gamma f(y) - \gamma f(x) &= f(q) - f(p) + (\|y - q\|^2 - \|x - p\|^2)/(2\gamma) \\ &\geq (2\langle q - p \mid x - p \rangle + \|y - q\|^2 - \|x - p\|^2)/(2\gamma) \\ &= (\|y - q - x + p\|^2 + 2\langle y - x \mid x - p \rangle)/(2\gamma) \\ &\geq \langle y - x \mid x - p \rangle / \gamma. \end{aligned} \quad (12.29)$$

Likewise,  $\gamma f(y) - \gamma f(x) \leq \langle y - x \mid y - q \rangle / \gamma$ . Combining the last two inequalities and using the firm nonexpansiveness of  $\text{Prox}_f$  (Proposition 12.27), we get

$$\begin{aligned} 0 &\leq \gamma f(y) - \gamma f(x) - \langle y - x \mid x - p \rangle / \gamma \\ &\leq \langle y - x \mid (y - q) - (x - p) \rangle / \gamma \\ &\leq (\|y - x\|^2 - \|q - p\|^2) / \gamma \\ &\leq \|y - x\|^2 / \gamma. \end{aligned} \quad (12.30)$$

Thus,  $\lim_{y \rightarrow x} (\gamma f(y) - \gamma f(x) - \langle y - x \mid \gamma^{-1}(x - p) \rangle) / \|y - x\| = 0$ . Finally, Lipschitz continuity follows from Proposition 12.27.  $\square$

**Corollary 12.30** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $d_C^2$  is Fréchet differentiable on  $\mathcal{H}$  and*

$$\nabla d_C^2 = 2(\text{Id} - P_C). \quad (12.31)$$

*Proof.* Apply Proposition 12.29 with  $f = \iota_C$  and  $\gamma = 1/2$ , and use Example 12.21 and Example 12.25.  $\square$

**Proposition 12.31** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ , and set  $q = (1/2)\|\cdot\|^2$ . Then  $\nabla(q \circ P_K) = \nabla((1/2)d_{K^\ominus}^2) = P_K$ .*

*Proof.* Using Theorem 6.29(i) and Corollary 12.30, we obtain  $\nabla(q \circ P_K) = \nabla(q \circ (\text{Id} - P_{K^\ominus})) = \nabla((1/2)d_{K^\ominus}^2) = \text{Id} - P_{K^\ominus} = P_K$ .  $\square$

**Proposition 12.32** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $x \in \mathcal{H}$ . Then the net  $(\gamma f(x))_{\gamma \in \mathbb{R}_{++}}$  is decreasing and the following hold:*

- (i)  $\gamma f(x) \uparrow f(x)$  as  $\gamma \downarrow 0$ .
- (ii)  $\gamma f(x) \downarrow \inf f(\mathcal{H})$  as  $\gamma \uparrow +\infty$ .

*Proof.* In view of Proposition 12.9(ii)&(iv), we need to prove only (i). To this end, set  $\mu = \sup_{\gamma \in \mathbb{R}_{++}} \gamma f(x)$ . It follows from Proposition 12.9(ii) that  $\gamma f(x) \uparrow \mu \leq f(x)$  as  $\gamma \downarrow 0$ . Therefore, we assume that  $\mu < +\infty$ , and it is enough to show that  $\lim_{\gamma \downarrow 0} \gamma f(x) \geq f(x)$ . We deduce from (12.24) that

$$(\forall \gamma \in \mathbb{R}_{++}) \quad \mu \geq \gamma f(x) = f(\text{Prox}_{\gamma f} x) + \frac{1}{2\gamma} \|x - \text{Prox}_{\gamma f} x\|^2. \quad (12.32)$$

Now set  $g = f + (1/2)\|x - \cdot\|^2$ . Then (12.32) implies that  $(\forall \gamma \in ]0, 1[) \text{Prox}_{\gamma f} x \in \text{lev}_{\leq \mu} g$ . Since  $g$  is coercive by Corollary 11.15(i), we derive from Proposition 11.11 that  $\nu = \sup_{\gamma \in ]0, 1[} \|\text{Prox}_{\gamma f} x\| < +\infty$ . On the other hand, Theorem 9.19 asserts that there exist  $u \in \mathcal{H}$  and  $\eta \in \mathbb{R}$  such that  $f \geq \langle \cdot | u \rangle + \eta$ . Therefore, (12.32) yields

$$\begin{aligned} (\forall \gamma \in ]0, 1[) \quad \mu &\geq \langle \text{Prox}_{\gamma f} x | u \rangle + \eta + \frac{1}{2\gamma} \|x - \text{Prox}_{\gamma f} x\|^2 \\ &\geq -\nu \|u\| + \eta + \frac{1}{2\gamma} \|x - \text{Prox}_{\gamma f} x\|^2, \end{aligned} \quad (12.33)$$

from which we deduce that

$$\|x - \text{Prox}_{\gamma f} x\|^2 \leq 2\gamma(\mu + \nu \|u\| - \eta) \rightarrow 0 \quad \text{as } \gamma \downarrow 0. \quad (12.34)$$

In turn, since  $f$  is lower semicontinuous,

$$\begin{aligned} \lim_{\gamma \downarrow 0} \gamma f(x) &= \lim_{\gamma \downarrow 0} f(\text{Prox}_{\gamma f} x) + \frac{1}{2\gamma} \|x - \text{Prox}_{\gamma f} x\|^2 \\ &\geq \underline{\lim}_{\gamma \downarrow 0} f(\text{Prox}_{\gamma f} x) \end{aligned}$$

$$\geq f(x), \quad (12.35)$$

which provides the desired inequality.  $\square$

## 12.5 Infimal Postcomposition

**Definition 12.33** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ , let  $\mathcal{K}$  be a real Hilbert space, and let  $L: \mathcal{H} \rightarrow \mathcal{K}$ . The *infimal postcomposition* of  $f$  by  $L$  is

$$L \triangleright f: \mathcal{K} \rightarrow [-\infty, +\infty]: y \mapsto \inf f(L^{-1}\{y\}) = \inf_{\substack{x \in \mathcal{H} \\ Lx=y}} f(x), \quad (12.36)$$

and it is *exact at a point*  $y \in \mathcal{K}$  if  $(L \triangleright f)(y) = \min_{x \in L^{-1}\{y\}} f(x)$ , i.e.,

$$(\exists x \in \mathcal{H}) \quad Lx = y \quad \text{and} \quad (L \triangleright f)(y) = f(x) \in ]-\infty, +\infty]; \quad (12.37)$$

$L \triangleright f$  is *exact* if it is exact at every point of its domain, in which case it is denoted by  $L \triangleright f$ .

**Proposition 12.34** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $L: \mathcal{H} \rightarrow \mathcal{K}$ , where  $\mathcal{K}$  is a real Hilbert space. Then the following hold:

- (i)  $\text{dom}(L \triangleright f) = L(\text{dom } f)$ .
- (ii) Suppose that  $f$  is convex and that  $L$  is affine. Then  $L \triangleright f$  is convex.

*Proof.* (i): Take  $y \in \mathcal{K}$ . Then  $y \in \text{dom}(L \triangleright f) \Leftrightarrow [(\exists x \in \mathcal{H}) f(x) < +\infty \text{ and } Lx = y] \Leftrightarrow y \in L(\text{dom } f)$ .

(ii): The function  $F: \mathcal{K} \times \mathcal{H} \rightarrow [-\infty, +\infty]: (y, x) \mapsto f(x) + \iota_{\text{gra } L}(x, y)$  is convex and so is its marginal function  $L \triangleright f$  by Proposition 8.26.  $\square$

**Proposition 12.35** Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$  and set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, y) \mapsto x + y$ . Then  $f \square g = L \triangleright (f \oplus g)$ .

*Proof.* A direct consequence of Definition 12.1 and Definition 12.33.  $\square$

## Exercises

**Exercise 12.1** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Determine  $f \square 0$  and  $f \square \iota_{\{0\}}$ .

**Exercise 12.2** Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$ . Show that the set  $\text{epi } f + \text{epi } g$  is closed if and only if  $f \square g$  is lower semicontinuous and exact on  $\{z \in \mathcal{H} \mid (f \square g)(z) > -\infty\}$ .

**Exercise 12.3** Provide continuous and convex functions  $f$  and  $g$  from  $\mathbb{R}$  to  $\mathbb{R}$  such that  $\text{epi } f + \text{epi } g$  is strictly contained in  $\text{epi}(f \square g)$ .

**Exercise 12.4** Let  $f$ ,  $g$ , and  $h$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $f \square g = f \square h$ . Show that it does not follow that  $g = h$ .

**Exercise 12.5** Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ , let  $\phi: \mathbb{R} \rightarrow \mathbb{R}$  be an even convex function, and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \phi(\|P_K x\|)$ . Show that  $f$  is convex. Is this property still true if  $K$  is an arbitrary nonempty closed convex set?

**Exercise 12.6** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\gamma \in \mathbb{R}_{++}$ , let  $\alpha \in \mathbb{R}_{++}$ , and let  $u \in \mathcal{H}$ .

- (i) Set  $g = f^\vee$ . Show that  $\gamma g = (\gamma f)^\vee$ .
- (ii) Set  $g = f + (2\alpha)^{-1} \|\cdot\|^2$  and  $\beta = (\alpha\gamma)/(\alpha + \gamma)$ . Show that

$$\gamma g = \frac{1}{2(\alpha + \gamma)} \|\cdot\|^2 + (\beta f) \left( \frac{\alpha \cdot}{\alpha + \gamma} \right). \quad (12.38)$$

- (iii) Set  $g = f + \langle \cdot | u \rangle$ . Show that  $\gamma g = \langle \cdot | u \rangle - (\gamma/2) \|u\|^2 + \gamma f(\cdot - \gamma u)$ .

**Exercise 12.7** Compute the proximity operator and the Moreau envelope  ${}^1f$  for  $f \in \Gamma_0(\mathbb{R})$  in the following cases:

- (i)  $f = |\cdot|$ .
- (ii)  $f = |\cdot|^3$ .
- (iii)  $f = |\cdot| + |\cdot|^3$  (see [Figure 12.1](#)).

**Exercise 12.8** Let  $f \in \Gamma_0(\mathcal{H})$ , and set  $f_0 = \iota_{\{0\}}$  and  $(\forall n \in \mathbb{N}) \ f_{n+1} = f \square f_n$ . Show that

$$(\forall n \in \mathbb{N} \setminus \{0\}) \quad f_n = n f \left( \frac{\cdot}{n} \right). \quad (12.39)$$

**Exercise 12.9** Let  $\beta \in \mathbb{R}_{++}$ . Show that the  $\beta$ -Pasch–Hausdorff envelope of a convex function from  $\mathcal{H}$  to  $]-\infty, +\infty]$  is convex. In addition, find a function from  $\mathcal{H}$  to  $]-\infty, +\infty]$  that has a nonconvex  $\beta$ -Pasch–Hausdorff envelope.

**Exercise 12.10** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ , and let  $g: \mathcal{H} \rightarrow \mathbb{R}$  be upper semicontinuous. Show that

$$f \square g = \bar{f} \square g, \quad (12.40)$$

where  $\bar{f}$  is the lower semicontinuous envelope of  $f$  defined in (1.42).

**Exercise 12.11** Let  $A \in \mathcal{B}(\mathcal{H})$  be self-adjoint, strictly positive, and surjective, set  $q_A: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \frac{1}{2} \langle x | Ax \rangle$ , and let  $b \in \mathcal{H}$ . Show that

$$(\forall x \in \mathcal{H}) \quad q_A(x) - \langle x | b \rangle = q_A(A^{-1}b) - \langle A^{-1}b | b \rangle + q_A(x - A^{-1}b). \quad (12.41)$$

Deduce that the unique minimizer of the function  $q_A - \langle \cdot | b \rangle$  is  $A^{-1}b$ , at which point the value is  $-q_A(A^{-1}b) = -q_{A^{-1}}(b)$ .

**Exercise 12.12** Let  $A$  and  $B$  be self-adjoint, strictly positive, surjective operators in  $\mathcal{B}(\mathcal{H})$  such that  $A + B$  is surjective, and set  $q_A: x \mapsto \frac{1}{2} \langle x | Ax \rangle$ . Show that

$$q_A \square q_B = q_{(A^{-1}+B^{-1})^{-1}}. \quad (12.42)$$

**Exercise 12.13** For all  $y \in \mathcal{H}$  and  $\alpha \in \mathbb{R}_{++}$ , set  $q_{y,\alpha}: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \frac{1}{2}\|\alpha^{-1}(x - y)\|^2$  if  $\alpha > 0$ ; and  $q_{y,\alpha} = \iota_{\{y\}}$  if  $\alpha = 0$ . Let  $y$  and  $z$  be in  $\mathcal{H}$ , and let  $\alpha$  and  $\beta$  be in  $\mathbb{R}$ . Show that

$$q_{y,\alpha} \square q_{z,\beta} = q_{y+z, \sqrt{\alpha^2 + \beta^2}}. \quad (12.43)$$

**Exercise 12.14** Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Define the *strict epigraph* of  $f$  by

$$\text{epi}_{<} f = \{(x, \xi) \in \mathcal{H} \times \mathbb{R} \mid f(x) < \xi\}. \quad (12.44)$$

Show that  $\text{epi}_{<} f + \text{epi}_{<} g = \text{epi}_{<} (f \square g)$ .

**Exercise 12.15** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ , let  $\mathcal{K}$  be a real Hilbert space, let  $L: \mathcal{H} \rightarrow \mathcal{K}$ , and use the same notation as in (12.44). Show that  $\text{epi}_{<} (L \triangleright f) = (L \times \text{Id})(\text{epi}_{<} f)$ .

# Chapter 13

## Conjugation

In classical analysis, functional transforms make it possible to investigate problems from a different perspective and sometimes simplify their analysis. In convex analysis, the most suitable notion of a transform is the Legendre transform, which maps a function to its (Fenchel) conjugate. This transform is studied in detail in this chapter. In particular, it is shown that the conjugate of an infimal convolution is the sum of the conjugates. The key result of this chapter is the Fenchel–Moreau theorem, which states that the proper convex lower semicontinuous functions are precisely those functions that coincide with their biconjugates.

### 13.1 Definition and Examples

**Definition 13.1** Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . The *conjugate* (or *Legendre transform*, or *Legendre–Fenchel transform*, or *Fenchel conjugate*) of  $f$  is

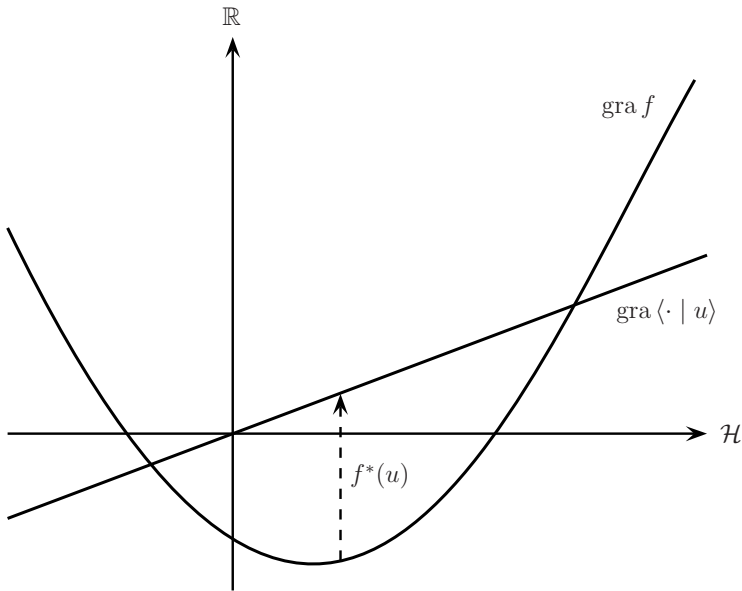
$$f^*: \mathcal{H} \rightarrow [-\infty, +\infty] : u \mapsto \sup_{x \in \mathcal{H}} (\langle x | u \rangle - f(x)), \quad (13.1)$$

and the *biconjugate* of  $f$  is  $f^{**} = (f^*)^*$ .

Let us illustrate Definition 13.1 through a variety of examples (see also [Figure 13.1](#)).

**Example 13.2** These examples concern the case  $\mathcal{H} = \mathbb{R}$ .

- (i) Let  $p \in ]1, +\infty[$  and set  $p^* = p/(p-1)$ . Then  $((1/p)| \cdot |^p)^* = (1/p^*)| \cdot |^{p^*}$ .
- (ii) Let  $f: x \mapsto \begin{cases} 1/x, & \text{if } x > 0; \\ +\infty, & \text{if } x \leq 0. \end{cases}$



**Fig. 13.1**  $f^*(u)$  is the supremum of the signed vertical difference between the graph of  $f$  and that of the continuous linear functional  $\langle \cdot | u \rangle$ .

$$\text{Then } f^*: u \mapsto \begin{cases} -2\sqrt{-u}, & \text{if } u \leq 0; \\ +\infty, & \text{if } u > 0. \end{cases}$$

$$\text{(iii) Let } f: x \mapsto \begin{cases} -\ln(x), & \text{if } x > 0; \\ +\infty, & \text{if } x \leq 0. \end{cases}$$

$$\text{Then } f^*: u \mapsto \begin{cases} -\ln(-u) - 1, & \text{if } u < 0; \\ +\infty, & \text{if } u \geq 0. \end{cases}$$

$$\text{(iv) } \cosh^*: u \mapsto u \operatorname{arcsinh}(u) - \sqrt{u^2 + 1}.$$

$$\text{(v) } \exp^*: u \mapsto \begin{cases} u \ln(u) - u, & \text{if } u > 0; \\ 0, & \text{if } u = 0; \\ +\infty, & \text{if } u < 0. \end{cases}$$

$$\text{(vi) Let } f: x \mapsto \sqrt{1 + x^2}. \text{ Then } f^*: u \mapsto \begin{cases} -\sqrt{1 - u^2}, & \text{if } |u| \leq 1; \\ +\infty, & \text{if } |u| > 1. \end{cases}$$

*Proof.* Exercise 13.1. □



**Example 13.3** Below are some direct applications of (13.1).

- (i) Let  $f = \iota_C$ , where  $C \subset \mathcal{H}$ . Then (7.4) yields  $f^* = \sigma_C$ .
- (ii) Let  $f = \iota_K$ , where  $K$  is a nonempty cone in  $\mathcal{H}$ . Then (i) yields  $f^* = \sigma_K = \iota_{K^\ominus}$ .
- (iii) Let  $f = \iota_V$ , where  $V$  is a linear subspace of  $\mathcal{H}$ . Then (ii) yields  $f^* = \iota_{V^\ominus} = \iota_{V^\perp}$ .
- (iv) Let  $f = \iota_{B(0;1)}$ . Then (i) yields  $f^* = \sigma_{B(0;1)} = \sup_{\|x\| \leq 1} \langle x | \cdot \rangle = \|\cdot\|$ .
- (v) Let  $f = \|\cdot\|$ . Then  $f^* = \iota_{B(0;1)}$ .

*Proof.* (i)–(iv): Exercise 13.2.

(v): Let  $u \in \mathcal{H}$ . If  $\|u\| \leq 1$ , then Cauchy–Schwarz yields  $0 = \langle 0 | u \rangle - \|0\| \leq \sup_{x \in \mathcal{H}} (\langle x | u \rangle - \|x\|) \leq \sup_{x \in \mathcal{H}} (\|x\|(\|u\| - 1)) = 0$ . Therefore  $f^*(u) = 0$ . On the other hand, if  $\|u\| > 1$ , then  $\sup_{x \in \mathcal{H}} (\langle x | u \rangle - \|x\|) \geq \sup_{\lambda \in \mathbb{R}_{++}} (\langle \lambda u | u \rangle - \|\lambda u\|) = \|u\|(\|u\| - 1) \sup_{\lambda \in \mathbb{R}_{++}} \lambda = +\infty$ . Altogether,  $f^* = \iota_{B(0;1)}$ .  $\square$

**Example 13.4** Let  $\varphi: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $\gamma \in \mathbb{R}_{++}$ , and set  $f = \varphi + \gamma^{-1}q$ , where  $q = (1/2)\|\cdot\|^2$ . Then

$$f^* = \gamma q - \gamma \varphi \circ \gamma \text{Id} = \gamma q - (\varphi \square (\gamma^{-1}q)) \circ \gamma \text{Id}. \quad (13.2)$$

*Proof.* Let  $u \in \mathcal{H}$ . Definition 12.20 yields

$$\begin{aligned} f^*(u) &= - \inf_{x \in \mathcal{H}} (f(x) - \langle x | u \rangle) \\ &= \frac{\gamma}{2} \|u\|^2 - \inf_{x \in \mathcal{H}} \left( \varphi(x) + \frac{1}{2\gamma} \|x - \gamma u\|^2 \right) \\ &= \frac{\gamma}{2} \|u\|^2 - \gamma \varphi(\gamma u), \end{aligned} \quad (13.3)$$

which provides (13.2).  $\square$

**Example 13.5** Let  $C$  be a nonempty subset of  $\mathcal{H}$  and let  $f = \iota_C + \|\cdot\|^2/2$ . Then  $f^* = (\|\cdot\|^2 - d_C^2)/2$ .

*Proof.* Set  $\varphi = \iota_C$  and  $\gamma = 1$  in Example 13.4.  $\square$

**Example 13.6** Set  $f = (1/2)\|\cdot\|^2$ . Then  $f^* = f$ .

*Proof.* Set  $C = \mathcal{H}$  in Example 13.5.  $\square$

**Example 13.7** Let  $\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be even. Then  $(\phi \circ \|\cdot\|)^* = \phi^* \circ \|\cdot\|$ .

*Proof.* If  $\mathcal{H} = \{0\}$ , then  $(\phi \circ \|\cdot\|)^*(0) = -\phi(0) = (\phi^* \circ \|\cdot\|)(0)$ . Now assume that  $\mathcal{H} \neq \{0\}$ . Then, for every  $u \in \mathcal{H}$ ,

$$\begin{aligned} (\phi \circ \|\cdot\|)^*(u) &= \sup_{\rho \in \mathbb{R}_+} \sup_{\|x\|=1} (\langle \rho x | u \rangle - \phi(\|\rho x\|)) \\ &= \sup_{\rho \in \mathbb{R}_+} (\rho \|u\| - \phi(\rho)) \end{aligned}$$

$$\begin{aligned}
&= \sup_{\rho \in \mathbb{R}} (\rho \|u\| - \phi(\rho)) \\
&= \phi^*(\|u\|),
\end{aligned} \tag{13.4}$$

as required.  $\square$

**Example 13.8** Let  $\varphi: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and let  $f$  be its perspective function, defined in (8.18). Then  $f^* = \iota_C$ , where  $C = \{(\nu, u) \in \mathbb{R} \times \mathcal{H} \mid \nu + \varphi^*(u) \leq 0\}$ .

*Proof.* Let  $\nu \in \mathbb{R}$  and  $u \in \mathcal{H}$ . It follows from (8.18) that

$$\begin{aligned}
f^*(\nu, u) &= \sup_{\xi \in \mathbb{R}_{++}} \left( \sup_{x \in \mathcal{H}} \xi \nu + \langle x \mid u \rangle - \xi \varphi(x/\xi) \right) \\
&= \sup_{\xi \in \mathbb{R}_{++}} \xi \left( \nu + \sup_{x \in \mathcal{H}} \langle x/\xi \mid u \rangle - \varphi(x/\xi) \right) \\
&= \sup_{\xi \in \mathbb{R}_{++}} \xi (\nu + \varphi^*(u)) \\
&= \begin{cases} 0, & \text{if } \nu + \varphi^*(u) \leq 0; \\ +\infty, & \text{otherwise,} \end{cases}
\end{aligned} \tag{13.5}$$

which completes the proof.  $\square$

## 13.2 Basic Properties

Let us first record some immediate consequences of Definition 13.1.

**Proposition 13.9** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then the following hold:*

- (i)  $f^*(0) = -\inf f(\mathcal{H})$ .
- (ii)  $-\infty \in f^*(\mathcal{H}) \Leftrightarrow f \equiv +\infty \Leftrightarrow f^* \equiv -\infty$ .
- (iii) *Suppose that  $f^*$  is proper. Then  $f$  is proper.*
- (iv) *Let  $u \in \mathcal{H}$ . Then*

$$f^*(u) = \sup_{x \in \text{dom } f} (\langle x \mid u \rangle - f(x)) = \sup_{(x, \xi) \in \text{epi } f} (\langle x \mid u \rangle - \xi).$$

- (v)  $f^* = \iota_{\text{gra } f}^*(\cdot, -1) = \iota_{\text{epi } f}^*(\cdot, -1)$ .

**Proposition 13.10** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then the following hold:*

- (i) *Let  $(u, \nu) \in \mathcal{H} \times \mathbb{R}$ . Then  $(u, \nu) \in \text{epi } f^* \Leftrightarrow \langle \cdot \mid u \rangle - \nu \leq f$ .*
- (ii)  *$f^* \equiv +\infty$  if and only if  $f$  possesses no continuous affine minorant.*
- (iii) *Suppose that  $\text{dom } f^* \neq \emptyset$ . Then  $f$  is bounded below on every bounded subset of  $\mathcal{H}$ .*

*Proof.* (i):  $(u, \nu) \in \text{epi } f^* \Leftrightarrow f^*(u) \leq \nu \Leftrightarrow (\forall x \in \mathcal{H}) \langle x | u \rangle - f(x) \leq \nu$ .

(ii): By (i), there is a bijection between the points of the epigraph of  $f^*$  and the continuous affine minorants of  $f$ , and  $\text{epi } f^* = \emptyset \Leftrightarrow f^* \equiv +\infty$ .

(iii): By (ii), if  $\text{dom } f^* \neq \emptyset$ , then  $f$  possesses a continuous affine minorant, say  $\langle \cdot | u \rangle + \nu$ . Now let  $C$  be a bounded set in  $\mathcal{H}$  and let  $\beta = \sup_{x \in C} \|x\|$ . Then, by Cauchy–Schwarz,  $(\forall x \in C) f(x) \geq \langle x | u \rangle + \nu \geq -\|x\| \|u\| + \nu \geq -\beta \|u\| - \nu > -\infty$ .  $\square$

**Proposition 13.11** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f^* \in \Gamma(\mathcal{H})$ .*

*Proof.* We assume that  $f \not\equiv +\infty$ . By Proposition 13.9(iv),  $f^*$  is the supremum of the lower semicontinuous convex functions  $(\langle x | \cdot \rangle - \xi)_{(x, \xi) \in \text{epi } f}$ . The result therefore follows from Proposition 9.3.  $\square$

**Example 13.12** Let  $\varphi: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and let  $\gamma \in \mathbb{R}_{++}$ . Then, by Proposition 13.11,  $(\gamma/2)\|\cdot\|^2 - \gamma\varphi(\gamma\cdot)$  is lower semicontinuous and convex as a conjugate function (see Example 13.4). Likewise, it follows from Example 13.5 that, for every nonempty subset  $C$  of  $\mathcal{H}$ ,  $\|\cdot\|^2 - d_C^2 \in \Gamma_0(\mathcal{H})$ .

**Proposition 13.13 (Fenchel–Young inequality)** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then*

$$(\forall x \in \mathcal{H})(\forall u \in \mathcal{H}) \quad f(x) + f^*(u) \geq \langle x | u \rangle. \quad (13.6)$$

*Proof.* Fix  $x$  and  $u$  in  $\mathcal{H}$ . Since  $f$  is proper, it follows from Proposition 13.9(ii) that  $-\infty \notin f^*(\mathcal{H})$ . Thus, if  $f(x) = +\infty$ , the inequality trivially holds. On the other hand, if  $f(x) < +\infty$ , then (13.1) yields  $f^*(u) \geq \langle x | u \rangle - f(x)$  and the inequality follows.  $\square$

**Proposition 13.14** *Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $[-\infty, +\infty]$ . Then the following hold:*

- (i)  $f^{**} \leq f$ .
- (ii)  $f \leq g \Rightarrow [f^* \geq g^* \text{ and } f^{**} \leq g^{**}]$ .
- (iii)  $f^{***} = f^*$ .
- (iv)  $(\check{f})^* = f^*$ .

*Proof.* (i) and (ii): Direct consequences of (13.1).

(iii): It follows from (i) and (ii) that  $f^{***} \geq f^*$ . On the other hand, (i) applied to  $f^*$  yields  $f^{***} \leq f^*$ .

(iv): It follows from Proposition 13.11 and Definition 9.7 that  $f^{**} \leq \check{f} \leq f$ . Hence, by (iii) and (ii),  $f^* = f^{***} \geq (\check{f})^* \geq f^*$ .  $\square$

**Example 13.15** Set  $f: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto -|x|$ . Then  $\check{f} \equiv -\infty$  and  $f^* \equiv +\infty$ .

**Proposition 13.16 (self-conjugacy)** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then*

$$f = f^* \quad \Leftrightarrow \quad f = (1/2)\|\cdot\|^2. \quad (13.7)$$

*Proof.* Set  $q = (1/2)\|\cdot\|^2$ . Then the identity  $q^* = q$  is known from Example 13.6. Conversely, if  $f = f^*$ , then  $f$  is proper by virtue of Proposition 13.9, and the Fenchel–Young inequality (Proposition 13.13) yields  $(\forall x \in \mathcal{H}) \ 2f(x) \geq \langle x | x \rangle$ , i.e.,  $f \geq q$ . Therefore, by Proposition 13.14(ii),  $q = q^* \geq f^* = f$  and we conclude that  $f = q$ .  $\square$

**Remark 13.17**

- (i) Suppose that  $\mathcal{H} \neq \{0\}$ . Since  $(1/2)\|\cdot\|^2$  is the only self-conjugate function, a convex cone  $K$  in  $\mathcal{H}$  cannot be self-polar since  $K^\ominus = K \Leftrightarrow \iota_K^* = \iota_K$ . In particular, we recover the well-known fact that a linear subspace of  $\mathcal{H}$  cannot be self-orthogonal.
- (ii) Let  $f = \iota_K$ , where  $K$  is a self-dual closed convex cone in  $\mathcal{H}$ . Then  $f^* = \sigma_K = \iota_{K^\ominus} = \iota_{K^\oplus}^\vee = f^\vee$ . Another function that satisfies the identity  $f^* = f^\vee$  is

$$f: \mathcal{H} = \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} -\ln(x) - \frac{1}{2}, & \text{if } x > 0; \\ +\infty, & \text{if } x \leq 0. \end{cases} \quad (13.8)$$

**Proposition 13.18** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be even. Then  $f^*$  is even.*

*Proof.* Let  $u \in \mathcal{H}$ . Then  $f^*(-u) = \sup_{x \in \mathcal{H}} \langle x | -u \rangle - f(x) = \sup_{x \in \mathcal{H}} \langle x | u \rangle - f(-x) = \sup_{x \in \mathcal{H}} \langle x | u \rangle - f(x) = f^*(u)$ .  $\square$

**Proposition 13.19** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be such that  $f \geq f(0) = 0$ . Then  $f^* \geq f^*(0) = 0$ .*

*Proof.* We have  $f^*(0) = -\inf f(\mathcal{H}) = -f(0) = 0$ . Moreover, for every  $u \in \mathcal{H}$ , (13.1) yields  $f^*(u) \geq \langle 0 | u \rangle - f(0) = 0$ .  $\square$

**Proposition 13.20** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ . Then the following hold:*

- (i)  $(\forall \alpha \in \mathbb{R}_{++}) \ (\alpha f)^* = \alpha f^*(\cdot/\alpha)$ .
- (ii)  $(\forall \alpha \in \mathbb{R}_{++}) \ (\alpha f(\cdot/\alpha))^* = \alpha f^*$ .
- (iii)  $(\forall y \in \mathcal{H})(\forall v \in \mathcal{H})(\forall \alpha \in \mathbb{R}) \ (\tau_y f + \langle \cdot | v \rangle + \alpha)^* = \tau_v f^* + \langle y | \cdot \rangle - \langle y | v \rangle - \alpha$ .
- (iv) Let  $L \in \mathcal{B}(\mathcal{H})$  be bijective. Then  $(f \circ L)^* = f^* \circ L^{*-1}$ .
- (v)  $f^{\vee*} = f^{*\vee}$ .
- (vi) Let  $V$  be a closed linear subspace of  $\mathcal{H}$  such that  $\text{dom } f \subset V$ . Then  $(f|_V)^* \circ P_V = f^* = f^* \circ P_V$ .

*Proof.* (i)–(iv): Straightforward from (13.1).

(v): Take  $L = -\text{Id}$  in (iv).

(vi): Let  $u \in \mathcal{H}$ . Then  $f^*(u) = \sup_{x \in V} \langle x | u \rangle - f(x) = \sup_{x \in V} \langle P_V x | u \rangle - f|_V(x) = \sup_{x \in V} \langle x | P_V u \rangle - f|_V(x) = (f|_V)^*(P_V u)$ . Consequently, we obtain  $(f|_V)^*(P_V u) = (f|_V)^*(P_V P_V u) = f^*(P_V u)$ .  $\square$

**Proposition 13.21** *Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Then the following hold:*

- (i)  $(f \square g)^* = f^* + g^*$ .
- (ii) *Suppose that  $f$  and  $g$  are proper. Then  $(f + g)^* \leq f^* \square g^*$ .*
- (iii)  $(\forall \gamma \in \mathbb{R}_{++}) (\gamma f)^* = f^* + (\gamma/2) \|\cdot\|^2$ .
- (iv) *Let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then  $(L \triangleright f)^* = f^* \circ L^*$ .*
- (v) *Let  $L \in \mathcal{B}(\mathcal{K}, \mathcal{H})$ . Then  $(f \circ L)^* \leq L^* \triangleright f^*$ .*

*Proof.* (i): For every  $u \in \mathcal{H}$ , we have

$$\begin{aligned} (f \square g)^*(u) &= \sup_{x \in \mathcal{H}} \left( \langle x \mid u \rangle - \inf_{y \in \mathcal{H}} (f(y) + g(x - y)) \right) \\ &= \sup_{y \in \mathcal{H}} \left( \langle y \mid u \rangle - f(y) + \sup_{x \in \mathcal{H}} (\langle x - y \mid u \rangle - g(x - y)) \right) \\ &= f^*(u) + g^*(u). \end{aligned} \quad (13.9)$$

(ii): Items (i) and (ii) in Proposition 13.14 yield successively  $f + g \geq f^{**} + g^{**}$  and  $(f + g)^* \leq (f^{**} + g^{**})^*$ . However, by (i) above,  $(f^{**} + g^{**})^* = (f^* \square g^*)^{**} \leq f^* \square g^*$ .

(iii): Take  $g = \|\cdot\|^2/(2\gamma)$  in (i).

(iv): Let  $v \in \mathcal{K}$ . Then

$$\begin{aligned} (L \triangleright f)^*(v) &= \sup_{y \in \mathcal{K}} \left( \langle y \mid v \rangle - \inf_{Lx=y} f(x) \right) \\ &= \sup_{y \in \mathcal{K}} \sup_{x \in L^{-1}\{y\}} (\langle Lx \mid v \rangle - f(x)) \\ &= \sup_{x \in \mathcal{H}} (\langle x \mid L^*v \rangle - f(x)) \\ &= f^*(L^*v). \end{aligned} \quad (13.10)$$

(v): By (iv) and Proposition 13.14(i),  $(L^* \triangleright f^*)^* = f^{**} \circ L^{**} = f^{**} \circ L \leq f \circ L$ . Hence, by Proposition 13.14(ii),  $L^* \triangleright f^* \geq (L^* \triangleright f^*)^{**} \geq (f \circ L)^*$ .  $\square$

**Corollary 13.22** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ , let  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then the following hold:*

- (i) *Suppose that  $-\infty \notin (L \triangleright f)(\mathcal{K})$ . Then  $((L \triangleright f) \square g)^* = (f^* \circ L^*) + g^*$ .*
- (ii) *Suppose that  $\text{dom } f \neq \emptyset$  and  $\text{dom } g \cap \text{ran } L \neq \emptyset$ . Then  $(f + (g \circ L))^* \leq f^* \square (L^* \triangleright g^*)$ .*

*Proof.* (i): By Proposition 13.21(i)&(iv),  $((L \triangleright f) \square g)^* = (L \triangleright f)^* + g^* = (f^* \circ L^*) + g^*$ .

(ii): By Proposition 13.21(ii)&(v),  $(f + (g \circ L))^* \leq f^* \square (g \circ L)^* \leq f^* \square (L^* \triangleright g^*)$ .  $\square$

**Example 13.23** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be increasing on  $\mathbb{R}_+$  and even, and set  $f = \phi \circ d_C$ . Then  $f^* = \sigma_C + \phi^* \circ \|\cdot\|$ .

*Proof.* Since  $\phi$  is increasing on  $\mathbb{R}_+$ , we have for every  $x \in \mathcal{H}$  and every  $y \in C$ ,

$$\inf_{z \in C} \phi(\|x - z\|) \leq \phi(\|x - P_C x\|) = \phi\left(\inf_{z \in C} \|x - z\|\right) \leq \phi(\|x - y\|). \quad (13.11)$$

Taking the infimum over  $y \in C$  then yields  $(\forall x \in \mathcal{H}) \inf_{z \in C} \phi(\|x - z\|) = \phi(\inf_{z \in C} \|x - z\|)$ . Thus,  $f = \iota_C \square (\phi \circ \|\cdot\|)$ . In turn, since  $\phi$  is even, we derive from Proposition 13.21(i), Example 13.3(i), and Example 13.7 that

$$f^* = (\iota_C \square (\phi \circ \|\cdot\|))^* = \iota_C^* + (\phi \circ \|\cdot\|)^* = \sigma_C + \phi^* \circ \|\cdot\|, \quad (13.12)$$

which completes the proof.  $\square$

**Example 13.24** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $p \in ]1, +\infty[$ , and set  $p^* = p/(p-1)$ .

- (i) Setting  $\phi = |\cdot|$  in Example 13.23 and using Example 13.3(v) yields  $d_C^* = \sigma_C + \iota_{B(0;1)}$ .
- (ii) If  $V$  is a closed linear subspace of  $\mathcal{H}$ , then (i) and Example 13.3(iii) yield  $d_V^* = \iota_{V^\perp \cap B(0;1)}$ .
- (iii) In view of Example 13.2(i), setting  $\phi = (1/p)|\cdot|^p$  in Example 13.23 yields  $((1/p)d_C^p)^* = \sigma_C + (1/p^*)\|\cdot\|^{p^*}$ .
- (iv) Setting  $C = \{0\}$  in (iii) yields  $((1/p)\|\cdot\|^p)^* = (1/p^*)\|\cdot\|^{p^*}$ .

Here is a generalization of Proposition 13.14(ii).

**Proposition 13.25** Let  $(f_i)_{i \in I}$  be a family of proper functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ . Then the following hold:

- (i)  $(\inf_{i \in I} f_i)^* = \sup_{i \in I} f_i^*$ .
- (ii)  $(\sup_{i \in I} f_i)^* \leq \inf_{i \in I} f_i^*$ .

*Proof.* (i): By definition,

$$\begin{aligned} (\forall u \in \mathcal{H}) \quad \left( \inf_{i \in I} f_i \right)^*(u) &= \sup_{x \in \mathcal{H}} \left( \langle x | u \rangle + \sup_{i \in I} -f_i(x) \right) \\ &= \sup_{i \in I} \sup_{x \in \mathcal{H}} \left( \langle x | u \rangle - f_i(x) \right) \\ &= \sup_{i \in I} f_i^*(u). \end{aligned} \quad (13.13)$$

(ii): Set  $g = \sup_{i \in I} f_i$ . For every  $i \in I$ ,  $f_i \leq g$  and therefore  $g^* \leq f_i^*$  by Proposition 13.14(ii). Hence,  $g^* \leq \inf_{i \in I} f_i^*$ .  $\square$

**Proposition 13.26** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $\gamma \in \mathbb{R}_{++}$ , and set  $g = (1/2)\|\cdot\|^2$ . Then

$$(\gamma f - q)^* = \gamma(\gamma q - f^*)^* - q \quad (13.14)$$

and  $(\gamma q - f^*)^*$  is  $\gamma^{-1}$ -strongly convex.

*Proof.* Set  $\varphi = \gamma q - f^*$ . Then

$$\varphi = \gamma q - \sup_{u \in \mathcal{H}} (\langle \cdot | u \rangle - f(u)) = \inf_{u \in \mathcal{H}} (\gamma q - \langle \cdot | u \rangle + f(u)). \quad (13.15)$$

Therefore, we derive from Proposition 13.25(i) that

$$\begin{aligned} \gamma \varphi^* &= \gamma \sup_{u \in \mathcal{H}} (\gamma q - \langle \cdot | u \rangle + f(u))^* \\ &= \sup_{u \in \mathcal{H}} (\gamma(\gamma q - \langle \cdot | u \rangle)^* - \gamma f(u)) \\ &= \sup_{u \in \mathcal{H}} (q(\cdot + u) - \gamma f(u)) \\ &= q + \sup_{u \in \mathcal{H}} (\langle \cdot | u \rangle - (\gamma f - q)(u)) \\ &= q + (\gamma f - q)^*, \end{aligned} \quad (13.16)$$

which yields (13.14). In turn,  $(\gamma q - f^*)^* - \gamma^{-1}q = \gamma^{-1}(\gamma f - q)^*$  is convex by Proposition 13.11. Hence, the second claim follows from Proposition 10.6.  $\square$

**Proposition 13.27** *Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces and, for every  $i \in I$ , let  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$ . Then*

$$\left( \bigoplus_{i \in I} f_i \right)^* = \bigoplus_{i \in I} f_i^*. \quad (13.17)$$

*Proof.* Set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$  and  $\mathbf{f} = \bigoplus_{i \in I} f_i$ , and let  $\mathbf{u} = (u_i)_{i \in I} \in \mathcal{H}$ . We have

$$\begin{aligned} \mathbf{f}^*(\mathbf{u}) &= \sup_{\mathbf{x} \in \mathcal{H}} (\langle \mathbf{u} | \mathbf{x} \rangle - \mathbf{f}(\mathbf{x})) \\ &= \sup_{\mathbf{x} \in \mathcal{H}} \left( \sum_{i \in I} \langle u_i | x_i \rangle - \sum_{i \in I} f_i(x_i) \right) \\ &= \sum_{i \in I} \sup_{x_i \in \mathcal{H}_i} (\langle u_i | x_i \rangle - f_i(x_i)) \\ &= \sum_{i \in I} f_i^*(u_i), \end{aligned} \quad (13.18)$$

and we obtain the conclusion.  $\square$

The remainder of this section concerns the conjugates of bivariate functions.

**Proposition 13.28** *Let  $\mathcal{K}$  be a real Hilbert space, let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$  be a proper function, and let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]: x \mapsto \inf F(x, \mathcal{K})$ . Then  $f^* = F^*(\cdot, 0)$ .*

*Proof.* Fix  $u \in \mathcal{H}$ . Then

$$\begin{aligned} f^*(u) &= \sup_{x \in \mathcal{H}} \left( \langle x \mid u \rangle - \inf_{y \in \mathcal{K}} F(x, y) \right) \\ &= \sup_{x \in \mathcal{H}} \sup_{y \in \mathcal{K}} \left( \langle x \mid u \rangle + \langle y \mid 0 \rangle - F(x, y) \right) \\ &= \sup_{(x, y) \in \mathcal{H} \times \mathcal{K}} \left( \langle (x, y) \mid (u, 0) \rangle - F(x, y) \right) \\ &= F^*(u, 0), \end{aligned} \tag{13.19}$$

which establishes the identity.  $\square$

**Definition 13.29** Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $F$  is *autoconjugate* if  $F^* = F^\top$ , where  $F^\top: \mathcal{H} \times \mathcal{H}: (u, x) \mapsto F(x, u)$ .

**Proposition 13.30** *Let  $F \in \Gamma(\mathcal{H} \times \mathcal{H})$ . Then  $F^{*\top} = F^{\top*}$ .*

*Proof.* Exercise 13.9.  $\square$

**Proposition 13.31** *Let  $F \in \Gamma(\mathcal{H} \times \mathcal{H})$  be autoconjugate. Then  $F \geq \langle \cdot \mid \cdot \rangle$  and  $F^* \geq \langle \cdot \mid \cdot \rangle$ .*

*Proof.* Take  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then, the Fenchel–Young inequality (Proposition 13.13) yields  $2F^*(u, x) = 2F(x, u) = F(x, u) + F^\top(u, x) = F(x, u) + F^*(u, x) \geq \langle (x, u) \mid (u, x) \rangle = 2 \langle x \mid u \rangle$  and the result follows.  $\square$

### 13.3 The Fenchel–Moreau Theorem

As seen in Proposition 13.14(i), a function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  is related to its biconjugate via the inequality  $f^{**} \leq f$ . In general  $f^{**} \neq f$  since the left-hand side is always lower semicontinuous and convex by Proposition 13.11, while the right-hand side need not be. The next theorem characterizes those functions that coincide with their biconjugates.

**Theorem 13.32 (Fenchel–Moreau)** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $f$  is lower semicontinuous and convex if and only if  $f = f^{**}$ . In this case,  $f^*$  is proper as well.*

*Proof.* If  $f = f^{**}$ , then  $f$  is lower semicontinuous and convex as a conjugate function by Proposition 13.11. Conversely, suppose that  $f$  is lower semicontinuous and convex. Fix  $x \in \mathcal{H}$ , take  $\xi \in ]-\infty, f(x)[$ , and set  $(p, \pi) = P_{\text{epi } f}(x, \xi)$ . Proposition 9.17 states that



$$\pi \geq \xi \quad \text{and} \quad (\forall y \in \text{dom } f) \quad \langle y - p \mid x - p \rangle \leq (\pi - \xi)(f(y) - \pi). \quad (13.20)$$

If  $\pi > \xi$ , then upon setting  $v = (x - p)/(\pi - \xi)$ , we get from (13.20) that

$$\begin{aligned} (\forall y \in \text{dom } f) \quad \langle y \mid v \rangle - f(y) &\leq \langle p \mid v \rangle - \pi \\ &= \langle x \mid v \rangle - (\pi - \xi)\|v\|^2 - \pi \\ &\leq \langle x \mid v \rangle - \pi. \end{aligned} \quad (13.21)$$

However, (13.21) implies that  $f^*(v) \leq \langle x \mid v \rangle - \pi$  and, in turn, that  $\pi \leq \langle x \mid v \rangle - f^*(v) \leq f^{**}(x)$ . To sum up,

$$\pi > \xi \quad \Rightarrow \quad f^{**}(x) > \xi. \quad (13.22)$$

We now show that  $f^{**}(x) = f(x)$ . Since  $\text{dom } f \neq \emptyset$ , we first consider the case  $x \in \text{dom } f$ . Then (9.15) yields  $\pi = f(p) > \xi$ , and it follows from (13.22) and Proposition 13.14(i) that  $f(x) \geq f^{**}(x) > \xi$ . Since  $\xi$  can be chosen arbitrarily in  $]-\infty, f(x)[$ , we get  $f^{**}(x) = f(x)$ . Thus,  $f$  and  $f^{**}$  coincide on  $\text{dom } f \neq \emptyset$ . Therefore  $+\infty \not\equiv f^{**} \not\equiv -\infty$  and it follows from Proposition 13.9(ii) that  $-\infty \notin f^*(\mathcal{H}) \neq \{+\infty\}$ , i.e., that  $f^*$  is proper. Now suppose that  $x \notin \text{dom } f$ . If  $\pi > \xi$ , it follows from (13.22) that  $f^{**}(x) > \xi$  and, since  $\xi$  can be any real number, we obtain  $f^{**}(x) = +\infty = f(x)$ . Otherwise,  $\pi = \xi$  and, since  $(x, \xi) \notin \text{epi } f$  and  $(p, \pi) \in \text{epi } f$ , we have  $\|x - p\| > 0$ . Now, fix  $w \in \text{dom } f^*$  and set  $u = x - p$ . Then it follows from (13.1) and (13.20) that

$$(\forall y \in \text{dom } f) \quad \langle y \mid w \rangle - f(y) \leq f^*(w) \quad \text{and} \quad \langle y \mid u \rangle \leq \langle p \mid u \rangle. \quad (13.23)$$

Next, let  $\lambda \in \mathbb{R}_{++}$ . Then (13.23) yields

$$(\forall y \in \text{dom } f) \quad \langle y \mid w + \lambda u \rangle - f(y) \leq f^*(w) + \langle p \mid \lambda u \rangle. \quad (13.24)$$

Hence,  $f^*(w + \lambda u) \leq f^*(w) + \langle \lambda u \mid p \rangle = f^*(w) + \langle w + \lambda u \mid x \rangle - \langle w \mid x \rangle - \lambda\|u\|^2$ . Consequently,  $f^{**}(x) \geq \langle w + \lambda u \mid x \rangle - f^*(w + \lambda u) \geq \langle w \mid x \rangle + \lambda\|u\|^2 - f^*(w)$ . Since  $\lambda$  can be arbitrarily large, we must have  $f^{**}(x) = +\infty$ .  $\square$

**Corollary 13.33** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f^* \in \Gamma_0(\mathcal{H})$  and  $f^{**} = f$ .*

*Proof.* Combine Theorem 13.32 and Proposition 13.11.  $\square$

**Corollary 13.34** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $g: \mathcal{H} \rightarrow [-\infty, +\infty]$ . Then  $f \leq g \Leftrightarrow f^* \geq g^*$ .*

*Proof.* Proposition 13.14(ii), Corollary 13.33, and Proposition 13.14(i) yield  $f \leq g \Rightarrow f^* \geq g^* \Rightarrow f = f^{**} \leq g^{**} \leq g$ .  $\square$

**Corollary 13.35** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Then  $f = g^* \Leftrightarrow g = f^*$ .*

**Corollary 13.36** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f$  is the supremum of its continuous affine minorants.*

*Proof.* As seen in Proposition 13.10(i), a continuous affine minorant of  $f$  assumes the form  $\langle \cdot | u \rangle - \nu$ , where  $(u, \nu) \in \text{epi } f^*$ . On the other hand, it follows from Theorem 13.32 and Proposition 13.9(iv) that  $(\forall x \in \mathcal{H}) f(x) = f^{**}(x) = \sup_{(u, \nu) \in \text{epi } f^*} (\langle u | x \rangle - \nu)$ .  $\square$

**Example 13.37** Here are some consequences of Theorem 13.32.

- (i) Let  $C$  be a nonempty subset of  $\mathcal{H}$ . Then it follows from Example 13.3(i) and Proposition 7.11 that  $\iota_C^{**} = \sigma_C^* = \sigma_{\overline{\text{conv}} C}^* = \iota_{\overline{\text{conv}} C}^{**} = \iota_{\overline{\text{conv}} C}$ .
- (ii) Let  $V$  be a closed linear subspace of  $\mathcal{H}$ . Then  $\iota_V^{**} = \iota_V$ . On the other hand,  $\iota_V^* = \sigma_V = \iota_{V^\perp}$  and therefore  $\iota_V^{**} = \sigma_V^* = \iota_{V^\perp}^* = \iota_{V^{\perp\perp}}$ . Altogether, we recover the well-known identity  $V^{\perp\perp} = V$ .
- (iii) More generally, let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ . Then  $\iota_K^{**} = \iota_K$ . On the other hand,  $\iota_K^* = \sigma_K = \iota_{K^\ominus}$  and therefore  $\iota_K^{**} = \sigma_K^* = \iota_{K^\ominus}^* = \iota_{K^{\ominus\ominus}}$ . Thus, we recover Corollary 6.33, namely  $K^{\ominus\ominus} = K$ .

**Proposition 13.38** *Let  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  be convex and let  $x \in \mathcal{H}$ . Suppose that  $f(x) \in \mathbb{R}$ . Then the following are equivalent:*

- (i)  $f$  is lower semicontinuous at  $x$ .
- (ii)  $f^{**}(x) = \check{f}(x) = \bar{f}(x) = f(x)$ .
- (iii)  $f^{**}(x) = f(x)$ .

Moreover, each of the above implies that  $f \geq \bar{f} = \check{f} = f^{**} \in \Gamma_0(\mathcal{H})$ .

*Proof.* In view of Proposition 13.11, Proposition 13.14(i), Proposition 9.8(i), and Corollary 9.10, we have  $f^{**} \leq \check{f} = \bar{f} \leq f$ . Furthermore,  $\check{f}(x) = \bar{f}(x) = f(x) \in \mathbb{R}$  and therefore Proposition 9.6 implies that  $\check{f} \in \Gamma_0(\mathcal{H})$ . Hence, we deduce from Proposition 13.14(iv) and Corollary 13.33 that  $f^{**} = \check{f}^{**} = \check{f} = \bar{f} \leq f$ . Thus, by Lemma 1.31(v), (i)  $\Leftrightarrow f(x) = \bar{f}(x) \Leftrightarrow$  (iii)  $\Leftrightarrow$  (ii).  $\square$

**Proposition 13.39** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ . If  $f$  has a continuous affine minorant (equivalently, if  $\text{dom } f^* \neq \emptyset$ ), then  $f^{**} = \check{f}$ ; otherwise  $f^{**} = -\infty$ .*

*Proof.* If  $f \equiv +\infty$ , then  $\check{f} = f$ , and Proposition 13.9(iv) yields  $f^{**} \equiv +\infty$  and hence  $f^{**} = \check{f}$ . Now suppose that  $f \not\equiv +\infty$ . As seen in Proposition 13.10(ii), if  $f$  possesses no continuous affine minorant, then  $f^* \equiv +\infty$  and therefore  $f^{**} \equiv -\infty$ . Otherwise, there exists a continuous affine function  $a: \mathcal{H} \rightarrow \mathbb{R}$  such that  $a \leq f$ . Hence,  $a = \check{a} \leq \check{f} \leq \bar{f}$ , and  $\check{f}$  is therefore proper. Thus, in view of Proposition 13.11, we have  $\bar{f} \in \Gamma_0(\mathcal{H})$ , and it follows from Proposition 13.14(iv) and Theorem 13.32 that  $f^{**} = (\bar{f})^{**} = \bar{f}$ .  $\square$

**Proposition 13.40** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a proper convex function that has a continuous affine minorant. Then the following hold:*

- (i)  $\text{dom } f \subset \text{dom } f^{**} \subset \overline{\text{dom } f}$ .
- (ii)  $\text{epi } f^{**} = \overline{\text{epi } f}$ .
- (iii)  $(\forall x \in \mathcal{H}) f^{**}(x) = \varliminf_{y \rightarrow x} f(y)$ .

*Proof.* Proposition 13.38 yields  $f^{**} = \check{f} = \bar{f}$ .

(i): Combine Proposition 8.2 and Proposition 9.8(iv).

(ii): Lemma 1.31(vi) yields  $\text{epi } f^{**} = \text{epi } \bar{f} = \overline{\text{epi } f}$ .

(iii): Let  $x \in \mathcal{H}$ . Then Proposition 13.38 and Lemma 1.31(iv) yield  $f^{**}(x) = \bar{f}(x) = \underline{\lim}_{y \rightarrow x} f(y)$ .  $\square$

Here is a sharpening of Proposition 13.25(ii).

**Proposition 13.41** *Let  $(f_i)_{i \in I}$  be a family of functions in  $\Gamma_0(\mathcal{H})$  such that  $\sup_{i \in I} f_i \not\equiv +\infty$ . Then  $(\sup_{i \in I} f_i)^* = (\inf_{i \in I} f_i^*)^\vee$ .*

*Proof.* Theorem 13.32 and Proposition 13.25(i) imply that  $\sup_{i \in I} f_i = \sup_{i \in I} f_i^{**} = (\inf_{i \in I} f_i^*)^*$ . Hence,  $(\sup_{i \in I} f_i)^* = (\inf_{i \in I} f_i^*)^{**}$  and the claim follows from Proposition 13.39.  $\square$

**Proposition 13.42** *Let  $\mathcal{K}$  be a real Hilbert space, let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{dom } g \cap \text{ran } L \neq \emptyset$ . Then  $(g \circ L)^* = (L^* \triangleright g^*)^{**} = (L^* \triangleright g^*)^\vee$ .*

*Proof.* Proposition 13.21(v) gives  $(g \circ L)^* \leq L^* \triangleright g^*$ . Thus, it follows from Proposition 13.14(ii), Proposition 13.21(iv), Corollary 13.33, and Proposition 13.14(i) that  $(g \circ L)^{**} \geq (L^* \triangleright g^*)^* = g^{**} \circ L^{**} = g \circ L \geq (g \circ L)^{**}$ . Consequently,  $(g \circ L)^* = (g \circ L)^{***} = (L^* \triangleright g^*)^{**}$  by Proposition 13.14(iii). On the other hand, Proposition 13.21(iv) yields  $\emptyset \neq \text{dom}(g \circ L) = \text{dom}(L^* \triangleright g^*)^*$ . Therefore, by Proposition 13.39,  $(L^* \triangleright g^*)^{**} = (L^* \triangleright g^*)^\vee$ .  $\square$

We conclude this section with a result on the conjugation of convex integral functions.

**Proposition 13.43** *Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space, let  $(\mathcal{H}, \langle \cdot | \cdot \rangle_{\mathcal{H}})$  be a real Hilbert space, and let  $\varphi \in \Gamma_0(\mathcal{H})$ . Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); \mathcal{H})$  and that one of the following holds:*

- (a)  $\mu(\Omega) < +\infty$ .
- (b)  $\varphi \geq \varphi(0) = 0$ .

Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$x \mapsto \begin{cases} \int_{\Omega} \varphi(x(\omega)) \mu(d\omega), & \text{if } \varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R}); \\ +\infty, & \text{otherwise,} \end{cases} \quad (13.25)$$

and

$$g: \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$u \mapsto \begin{cases} \int_{\Omega} \varphi^*(u(\omega)) \mu(d\omega), & \text{if } \varphi^* \circ u \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R}); \\ +\infty, & \text{otherwise.} \end{cases} \quad (13.26)$$

Then the following hold:

- (i)  $f \in \Gamma_0(\mathcal{H})$  and  $g \in \Gamma_0(\mathcal{H})$ .
- (ii) Suppose that  $\mathbf{H}$  is separable, and that  $(\Omega, \mathcal{F}, \mu)$  is complete (every subset of a set in  $\mathcal{F}$  of  $\mu$ -measure zero is in  $\mathcal{F}$ ) and  $\sigma$ -finite ( $\Omega$  is a countable union of sets in  $\mathcal{F}$  of finite  $\mu$ -measure). Then  $f^* = g$ .

*Proof.* (i): We have shown in Proposition 9.32 that  $f \in \Gamma_0(\mathcal{H})$ . Likewise, since Corollary 13.33 implies that  $\varphi^* \in \Gamma_0(\mathbf{H})$  and Proposition 13.19 implies that  $\varphi^* \geq \varphi^*(0) = 0$ ,  $g$  is a well-defined function in  $\Gamma_0(\mathcal{H})$ .

(ii): This follows from (i) and [223] (see also [224, Theorem 21(a)]).  $\square$

**Example 13.44** Let  $(\Omega, \mathcal{F}, \mathbf{P})$  be a complete probability space, let  $\mathcal{H}$  be the space of square-integrable random variables (see Example 2.8), and set  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]: X \mapsto (1/p)\mathbf{E}|X|^p$ , where  $p \in ]1, +\infty[$ . Then  $f^*: \mathcal{H} \rightarrow ]-\infty, +\infty]: X \mapsto (1/p^*)\mathbf{E}|X|^{p^*}$ , where  $p^* = p/(p-1)$ .

*Proof.* This follows from Proposition 13.43 and Example 13.2(i).  $\square$

## Exercises

**Exercise 13.1** Prove Example 13.2.

**Exercise 13.2** Prove items (i)–(iv) in Example 13.3.

**Exercise 13.3** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\alpha \in \mathbb{R}_{++}$ . Define  $\alpha \star f = \alpha f \circ \alpha^{-1}\text{Id}$ . Prove the following:

- (i)  $\text{epi}(\alpha \star f) = \alpha \text{epi } f$ .
- (ii)  $(\alpha f)^* = \alpha \star f^*$ .
- (iii)  $(\alpha \star f)^* = \alpha f^*$ .
- (iv)  $(\alpha^2 f \circ \alpha^{-1}\text{Id})^* = \alpha^2 f^* \circ \alpha^{-1}\text{Id}$ .

The operation  $(\alpha, f) \mapsto \alpha \star f$  is sometimes called *epi-multiplication*, it is the property dual to pointwise multiplication under conjugation.

**Exercise 13.4** Suppose that  $p \in ]0, 1[$  and set

$$f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} -\frac{1}{p}x^p, & \text{if } x \geq 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (13.27)$$

Show that

$$f^*: \mathbb{R} \rightarrow ]-\infty, +\infty]: u \mapsto \begin{cases} -\frac{1}{p^*}|u|^{p^*}, & \text{if } u < 0; \\ +\infty, & \text{otherwise,} \end{cases} \quad (13.28)$$

where  $\frac{1}{p} + \frac{1}{p^*} = 1$ .

**Exercise 13.5** Let  $f$  and  $g$  be functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$  and let  $\alpha \in ]0, 1[$ . Show that  $(\alpha f + (1 - \alpha)g)^* \leq \alpha f^* + (1 - \alpha)g^*$ .

**Exercise 13.6** Derive Proposition 13.21(i) from Proposition 13.28.

**Exercise 13.7** Set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \frac{1}{2} \overline{\lim} \|x - z_n\|^2$ , where  $(z_n)_{n \in \mathbb{N}}$  is a bounded sequence in  $\mathcal{H}$  (see also Example 8.17 and Proposition 11.17) and define  $g: \mathcal{H} \rightarrow \mathbb{R}: u \mapsto \frac{1}{2} \|u\|^2 + \underline{\lim} \langle z_n | u \rangle$ . Show that  $f^* \leq g$ , and provide an example of a sequence  $(z_n)_{n \in \mathbb{N}}$  for which  $f^* = g$  and another one for which  $f^* < g$ .

**Exercise 13.8** Prove Proposition 6.26 via Proposition 13.21(i).

**Exercise 13.9** Prove Proposition 13.30.

**Exercise 13.10** Let  $C$  and  $D$  be nonempty subsets of  $\mathcal{H}$  such that  $D$  is closed and convex. Show that  $C \subset D \Leftrightarrow \sigma_C \leq \sigma_D$ .



# Chapter 14

## Further Conjugation Results

In this chapter, we exhibit several deeper results on conjugation. We first discuss Moreau's decomposition principle, whereby a vector is decomposed in terms of the proximity operator of a lower semicontinuous function and that of its conjugate. This powerful nonlinear principle extends the standard linear decomposition with respect to a closed linear subspace and its orthogonal complement. Basic results concerning the proximal average and positively homogeneous functions are also presented. Also discussed are the Moreau–Rockafellar theorem, which characterizes coercivity in terms of an interiority condition, and the Toland–Singer theorem, which provides an appealing formula for the conjugate of the difference.

### 14.1 Moreau's Decomposition

In this section, we take a closer look at the infimal convolution of a convex function in  $\Gamma_0(\mathcal{H})$  and the function  $q = (1/2)\|\cdot\|^2$ .

**Proposition 14.1** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\gamma \in \mathbb{R}_{++}$ , and set  $q = (1/2)\|\cdot\|^2$ . Then*

$$(f + \gamma q)^* = f^* \square (\gamma^{-1} q) = \gamma(f^*). \quad (14.1)$$

*Proof.* It follows from Corollary 13.33, Proposition 13.16, and Proposition 13.21(i) that  $f + \gamma q = f^{**} + (\gamma^{-1} q)^* = (f^* \square (\gamma^{-1} q))^*$ . Since Corollary 13.33 and Proposition 12.15 imply that  $f^* \square (\gamma^{-1} q) = f^* \square (\gamma^{-1} q) \in \Gamma_0(\mathcal{H})$ , we deduce from Theorem 13.32 and (12.20) that  $(f + \gamma q)^* = (f^* \square (\gamma^{-1} q))^{**} = f^* \square (\gamma^{-1} q) = \gamma(f^*)$ .  $\square$

The next proposition characterizes functions with strongly convex conjugates.

**Proposition 14.2** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\gamma \in \mathbb{R}_{++}$ , and set  $q = (1/2)\|\cdot\|^2$ . Then the following are equivalent:*

- (i)  $f^* - \gamma^{-1}q$  is convex, i.e.,  $f^*$  is  $\gamma^{-1}$ -strongly convex.
- (ii)  $\gamma q - f$  is convex.

*Proof.* (i) $\Rightarrow$ (ii): Set  $h = f^* - \gamma^{-1}q$ . Since  $f \in \Gamma_0(\mathcal{H})$  and  $h$  is convex, we have  $h \in \Gamma_0(\mathcal{H})$  and, by Corollary 13.33,  $h^* \in \Gamma_0(\mathcal{H})$ . Hence, using Theorem 13.32 and Example 13.4, we obtain

$$f = f^{**} = (h + \gamma^{-1}q)^* = \gamma q - \gamma h \circ \gamma \text{Id}. \quad (14.2)$$

Hence, since it follows from Proposition 12.15 that  $\gamma h$  is convex, we deduce that  $\gamma q - f = \gamma h \circ \gamma \text{Id}$  is convex.

(ii) $\Rightarrow$ (i): Set  $g = \gamma q - f$ . Then  $g \in \Gamma_0(\mathcal{H})$  and therefore Corollary 13.33 yields  $g = g^{**}$ . Thus,  $f = \gamma q - g = \gamma q - (g^*)^*$ . In turn, Proposition 13.26 yields  $f^* = ((\gamma g^* - q)^* + q)/\gamma$ . Thus, Proposition 13.11 implies that  $f^* - q/\gamma = (\gamma g^* - q)^*/\gamma$  is convex.  $\square$

**Theorem 14.3** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\gamma \in \mathbb{R}_{++}$ . Then the following hold:*

- (i) Set  $q = (1/2)\|\cdot\|^2$ . Then

$$\gamma^{-1}q = f \square (\gamma^{-1}q) + (f^* \square (\gamma q)) \circ \gamma^{-1}\text{Id} = \gamma f + {}^{1/\gamma}(f^*) \circ \gamma^{-1}\text{Id}. \quad (14.3)$$

- (ii)  $\text{Id} = \text{Prox}_{\gamma f} + \gamma \text{Prox}_{f^*/\gamma} \circ \gamma^{-1}\text{Id}$ .

- (iii) Let  $x \in \mathcal{H}$ . Then

$$\begin{aligned} f(\text{Prox}_{\gamma f} x) + f^*(\text{Prox}_{f^*/\gamma}(x/\gamma)) \\ = \langle \text{Prox}_{\gamma f} x \mid \text{Prox}_{f^*/\gamma}(x/\gamma) \rangle. \end{aligned} \quad (14.4)$$

*Proof.* (i): It follows from Example 13.4, Proposition 12.15, and Proposition 14.1 that

$$\begin{aligned} \gamma^{-1}q &= (f^* + \gamma q)^* + (f^* \square (\gamma q)) \circ \gamma^{-1}\text{Id} \\ &= f \square (\gamma^{-1}q) + (f^* \square (\gamma q)) \circ \gamma^{-1}\text{Id}. \end{aligned} \quad (14.5)$$

- (ii): Take the Fréchet derivative in (i) using Proposition 12.29.

(iii): Set  $p = \text{Prox}_{\gamma f} x$ ,  $p^* = \text{Prox}_{f^*/\gamma}(x/\gamma)$ , and let  $y \in \mathcal{H}$ . Then (ii) and Proposition 12.26 yield  $\langle y - p \mid p^* \rangle + f(p) \leq f(y)$ . Therefore, it follows from Proposition 13.13 that  $f(p) + f^*(p^*) = f(p) + \sup_{y \in \mathcal{H}} (\langle y \mid p^* \rangle - f(y)) \leq \langle p \mid p^* \rangle \leq f(p) + f^*(p^*)$ . Hence,  $f(p) + f^*(p^*) = \langle p \mid p^* \rangle$ .  $\square$

The striking symmetry obtained when  $\gamma = 1$  in Theorem 14.3 deserves to be noted.

**Remark 14.4** Let  $f \in \Gamma_0(\mathcal{H})$  and set  $q = (1/2)\|\cdot\|^2$ . Then Theorem 14.3 yields

$$(f \square q) + (f^* \square q) = q \quad \text{and} \quad \text{Prox}_f + \text{Prox}_{f^*} = \text{Id}. \quad (14.6)$$

Thus, using Proposition 12.29, we obtain



$$\text{Prox}_f = \text{Id} - \nabla(f \boxminus q) = \nabla(f^* \boxminus q). \quad (14.7)$$

If  $f = \iota_K$  in (14.6), where  $K$  is a nonempty closed convex cone in  $\mathcal{H}$ , then we recover Moreau's conical decomposition (Theorem 6.29). In particular, if  $K$  is a closed linear subspace, we recover the identities  $d_K^2 + d_{K^\perp}^2 = \|\cdot\|^2$  and  $P_K + P_{K^\perp} = \text{Id}$  already established in Corollary 3.22.

**Example 14.5** Set  $f = \|\cdot\|$  and let  $x \in \mathcal{H}$ . Then Example 13.3(v) yields  $f^* = \iota_{B(0;1)}$ . We therefore derive from (14.6), Example 12.25, and (3.9) that

$$\text{Prox}_f x = \begin{cases} (1 - 1/\|x\|)x, & \text{if } \|x\| > 1; \\ 0, & \text{if } \|x\| \leq 1. \end{cases} \quad (14.8)$$

In other words,  $\text{Prox}_f$  is the soft thresholder of Example 4.9.

## 14.2 Proximal Average

**Definition 14.6** Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . The *proximal average* of  $f$  and  $g$  is

$$\begin{aligned} \text{pav}(f, g) : \mathcal{H} &\rightarrow ]-\infty, +\infty] \\ x &\mapsto \frac{1}{2} \inf_{\substack{(y,z) \in \mathcal{H} \times \mathcal{H} \\ y+z=2x}} \left( f(y) + g(z) + \frac{1}{4} \|y - z\|^2 \right). \end{aligned} \quad (14.9)$$

**Proposition 14.7** Set  $L : \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} : (y, z) \mapsto (y + z)/2$ , let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and set  $F : \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty] : (y, z) \mapsto \frac{1}{2}f(y) + \frac{1}{2}g(z) + \frac{1}{8}\|y - z\|^2$ . Then the following hold:

- (i)  $\text{pav}(f, g) = \text{pav}(g, f)$ .
- (ii)  $\text{pav}(f, g) = L \triangleright F$ .
- (iii)  $\text{dom } \text{pav}(f, g) = \frac{1}{2} \text{dom } f + \frac{1}{2} \text{dom } g$ .
- (iv)  $\text{pav}(f, g)$  is a proper convex function.

*Proof.* (i): Clear from Definition 14.6.

(ii): Definition 12.33 and Definition 14.6 imply that  $\text{pav}(f, g) = L \triangleright F$ . Take  $x \in \text{dom } \text{pav}(f, g)$  and set  $h : y \mapsto \|y\|^2 + (f(y) + g(2x - y) - 2\langle y | x \rangle + \|x\|^2)$ . It suffices to show that  $h$  has a minimizer. Since  $h$  is a strongly convex function in  $\Gamma_0(\mathcal{H})$ , this follows from Corollary 11.16, which asserts that  $h$  has a unique minimizer.

(iii): This follows from (ii) and Proposition 12.34(i).

(iv): Since  $F$  is convex and  $L$  is linear, the convexity of  $\text{pav}(f, g)$  follows from (ii) and Proposition 12.34(ii). On the other hand, properness follows from (iii).  $\square$

**Corollary 14.8** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and set  $q = (1/2)\|\cdot\|^2$ . Then the following hold:*

- (i)  $\text{pav}(f, g) \in \Gamma_0(\mathcal{H})$ .
- (ii)  $(\text{pav}(f, g))^* = \text{pav}(f^*, g^*)$ .
- (iii)  $\text{pav}(f, g) \square q = \frac{1}{2}(f \square q) + \frac{1}{2}(g \square q)$ .
- (iv)  $\text{Prox}_{\text{pav}(f, g)} = \frac{1}{2}\text{Prox}_f + \frac{1}{2}\text{Prox}_g$ .

*Proof.* We define an operator  $\Theta$  on  $\Gamma_0(\mathcal{H}) \times \Gamma_0(\mathcal{H})$  by

$$\Theta: (f_1, f_2) \mapsto \left(\frac{1}{2}(f_1 + q) \circ (2\text{Id})\right) \square \left(\frac{1}{2}(f_2 + q) \circ (2\text{Id})\right). \quad (14.10)$$

Definition 14.6 and Lemma 2.11(ii) yield, for every  $f_1$  and  $f_2$  in  $\Gamma_0(\mathcal{H})$ , and for every  $x \in \mathcal{H}$ ,

$$\text{pav}(f_1, f_2)(x) = \inf_{\substack{(y, z) \in \mathcal{H} \times \mathcal{H} \\ y+z=x}} \left( \frac{1}{2}f_1(2y) + \frac{1}{2}f_2(2z) + 2q(y) + 2q(z) \right) - q(x). \quad (14.11)$$

Hence

$$(\forall f_1 \in \Gamma_0(\mathcal{H}))(\forall f_2 \in \Gamma_0(\mathcal{H})) \quad \text{pav}(f_1, f_2) = \Theta(f_1, f_2) - q. \quad (14.12)$$

Proposition 13.21(i), Proposition 13.20(ii), and Proposition 14.1 yield

$$\begin{aligned} (\Theta(f, g))^* &= \left(\frac{1}{2}(f + q) \circ (2\text{Id})\right)^* + \left(\frac{1}{2}(g + q) \circ (2\text{Id})\right)^* \\ &= \frac{1}{2}(f + q)^* + \frac{1}{2}(g + q)^* \\ &= \frac{1}{2}(f^* \square q) + \frac{1}{2}(g^* \square q). \end{aligned} \quad (14.13)$$

In view of (14.12), Proposition 13.26, (14.13), (14.6), Proposition 14.1, Proposition 13.20(i), Proposition 13.21(i), and (14.10), we get

$$\begin{aligned} (\text{pav}(f, g))^* &= (\Theta(f, g) - q)^* \\ &= (q - (\Theta(f, g))^*)^* - q \\ &= \left(\frac{1}{2}(q - (f^* \square q)) + \frac{1}{2}(q - (g^* \square q))\right)^* - q \\ &= \left(\frac{1}{2}(f \square q) + \frac{1}{2}(g \square q)\right)^* - q \\ &= \left(\frac{1}{2}(f \square q)\right)^* \square \left(\frac{1}{2}(g \square q)\right)^* - q \\ &= \left(\frac{1}{2}(f^* + q) \circ (2\text{Id})\right) \square \left(\frac{1}{2}(g^* + q) \circ (2\text{Id})\right) - q \\ &= \Theta(f^*, g^*) - q. \end{aligned} \quad (14.14)$$

On the other hand,  $\text{pav}(f^*, g^*) = \Theta(f^*, g^*) - q$  by (14.12). Combining this with (14.14), we obtain (ii). In turn,  $(\text{pav}(f, g))^{**} = (\text{pav}(f^*, g^*))^* =$

$\text{pav}(f^{**}, g^{**}) = \text{pav}(f, g)$ , which implies (i) by virtue of Proposition 14.7(iv) and Proposition 13.11. Furthermore, using (ii), Proposition 14.1, (14.12), and (14.13), we deduce that

$$\begin{aligned} \text{pav}(f^*, g^*) \square q &= (\text{pav}(f, g))^* \square q \\ &= (\text{pav}(f, g) + q)^* \\ &= (\Theta(f, g))^* \\ &= \frac{1}{2}(f^* \square q) + \frac{1}{2}(g^* \square q). \end{aligned} \quad (14.15)$$

Hence, upon replacing in (14.15)  $f$  by  $f^*$  and  $g$  by  $g^*$ , we obtain (iii). Finally, (14.15) and (14.7) yield

$$\text{Prox}_{\text{pav}(f, g)} = \nabla((\text{pav}(f, g))^* \square q) = \nabla\left(\frac{1}{2}(f^* \square q) + \frac{1}{2}(g^* \square q)\right), \quad (14.16)$$

which implies (iv).  $\square$

The proofs of the following results are left as Exercise 14.3 and Exercise 14.4.

**Proposition 14.9** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Then*

$$\left(\frac{1}{2}f^* + \frac{1}{2}g^*\right)^* \leq \text{pav}(f, g) \leq \frac{1}{2}f + \frac{1}{2}g. \quad (14.17)$$

**Proposition 14.10** *Let  $F$  and  $G$  be in  $\Gamma_0(\mathcal{H} \times \mathcal{H})$ . Then  $(\text{pav}(F, G))^\top = \text{pav}(F^\top, G^\top)$ .*

## 14.3 Positively Homogeneous Functions

**Proposition 14.11** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and set*

$$C = \{u \in \mathcal{H} \mid (\forall x \in \mathcal{H}) \quad \langle x \mid u \rangle \leq f(x)\}. \quad (14.18)$$

*Then the following are equivalent:*

- (i)  $f$  is positively homogeneous and  $f \in \Gamma_0(\mathcal{H})$ .
- (ii)  $f = \sigma_C$  and  $C$  is nonempty, closed, and convex.
- (iii)  $f$  is the support function of a nonempty closed convex subset of  $\mathcal{H}$ .

*Proof.* (i) $\Rightarrow$ (ii): Let  $u \in \mathcal{H}$ . We deduce from Proposition 9.14 that, given  $y \in \text{dom } f$ ,  $f(0) = \lim_{\alpha \downarrow 0} f((1 - \alpha)0 + \alpha y) = \lim_{\alpha \downarrow 0} \alpha f(y) = 0$ . Thus, if  $u \in C$ , we obtain  $f^*(u) = \sup_{x \in \mathcal{H}} \langle x \mid u \rangle - f(x) \leq 0 = \langle 0 \mid u \rangle - f(0) \leq f^*(u)$  and, in turn,  $f^*(u) = 0$ . On the other hand, if  $u \notin C$ , then there exists  $z \in \mathcal{H}$  such that  $\langle z \mid u \rangle - f(z) > 0$ . Consequently,  $(\forall \lambda \in \mathbb{R}_{++}) \quad f^*(u) \geq \langle \lambda z \mid u \rangle - f(\lambda z) = \lambda(\langle z \mid u \rangle - f(z))$ . Since  $\lambda$  can be arbitrarily large, we

conclude that  $f^*(u) = +\infty$ . Altogether,  $f^* = \iota_C$ . However, since  $f \in \Gamma_0(\mathcal{H})$ , Corollary 13.33 yields  $\iota_C = f^* \in \Gamma_0(\mathcal{H})$ , which shows that  $C$  is a nonempty closed convex set. On the other hand, we deduce from Theorem 13.32 that  $f = f^{**} = \iota_C^* = \sigma_C$ .

(ii) $\Rightarrow$ (iii): Clear.

(iii) $\Rightarrow$ (i): Example 11.2.  $\square$

The next result establishes a connection between the polar set and the Minkowski gauge.

**Proposition 14.12** *Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $0 \in C$ . Then  $m_C^* = \iota_{C^\circ}$ .*

*Proof.* Fix  $u \in \mathcal{H}$ . First, suppose that  $u \in C^\circ$ , let  $x \in \text{dom } m_C$ , and let  $\lambda \in ]m_C(x), +\infty[$ . Then, by (8.25), there exists  $y \in C$  such that  $x = \lambda y$  and, in turn, such that  $\langle x | u \rangle = \lambda \langle y | u \rangle \leq \lambda$ . Taking the limit as  $\lambda \downarrow m_C(x)$  yields  $\langle x | u \rangle \leq m_C(x)$ , and we deduce from Proposition 13.9(iv) that  $m_C^*(u) \leq 0$ . On the other hand,  $m_C^*(u) \geq \langle 0 | u \rangle - m_C(0) = 0$ . Altogether,  $m_C^*$  and  $\iota_{C^\circ}$  coincide on  $C^\circ$ . Now, suppose that  $u \notin C^\circ$ . Then there exists  $x \in C$  such that  $\langle x | u \rangle > 1 \geq m_C(x)$  and, using (8.24), we deduce that  $m_C^*(u) \geq \sup_{\lambda \in \mathbb{R}_{++}} \langle \lambda x | u \rangle - m_C(\lambda x) = \sup_{\lambda \in \mathbb{R}_{++}} \lambda(\langle x | u \rangle - m_C(x)) = +\infty$ . Therefore,  $m_C^*$  and  $\iota_{C^\circ}$  coincide also on  $\mathcal{H} \setminus C^\circ$ .  $\square$

**Corollary 14.13** *Let  $C$  be a closed convex subset of  $\mathcal{H}$  such that  $0 \in C$ . Then the following hold:*

(i)  $C = \text{lev}_{\leq 1} m_C$ .

(ii) Suppose that  $\text{int } C \neq \emptyset$ . Then  $\text{int } C = \text{lev}_{< 1} m_C$ .

*Proof.* (i): It is clear that  $C \subset \text{lev}_{\leq 1} m_C$ . Now assume the existence of a vector  $x \in (\text{lev}_{\leq 1} m_C) \setminus C$ . Theorem 3.38 provides  $u \in \mathcal{H} \setminus \{0\}$  such that  $\langle x | u \rangle > \sigma_C(u) \geq 0$  since  $0 \in C$ . Hence, after scaling  $u$  if necessary, we assume that  $\langle x | u \rangle > 1 \geq \sigma_C(u)$  so that  $u \in C^\circ$ . Using Proposition 14.12 and Proposition 13.14(i), we obtain the contradiction  $1 < \langle u | x \rangle \leq \sigma_{C^\circ}(x) = \iota_{C^\circ}^*(x) = m_C^{**}(x) \leq m_C(x) \leq 1$ .

(ii): Example 8.34 states that  $m_C$  is continuous on  $\mathcal{H}$ . In view of (i) and Corollary 8.38(ii) applied to  $m_C - 1$ , we deduce that  $\text{int } C = \text{int}(\text{lev}_{\leq 1} m_C) = \text{lev}_{< 1} m_C$ .  $\square$

## 14.4 Coercivity

We investigate some aspects of the interplay between coercivity and conjugation.

**Proposition 14.14** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\alpha \in \mathbb{R}_{++}$ , and consider the following properties:*

- (i)  $\underline{\lim}_{\|x\| \rightarrow +\infty} f(x)/\|x\| > \alpha$ .
- (ii)  $(\exists \beta \in \mathbb{R}) \ f \geq \alpha\|\cdot\| + \beta$ .
- (iii)  $(\exists \gamma \in \mathbb{R}) \ f^*|_{B(0;\alpha)} \leq \gamma$ .
- (iv)  $\underline{\lim}_{\|x\| \rightarrow +\infty} f(x)/\|x\| \geq \alpha$ .

Then (i) $\Rightarrow$ (ii) $\Leftrightarrow$ (iii) $\Rightarrow$ (iv).

*Proof.* (i) $\Rightarrow$ (ii): There exists  $\rho \in \mathbb{R}_{++}$  such that

$$(\forall x \in \mathcal{H} \setminus B(0; \rho)) \quad f(x) \geq \alpha\|x\|. \quad (14.19)$$

Now set  $\mu = \inf f(B(0; \rho))$ . Then we deduce from Proposition 13.10(iii) that  $\mu > -\infty$ . Thus  $(\forall x \in B(0; \rho)) \ \alpha\|x\| \leq \alpha\rho \leq (\alpha\rho - \mu) + f(x)$ . Hence,

$$(\forall x \in B(0; \rho)) \quad f(x) \geq \alpha\|x\| + (\mu - \alpha\rho). \quad (14.20)$$

Altogether, (ii) holds with  $\beta = \min\{0, \mu - \alpha\rho\}$ .

(ii) $\Leftrightarrow$ (iii): Corollary 13.34 and Example 13.3(v) yield  $\alpha\|\cdot\| + \beta \leq f \Leftrightarrow f^* \leq (\alpha\|\cdot\| + \beta)^* \Leftrightarrow f^* \leq \iota_{B(0;\alpha)} - \beta$ .

(ii) $\Rightarrow$ (iv):  $\underline{\lim}_{\|x\| \rightarrow +\infty} f(x)/\|x\| \geq \underline{\lim}_{\|x\| \rightarrow +\infty} (\alpha + \beta/\|x\|) = \alpha$ .  $\square$

Proposition 14.14 yields at once the following result.

**Proposition 14.15** *Let  $f \in \Gamma_0(\mathcal{H})$  and consider the following properties:*

- (i)  $f$  is supercoercive.
- (ii)  $f^*$  is bounded on every bounded subset of  $\mathcal{H}$ .
- (iii)  $\text{dom } f^* = \mathcal{H}$ .

Then (i) $\Leftrightarrow$ (ii) $\Rightarrow$ (iii).

**Proposition 14.16** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then the following are equivalent:*

- (i)  $f$  is coercive.
- (ii) The level sets  $(\text{lev}_{\leq \xi} f)_{\xi \in \mathbb{R}}$  are bounded.
- (iii)  $\underline{\lim}_{\|x\| \rightarrow +\infty} f(x)/\|x\| > 0$ .
- (iv)  $(\exists (\alpha, \beta) \in \mathbb{R}_{++} \times \mathbb{R}) \ f \geq \alpha\|\cdot\| + \beta$ .
- (v)  $f^*$  is bounded above on a neighborhood of 0.
- (vi)  $0 \in \text{int dom } f^*$ .

*Proof.* (i) $\Leftrightarrow$ (ii): Proposition 11.11.

(ii) $\Rightarrow$ (iii): Suppose that  $\underline{\lim}_{\|x\| \rightarrow +\infty} f(x)/\|x\| \leq 0$  and set  $(\forall n \in \mathbb{N}) \ \alpha_n = n + 1$ . Then for every  $n \in \mathbb{N}$ , there exists  $x_n \in \mathcal{H}$  such that  $\|x_n\| \geq \alpha_n^2$  and  $f(x_n)/\|x_n\| \leq 1/\alpha_n$ . We thus obtain a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\text{dom } f$  such that  $0 < \alpha_n/\|x_n\| \leq 1/\alpha_n \rightarrow 0$  and  $f(x_n)/\|x_n\| \leq 1/\alpha_n$ . Now fix  $z \in \text{dom } f$  and set

$$(\forall n \in \mathbb{N}) \quad y_n = \left(1 - \frac{\alpha_n}{\|x_n\|}\right) z + \frac{\alpha_n}{\|x_n\|} x_n. \quad (14.21)$$

The convexity of  $f$  implies that  $\sup_{n \in \mathbb{N}} f(y_n) \leq |f(z)| + 1$ . Therefore,  $(y_n)_{n \in \mathbb{N}}$  lies in  $\text{lev}_{\leq |f(z)|+1} f$  and it is therefore bounded. On the other hand, since  $\|y_n\| \geq \alpha_n - \|z\| \rightarrow +\infty$ , we reach a contradiction.

(iii)  $\Rightarrow$  (i): Clear.

(iii)  $\Leftrightarrow$  (iv)  $\Leftrightarrow$  (v)  $\Leftrightarrow$  (vi): Proposition 14.14 and Theorem 8.29.  $\square$

**Theorem 14.17 (Moreau–Rockafellar)** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $u \in \mathcal{H}$ . Then  $f - \langle \cdot | u \rangle$  is coercive if and only if  $u \in \text{int dom } f^*$ .*

*Proof.* Using Proposition 14.16 and Proposition 13.20(iii), we obtain the equivalences  $f - \langle \cdot | u \rangle$  is coercive  $\Leftrightarrow 0 \in \text{int dom}(f - \langle \cdot | u \rangle)^* \Leftrightarrow 0 \in \text{int dom}(\tau_{-u} f^*) \Leftrightarrow u \in \text{int dom } f^*$ .  $\square$

**Corollary 14.18** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and suppose that  $f$  is supercoercive. Then the following hold:*

(i)  $f \square g$  is coercive if and only if  $g$  is coercive.

(ii)  $f \square g$  is supercoercive if and only if  $g$  is supercoercive.

*Proof.* Proposition 14.15 asserts that  $f^*$  is bounded on every bounded subset of  $\mathcal{H}$  and that  $\text{dom } f^* = \mathcal{H}$ . Furthermore,  $f \square g = f \square g \in \Gamma_0(\mathcal{H})$  by Proposition 12.14(i).

(i): Using Proposition 14.16 and Proposition 13.21(i), we obtain the equivalences  $f \square g$  is coercive  $\Leftrightarrow 0 \in \text{int dom}(f \square g)^* \Leftrightarrow 0 \in \text{int dom}(f^* + g^*) \Leftrightarrow 0 \in \text{int}(\text{dom } f^* \cap \text{dom } g^*) \Leftrightarrow 0 \in \text{int dom } g^* \Leftrightarrow g$  is coercive.

(ii): Using Proposition 14.15 and Proposition 13.21(i), we obtain the equivalences  $f \square g$  is supercoercive  $\Leftrightarrow (f \square g)^*$  is bounded on bounded sets  $\Leftrightarrow f^* + g^*$  is bounded on bounded sets  $\Leftrightarrow g^*$  is bounded on bounded sets  $\Leftrightarrow g$  is supercoercive.  $\square$

## 14.5 The Conjugate of the Difference

**Proposition 14.19** *Let  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $h \in \Gamma_0(\mathcal{H})$ , and set*

$$f: \mathcal{H} \rightarrow [-\infty, +\infty] : x \mapsto \begin{cases} g(x) - h(x), & \text{if } x \in \text{dom } g; \\ +\infty, & \text{if } x \notin \text{dom } g. \end{cases} \quad (14.22)$$

*Then*

$$(\forall u \in \mathcal{H}) \quad f^*(u) = \sup_{v \in \text{dom } h^*} (g^*(u + v) - h^*(v)). \quad (14.23)$$

*Proof.* Fix  $u \in \mathcal{H}$ . Using Corollary 13.33, we obtain

$$f^*(u) = \sup_{x \in \text{dom } g} (\langle x | u \rangle - g(x) + h(x))$$

$$\begin{aligned}
&= \sup_{x \in \text{dom } g} (\langle x \mid u \rangle - g(x) + h^{**}(x)) \\
&= \sup_{x \in \text{dom } g} \left( \langle x \mid u \rangle - g(x) + \sup_{v \in \text{dom } h^*} (\langle v \mid x \rangle - h^*(v)) \right) \\
&= \sup_{v \in \text{dom } h^*} \left( \sup_{x \in \text{dom } g} (\langle x \mid u + v \rangle - g(x)) - h^*(v) \right) \\
&= \sup_{v \in \text{dom } h^*} (g^*(u + v) - h^*(v)), \tag{14.24}
\end{aligned}$$

as required.  $\square$

**Corollary 14.20 (Toland–Singer)** *Let  $g$  and  $h$  be in  $\Gamma_0(\mathcal{H})$ . Then*

$$\inf_{x \in \text{dom } g} (g(x) - h(x)) = \inf_{v \in \text{dom } h^*} (h^*(v) - g^*(v)). \tag{14.25}$$

*Proof.* Set  $u = 0$  in (14.23).  $\square$

## Exercises

**Exercise 14.1** Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $\text{pav}(f, f^*) = (1/2)\|\cdot\|^2$ .

**Exercise 14.2** Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Show that

$$\text{pav}(f, g) = -^1 \left( -\frac{^1f + ^1g}{2} \right). \tag{14.26}$$

**Exercise 14.3** Prove Proposition 14.9.

**Exercise 14.4** Prove Proposition 14.10.

**Exercise 14.5** Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ . Show that  $\text{pav}(F, F^*\mathfrak{T})$  is autoconjugate.

**Exercise 14.6** Let  $\alpha \in \mathbb{R}_{++}$  and  $\beta \in \mathbb{R}_{++}$ . Set  $q = (1/2)\|\cdot\|^2$ ,  $f = \alpha q$ , and  $g = \beta q$ . Show that (14.17) becomes

$$\frac{2\alpha\beta}{\alpha + \beta} q \leq \frac{\alpha + \beta + 2\alpha\beta}{2 + \alpha + \beta} q \leq \frac{\alpha + \beta}{2} q, \tag{14.27}$$

which illustrates that the coefficient of  $q$  in the middle term (which corresponds to the proximal average) is bounded below by the harmonic mean of  $\alpha$  and  $\beta$ , and bounded above by their arithmetic mean.

**Exercise 14.7** Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$  and  $0 \in \text{sri}(\text{dom } f^* - \text{dom } g^*)$ , and let  $\alpha \in \mathbb{R}_{++}$  be such that  $\text{dom } f \cap \alpha \text{dom } f \neq \emptyset$  and  $\text{dom } f^* \cap \alpha \text{dom } f^* \neq \emptyset$ . Set

$$f \star g = \text{pav}(f + g, f \square g) \quad \text{and} \quad \alpha \star f = \text{pav}(\alpha f, \alpha f(\cdot/\alpha)), \quad (14.28)$$

and show that

$$(f \star g)^* = f^* \star g^* \quad \text{and} \quad (\alpha \star f)^* = \alpha \star f^*. \quad (14.29)$$

**Exercise 14.8** Let  $C$  be a closed convex subset of  $\mathcal{H}$  such that  $0 \in C$ . Show that  $C$  is bounded if and only if  $0 \in \text{int } C^\odot$ .

**Exercise 14.9** Let  $h: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and suppose, for every  $g \in \Gamma_0(\mathcal{H})$ , that  $f$  is as in (14.22) and that (14.23) holds. Let  $h: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and suppose that, for every  $g \in \Gamma_0(\mathcal{H})$ , (14.22)–(14.23) hold. Show that  $h \in \Gamma_0(\mathcal{H})$ .

**Exercise 14.10** Let  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, and set  $q = (1/2)\|\cdot\|^2$ . Then it follows from Proposition 13.26 that  $(g - q)^* = (q - g^*)^* - q$ . Prove this result using Proposition 14.19.



# Chapter 15

## Fenchel–Rockafellar Duality

Of central importance in convex analysis are conditions guaranteeing that the conjugate of a sum is the infimal convolution of the conjugates. The main result in this direction is a theorem due to Attouch and Brézis. In turn, it gives rise to the Fenchel–Rockafellar duality framework for convex optimization problems. The applications we discuss include von Neumann’s minimax theorem as well as several results on the closure of the sum of linear subspaces.

Throughout this chapter,  $\mathcal{K}$  is a real Hilbert space.

### 15.1 The Attouch–Brézis Theorem

**Proposition 15.1** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $\text{dom } f \cap \text{dom } g \neq \emptyset$ . Then  $f^* \square g^*$  is proper and convex, and it possesses a continuous affine minorant. Moreover,*

$$(f + g)^* = (f^* \square g^*)^{**} = (f^* \square g^*)^\circ. \quad (15.1)$$

*Proof.* Proposition 13.21(i) and Corollary 13.33 yield  $(f^* \square g^*)^* = f^{**} + g^{**} = f + g \in \Gamma_0(\mathcal{H})$ . In turn, Proposition 13.9(iii), Proposition 12.11, and Proposition 13.10(ii) imply that  $f^* \square g^*$  is proper and convex, and that it possesses a continuous affine minorant. Therefore, invoking Proposition 13.39, we deduce that  $(f + g)^* = (f^* \square g^*)^{**} = (f^* \square g^*)^\circ$ .  $\square$

**Proposition 15.2** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $0 \in \text{core}(\text{dom } f - \text{dom } g)$ . Then  $(f + g)^* = f^* \square g^*$ .*

*Proof.* We work in the Hilbert direct sum  $\mathcal{H} \oplus \mathcal{H}$ . For  $\eta \in \mathbb{R}$  and  $\rho \in \mathbb{R}_+$ , set

$$S_{\eta, \rho} = \{(u, v) \in \mathcal{H} \times \mathcal{H} \mid f^*(u) + g^*(v) \leq \eta \text{ and } \|u + v\| \leq \rho\}. \quad (15.2)$$

Now take  $a$  and  $b$  in  $\mathcal{H}$ . Since  $\text{cone}(\text{dom } f - \text{dom } g) = \mathcal{H}$ , there exist  $x \in \text{dom } f$ ,  $y \in \text{dom } g$ , and  $\gamma \in \mathbb{R}_{++}$  such that  $a - b = \gamma(x - y)$ . Now set  $\beta_{a,b} = \rho\|b - \gamma y\| + \gamma(f(x) + g(y) + \eta)$  and assume that  $(u, v) \in S_{\eta,\rho}$ . Then, by Cauchy–Schwarz and Fenchel–Young (Proposition 13.13),

$$\begin{aligned} \langle (a, b) \mid (u, v) \rangle &= \langle a \mid u \rangle + \langle b \mid v \rangle \\ &= \langle b - \gamma y \mid u + v \rangle + \gamma(\langle x \mid u \rangle + \langle y \mid v \rangle) \\ &\leq \|b - \gamma y\| \|u + v\| + \gamma(f(x) + f^*(u) + g(y) + g^*(v)) \\ &\leq \beta_{a,b}. \end{aligned} \tag{15.3}$$

Thus,

$$(\forall (a, b) \in \mathcal{H} \times \mathcal{H}) \quad \sup_{(u,v) \in S_{\eta,\rho}} |\langle (a, b) \mid (u, v) \rangle| \leq \max\{\beta_{a,b}, \beta_{-a,-b}\} < +\infty. \tag{15.4}$$

It follows from Lemma 2.16 applied to the linear functionals  $(a, b) \mapsto \langle (a, b) \mid (u, v) \rangle$  that  $\sup_{(u,v) \in S_{\eta,\rho}} \|(u, v)\| < +\infty$ . Hence,  $S_{\eta,\rho}$  is bounded. On the other hand,  $S_{\eta,\rho}$  is closed and convex. Altogether, Theorem 3.33 implies that  $S_{\eta,\rho}$  is weakly compact. Since  $+: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}$  is weakly continuous, Lemma 1.20 asserts that it maps  $S_{\eta,\rho}$  to the weakly compact set

$$W_{\eta,\rho} = \{u + v \in \mathcal{H} \mid (u, v) \in S_{\eta,\rho}, f^*(u) + g^*(v) \leq \eta, \text{ and } \|u + v\| \leq \rho\}. \tag{15.5}$$

Hence  $W_{\eta,\rho}$  is closed and, by Lemma 1.39, so is

$$W_\eta = \bigcup_{\rho \geq 0} W_{\eta,\rho} = \{u + v \in \mathcal{H} \mid (u, v) \in \mathcal{H} \times \mathcal{H}, f^*(u) + g^*(v) \leq \eta\}. \tag{15.6}$$

Thus, for every  $\nu \in \mathbb{R}$ ,

$$\begin{aligned} \text{lev}_{\leq \nu}(f^* \square g^*) &= \left\{ w \in \mathcal{H} \mid \inf_{u \in \mathcal{H}} f^*(u) + g^*(w - u) \leq \nu \right\} \\ &= \bigcap_{\eta > \nu} \{w \in \mathcal{H} \mid (\exists u \in \mathcal{H}) f^*(u) + g^*(w - u) \leq \eta\} \\ &= \bigcap_{\eta > \nu} W_\eta \end{aligned} \tag{15.7}$$

is closed and we deduce from Lemma 1.24 that  $f^* \square g^*$  is lower semicontinuous. On the other hand, by Proposition 15.1,  $f^* \square g^*$  is proper and convex. Altogether,  $f^* \square g^* \in \Gamma_0(\mathcal{H})$ . Therefore, Corollary 13.33 and (15.1) imply that

$$f^* \square g^* = (f^* \square g^*)^{**} = (f + g)^*. \tag{15.8}$$

It remains to show that  $f^* \square g^*$  is exact. Fix  $w \in \mathcal{H}$ . If  $w \notin \text{dom}(f^* \square g^*)$ , then  $f^* \square g^*$  is exact at  $w$ . Now suppose that  $w \in \text{dom}(f^* \square g^*)$ . Set  $F: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (u, v) \mapsto f^*(u) + g^*(v)$ ,  $C = \{(u, v) \in \mathcal{H} \times \mathcal{H} \mid u + v = w\}$ , and

$D = C \cap \text{lev}_{\leq \eta} F$ , where  $\eta \in ](f^* \square g^*)(w), +\infty[$ . Then  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ ,  $C$  is closed and convex, and  $D$  is nonempty. Moreover,  $D \subset S_{\eta, \|w\|}$  and, as shown above,  $S_{\eta, \|w\|}$  is bounded. Hence,  $D$  is bounded and it follows from Theorem 11.9 that  $F$  achieves its infimum on  $C$ . We conclude that  $f^* \square g^*$  is exact at  $w$ .  $\square$

**Theorem 15.3 (Attouch–Brézis)** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that the conical hull of  $\text{dom } f - \text{dom } g$  is a closed linear subspace, i.e.,*

$$0 \in \text{sri}(\text{dom } f - \text{dom } g). \quad (15.9)$$

*Then  $(f + g)^* = f^* \square g^* \in \Gamma_0(\mathcal{H})$ .*

*Proof.* Since  $f + g \in \Gamma_0(\mathcal{H})$ , Corollary 13.33 implies that  $(f + g)^* \in \Gamma_0(\mathcal{H})$ . Let us fix  $z \in \text{dom } f \cap \text{dom } g$ , which is nonempty by (15.9), and let us set  $\varphi: x \mapsto f(x + z)$  and  $\psi: y \mapsto g(y + z)$ . Note that  $0 \in \text{dom } \varphi \cap \text{dom } \psi$  and that  $\text{dom } \varphi - \text{dom } \psi = \text{dom } f - \text{dom } g$ . Now set  $K = \text{cone}(\text{dom } \varphi - \text{dom } \psi) = \overline{\text{span}}(\text{dom } \varphi - \text{dom } \psi)$ . Then

$$\text{dom } \varphi \subset K \quad \text{and} \quad \text{dom } \psi \subset K. \quad (15.10)$$

It follows from (15.9) that, in the Hilbert space  $K$ , we have

$$0 \in \text{core}(\text{dom } \varphi|_K - \text{dom } \psi|_K). \quad (15.11)$$

Now set  $h = \langle z \mid \cdot \rangle$  and let  $u \in \mathcal{H}$ . By invoking Proposition 13.20(iii), (15.10), Proposition 13.20(vi), and (15.11), and then applying Proposition 15.2 in  $K$  to the functions  $\varphi|_K \in \Gamma_0(K)$  and  $\psi|_K \in \Gamma_0(K)$ , we obtain

$$\begin{aligned} (f + g)^*(u) - h(u) &= (\varphi + \psi)^*(u) \\ &= (\varphi|_K + \psi|_K)^*(P_K u) \\ &= ((\varphi|_K)^* \square (\psi|_K)^*)(P_K u) \\ &= \min_{v \in K} \left( (\varphi|_K)^*(v) + (\psi|_K)^*(P_K u - v) \right) \\ &= \min_{w \in \mathcal{H}} \left( (\varphi|_K)^*(P_K w) + (\psi|_K)^*(P_K(u - w)) \right) \\ &= \min_{w \in \mathcal{H}} \left( \varphi^*(w) + \psi^*(u - w) \right) \\ &= (\varphi^* \square \psi^*)(u) \\ &= ((f^* - h) \square (g^* - h))(u) \\ &= (f^* \square g^*)(u) - h(u). \end{aligned} \quad (15.12)$$

Consequently,  $(f + g)^*(u) = (f^* \square g^*)(u)$ .  $\square$

**Remark 15.4** The following examples show that the assumptions in Theorem 15.3 are tight.

- (i) Suppose that  $\mathcal{H} = \mathbb{R}^2$ , let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x = (\xi_1, \xi_2) \mapsto -\sqrt{\xi_1 \xi_2}$  if  $x \in \mathbb{R}_+^2$ ;  $+\infty$  otherwise, and let  $g = \iota_{\{0\} \times \mathbb{R}}$ . Then  $f + g = \iota_{\{0\} \times \mathbb{R}_+}$ ,  $f^*(\nu_1, \nu_2) = 0$  if  $\nu_1 \leq 1/(4\nu_2) < 0$ ;  $+\infty$  otherwise, and  $g^* = \iota_{\mathbb{R} \times \{0\}}$ . Thus,  $(f + g)^* = \iota_{\mathbb{R} \times \mathbb{R}_-} \neq \iota_{\mathbb{R} \times \mathbb{R}_-} = f^* \square g^*$ . Here  $\text{cone}(\text{dom } f - \text{dom } g) = \mathbb{R}_+ \times \mathbb{R}$  is only a closed cone, not a closed linear subspace.
- (ii) Suppose that  $\mathcal{H}$  is infinite-dimensional, and let  $U$  and  $V$  be closed linear subspaces of  $\mathcal{H}$  such that  $U + V$  is not closed (see Example 3.34) or, equivalently, such that  $U^\perp + V^\perp$  is not closed (see Corollary 15.35 below). Set  $f = \iota_U$  and  $g = \iota_V$ . Then Proposition 6.34 implies that  $(f + g)^* = \iota_{\overline{U^\perp + V^\perp}}$ , whereas  $f^* \square g^* = \iota_{U^\perp + V^\perp}$ . In this example,  $\text{cone}(\text{dom } f - \text{dom } g)$  is a linear subspace (equivalently,  $0 \in \text{ri}(\text{dom } f - \text{dom } g)$ ) that is not closed. Therefore, Theorem 15.3 fails if the strong relative interior is replaced by the relative interior.
- (iii) Suppose that  $\mathcal{H}$  is infinite-dimensional, let  $f$  be as in Example 9.21, and let  $g = \iota_{\{0\}}$ . Then  $f + g = g$  and  $(f + g)^* \equiv 0$ . Since  $f^* \equiv +\infty$  by Proposition 13.10(ii), we have  $f^* \square g^* \equiv +\infty$ . Hence  $(f + g)^* \neq f^* \square g^*$  even though  $\text{dom } f - \text{dom } g = \mathcal{H}$ . Therefore, assuming the lower semicontinuity of  $f$  and  $g$  is necessary.

**Proposition 15.5** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Suppose that  $\text{dom } f \cap \text{dom } g \neq \emptyset$  and that one of the following holds:*

- (i)  $\text{cone}(\text{dom } f - \text{dom } g) = \overline{\text{span}}(\text{dom } f - \text{dom } g)$ .
- (ii)  $0 \in \text{core}(\text{dom } f - \text{dom } g)$ .
- (iii)  $0 \in \text{int}(\text{dom } f - \text{dom } g)$ .
- (iv)  $\text{cont } f \cap \text{dom } g \neq \emptyset$ .
- (v)  $\mathcal{H}$  is finite-dimensional and  $\text{ri}(\text{dom } f) \cap \text{ri}(\text{dom } g) \neq \emptyset$ .

Then  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$ , i.e., (15.9) holds.

*Proof.* The assertions follow from Proposition 6.19 and Corollary 8.30.  $\square$

**Remark 15.6** The conditions in Proposition 15.5 are not equivalent. For instance, in  $\mathcal{H} = \mathbb{R}^2$ , take  $\text{dom } f = \mathbb{R} \times \{0\}$  and  $\text{dom } g = \{0\} \times \mathbb{R}$ . Then (iv) is not satisfied but (iii) is. On the other hand, if  $\text{dom } f = [0, 1] \times \{0\}$  and  $\text{dom } g = [0, 1] \times \{0\}$ , then (ii) is not satisfied but (15.9) and (v) are.

The following result, which extends Proposition 12.14, provides conditions under which the infimal convolution is lower semicontinuous.

**Proposition 15.7** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Suppose that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } f^* - \text{dom } g^*)$ .
- (ii)  $f + g^\vee$  is coercive and  $0 \in \text{sri}(\text{dom } f - \text{dom } g^\vee)$ .
- (iii)  $f$  is coercive and  $g$  is bounded below.
- (iv)  $\text{dom } f^* = \mathcal{H}$ .
- (v)  $f$  is supercoercive.

Then  $f \sqcap g = f \sqcap g \in \Gamma_0(\mathcal{H})$ .

*Proof.* (i): Apply Theorem 15.3 to  $f^*$  and  $g^*$ .

(ii) $\Rightarrow$ (i): Theorem 15.3 implies that  $(f + g^\vee)^* = f^* \sqcap g^{*\vee} \in \Gamma_0(\mathcal{H})$ . Hence, by Proposition 14.16,  $0 \in \text{int dom}(f + g^\vee)^* = \text{int dom}(f^* \sqcap g^{*\vee}) = \text{int}(\text{dom } f^* + \text{dom } g^{*\vee}) = \text{int}(\text{dom } f^* - \text{dom } g^*) \subset \text{sri}(\text{dom } f^* - \text{dom } g^*)$ .

(iii) $\Rightarrow$ (i): Proposition 14.16 yields  $0 \in \text{int dom } f^*$ . On the other hand,  $0 \in \text{dom } g^*$  since  $g^*(0) = -\inf g(\mathcal{H}) < +\infty$ . Hence,  $0 \in \text{int}(\text{dom } f^* - \text{dom } g^*) \subset \text{sri}(\text{dom } f^* - \text{dom } g^*)$ .

(iv) $\Rightarrow$ (i): Clear.

(v) $\Rightarrow$ (iv): Proposition 14.15. □

**Corollary 15.8** *Let  $\varphi$  and  $\psi$  be functions in  $\Gamma_0(\mathcal{H} \times \mathcal{K})$ . Set*

$$F: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]: (x, y) \mapsto (\varphi(x, \cdot) \sqcap \psi(x, \cdot))(y), \quad (15.13)$$

*and assume that*

$$0 \in \text{sri } Q_1(\text{dom } \varphi - \text{dom } \psi), \quad (15.14)$$

*where  $Q_1: \mathcal{H} \times \mathcal{K} \rightarrow \mathcal{H}: (x, y) \mapsto x$ . Then*

$$F^*: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]: (u, v) \mapsto (\varphi^*(\cdot, v) \sqcap \psi^*(\cdot, v))(u). \quad (15.15)$$

*Proof.* Set  $\mathcal{H} = \mathcal{H} \times \mathcal{K} \times \mathcal{K}$  and define  $\Phi: \mathcal{H} \rightarrow ]-\infty, +\infty]: (x, y_1, y_2) \mapsto \varphi(x, y_1)$  and  $\Psi: \mathcal{H} \rightarrow ]-\infty, +\infty]: (x, y_1, y_2) \mapsto \psi(x, y_2)$ . Then  $\Phi$  and  $\Psi$  belong to  $\Gamma_0(\mathcal{H})$ , and

$$(\forall (u, v_1, v_2) \in \mathcal{H}) \quad \begin{cases} \Phi^*(u, v_1, v_2) = \varphi^*(u, v_1) + \iota_{\{0\}}(v_2), \\ \Psi^*(u, v_1, v_2) = \psi^*(u, v_2) + \iota_{\{0\}}(v_1). \end{cases} \quad (15.16)$$

Now define  $G \in \Gamma_0(\mathcal{H})$  by

$$G: (x, y_1, y_2) \mapsto \varphi(x, y_1) + \psi(x, y_2) = \Phi(x, y_1, y_2) + \Psi(x, y_1, y_2) \quad (15.17)$$

and set

$$L: \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{K}: (x, y_1, y_2) \mapsto (x, y_1 + y_2). \quad (15.18)$$

Then  $L \triangleright G = F$ ,  $L^*: \mathcal{H} \times \mathcal{K} \rightarrow \mathcal{H}: (u, v) \mapsto (u, v, v)$ , and since (15.14) is equivalent to  $0 \in \text{sri}(\text{dom } \Phi - \text{dom } \Psi)$ , Theorem 15.3 yields  $G^* = \Phi^* \sqcap \Psi^*$ . Altogether, Proposition 13.21(iv) implies that  $(\forall (u, v) \in \mathcal{H} \times \mathcal{K})$   $F^*(u, v) = (L \triangleright G)^*(u, v) = G^*(L^*(u, v)) = (\Phi^* \sqcap \Psi^*)(u, v, v)$ . In view of (15.16), this is precisely (15.15). □

## 15.2 Fenchel Duality

We consider the problem of minimizing the sum of two proper functions.

**Proposition 15.9** *Let  $f$  and  $g$  be proper functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$ . Then*

$$(\forall x \in \mathcal{H})(\forall u \in \mathcal{H}) \quad f(x) + g(x) \geq -f^*(u) - g^*(-u) \quad (15.19)$$

and

$$\inf(f + g)(\mathcal{H}) \geq -\inf(f^* + g^{*\vee})(\mathcal{H}). \quad (15.20)$$

*Proof.* Using Proposition 13.9(ii) and Proposition 13.13, we see that

$$\begin{aligned} (\forall x \in \mathcal{H})(\forall u \in \mathcal{H}) \quad f(x) + g(x) + f^*(u) + g^*(-u) &\geq \langle x \mid u \rangle + \langle x \mid -u \rangle \\ &= 0, \end{aligned} \quad (15.21)$$

and the results follow.  $\square$

**Definition 15.10** The *primal problem* associated with the sum of two proper functions  $f: \mathcal{H} \rightarrow ] -\infty, +\infty]$  and  $g: \mathcal{H} \rightarrow ] -\infty, +\infty]$  is

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(x), \quad (15.22)$$

its *dual problem* is

$$\underset{u \in \mathcal{H}}{\text{minimize}} \quad f^*(u) + g^*(-u), \quad (15.23)$$

the *primal optimal value* is  $\mu = \inf(f + g)(\mathcal{H})$ , the *dual optimal value* is  $\mu^* = \inf(f^* + g^{*\vee})(\mathcal{H})$ , and the *duality gap* is

$$\Delta(f, g) = \begin{cases} 0, & \text{if } \mu = -\mu^* \in \{-\infty, +\infty\}; \\ \mu + \mu^*, & \text{otherwise.} \end{cases} \quad (15.24)$$

**Remark 15.11** Technically, the dual problem depends on the ordered pair  $(f, g)$ . We follow here the common usage.

**Proposition 15.12** *Let  $f$  and  $g$  be proper functions from  $\mathcal{H}$  to  $] -\infty, +\infty]$ , and set  $\mu = \inf(f + g)(\mathcal{H})$  and  $\mu^* = \inf(f^* + g^{*\vee})(\mathcal{H})$ . Then the following hold:*

- (i)  $\mu \geq -\mu^*$ .
- (ii)  $\Delta(f, g) \in [0, +\infty]$ .
- (iii)  $\mu = -\mu^* \Leftrightarrow \Delta(f, g) = 0$ .

*Proof.* Clear from Proposition 15.9 and Definition 15.10.  $\square$

The next proposition describes a situation in which the duality gap (15.24) is 0 and in which the dual problem admits a solution.

**Proposition 15.13** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that*

$$0 \in \text{sri}(\text{dom } f - \text{dom } g). \quad (15.25)$$

*Then  $\inf(f + g)(\mathcal{H}) = -\min(f^* + g^{*\vee})(\mathcal{H})$ .*

*Proof.* Proposition 13.9(i) and Theorem 15.3 imply that  $\inf(f + g)(\mathcal{H}) = -(f + g)^*(0) = -(f^* \square g^*)(0) = -\min(f^* + g^{*\vee})(\mathcal{H})$ .  $\square$

**Corollary 15.14** *Let  $f \in \Gamma_0(\mathcal{H})$  and let  $K$  be a closed convex cone in  $\mathcal{H}$  such that  $0 \in \text{sri}(K - \text{dom } f)$ . Then  $\inf f(K) = -\min f^*(K^\oplus)$ .*

*Proof.* Set  $g = \iota_K$  in Proposition 15.13, and use Example 13.3(ii) and Definition 6.9.  $\square$

**Corollary 15.15** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$ . Suppose that  $f + g \geq 0$  and that  $g^* = g \circ L$ , where  $L \in \mathcal{B}(\mathcal{H})$ . Then there exists  $v \in \mathcal{H}$  such that  $f^*(v) + g(-Lv) \leq 0$ .*

*Proof.* By Proposition 15.13, there exists  $v \in \mathcal{H}$  such that  $0 \leq \inf(f + g)(\mathcal{H}) = -f^*(v) - g^*(-v)$ . Hence  $0 \geq f^*(v) + g(-Lv)$ .  $\square$

**Corollary 15.16** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$ . Suppose that  $g^* = g^\vee$  and that  $f + g \geq 0$ . Then there exists  $v \in \mathcal{H}$  such that  $f^*(v) + g(v) \leq 0$ , i.e., such that  $(\forall x \in \mathcal{H}) f(x) + g(x) \geq g(x) + \langle x | v \rangle + g(v) \geq 0$ .*

*Proof.* Applying Corollary 15.15 with  $L = -\text{Id}$ , we obtain the existence of  $v \in \mathcal{H}$  such that  $0 \geq f^*(v) + g(v) = \sup_{x \in \mathcal{H}} (\langle x | v \rangle - f(x)) + g(v)$ . Hence, Proposition 13.13 yields  $(\forall x \in \mathcal{H}) f(x) + g(x) \geq g(x) + \langle x | v \rangle + g(v) = g(x) + g^*(-v) - \langle x | -v \rangle \geq 0$ .  $\square$

**Corollary 15.17** *Let  $f \in \Gamma_0(\mathcal{H})$  and set  $q = (1/2)\|\cdot\|^2$ . Suppose that  $f + q \geq 0$ . Then there exists a vector  $w \in \mathcal{H}$  such that  $(\forall x \in \mathcal{H}) f(x) + q(x) \geq q(x - w)$ .*

*Proof.* Applying Corollary 15.16 with  $g = q$  yields a vector  $v$  in  $\mathcal{H}$  such that  $(\forall x \in \mathcal{H}) f(x) + q(x) \geq q(x) + \langle x | v \rangle + q(v) = q(x + v)$ . Hence, the conclusion follows with  $w = -v$ .  $\square$

## 15.3 Fenchel–Rockafellar Duality

We now turn our attention to a more general variational problem involving a linear operator.

**Proposition 15.18** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$  be proper, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then*

$$(\forall x \in \mathcal{H})(\forall v \in \mathcal{K}) \quad f(x) + g(Lx) \geq -f^*(L^*v) - g^*(-v) \quad (15.26)$$

and

$$\inf(f + g \circ L)(\mathcal{H}) \geq -\inf(f^* \circ L^* + g^{*\vee})(\mathcal{K}). \quad (15.27)$$

*Proof.* Using Proposition 13.9(ii) and Proposition 13.13, we see that for every  $x \in \mathcal{H}$  and every  $v \in \mathcal{K}$ ,

$$f(x) + g(Lx) + f^*(L^*v) + g^*(-v) \geq \langle x \mid L^*v \rangle + \langle Lx \mid -v \rangle = 0, \quad (15.28)$$

and the result follows.  $\square$

**Definition 15.19** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ , let  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . The *primal problem* associated with the composite function  $f + g \circ L$  is

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(Lx), \quad (15.29)$$

its *dual problem* is

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad f^*(L^*v) + g^*(-v), \quad (15.30)$$

the *primal optimal value* is  $\mu = \inf (f + g \circ L)(\mathcal{H})$ , the *dual optimal value* is  $\mu^* = \inf (f^* \circ L^* + g^{*\vee})(\mathcal{K})$ , and the *duality gap* is

$$\Delta(f, g, L) = \begin{cases} 0, & \text{if } \mu = -\mu^* \in \{-\infty, +\infty\}; \\ \mu + \mu^*, & \text{otherwise.} \end{cases} \quad (15.31)$$

**Remark 15.20** As observed in Remark 15.11, the dual problem depends on the ordered triple  $(f, g, L)$ . We follow here the common usage.

The next proposition extends Proposition 15.12.

**Proposition 15.21** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$  be proper, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Set  $\mu = \inf (f + g \circ L)(\mathcal{H})$  and  $\mu^* = \inf (f^* \circ L^* + g^{*\vee})(\mathcal{K})$ . Then the following hold:

- (i)  $\mu \geq -\mu^*$ .
- (ii)  $\Delta(f, g, L) \in [0, +\infty]$ .
- (iii)  $\mu = -\mu^* \Leftrightarrow \Delta(f, g, L) = 0$ .

*Proof.* Clear from Proposition 15.18 and Definition 15.19.  $\square$

A zero duality gap is not automatic, but it is guaranteed under additional assumptions.

**Proposition 15.22** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that  $0 \in \text{core}(\text{dom } g - L(\text{dom } f))$ . Then

$$\inf (f + g \circ L)(\mathcal{H}) = -\min (f^* \circ L^* + g^{*\vee})(\mathcal{K}). \quad (15.32)$$

*Proof.* Set  $\varphi = f \oplus g$  and  $V = \text{gra } L$ . Now take  $(x, y) \in \mathcal{H} \times \mathcal{K}$ . Since  $0 \in \text{core}(\text{dom } g - L(\text{dom } f))$ , there exist  $\gamma \in \mathbb{R}_{++}$ ,  $a \in \text{dom } f$ , and  $b \in \text{dom } g$  such that  $y - Lx = \gamma(b - La)$ . Upon setting  $z = a - x/\gamma$ , we obtain  $x = \gamma(a - z)$  and  $y = \gamma(b - Lz)$ . Therefore,  $(x, y) = \gamma((a, b) - (z, Lz)) \in \text{cone}((\text{dom } \varphi) - V)$ . We have thus shown that  $\text{cone}((\text{dom } \varphi) - V) = \mathcal{H} \times \mathcal{K}$ , which implies that  $0 \in$



$\text{core}(V - \text{dom } \varphi) \subset \text{sri}(V - \text{dom } \varphi)$ . It then follows from Corollary 15.14 that  $\inf \varphi(V) = -\min \varphi^*(V^\perp)$ . However,  $V^\perp = \{(u, v) \in \mathcal{H} \times \mathcal{K} \mid u = -L^*v\}$  and, by Proposition 13.27,  $\varphi^* = f^* \oplus g^*$ . Therefore,  $\inf(f + g \circ L)(\mathcal{H}) = \inf \varphi(V) = -\min \varphi^*(V^\perp) = -\min((f^* \circ L^*)^\vee + g^*)(\mathcal{K}) = -\min(f^* \circ L^* + g^{*\vee})(\mathcal{K})$ .  $\square$

**Theorem 15.23** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that (see Proposition 6.19 for special cases)*

$$0 \in \text{sri}(\text{dom } g - L(\text{dom } f)). \quad (15.33)$$

Then  $\inf(f + g \circ L)(\mathcal{H}) = -\min(f^* \circ L^* + g^{*\vee})(\mathcal{K})$ .

*Proof.* We first consider the special case  $0 \in \text{dom } f$  and  $0 \in \text{dom } g$ . Set  $\mathcal{A} = \overline{\text{span}} \text{dom } f$  and  $\mathcal{B} = \text{cone}(\text{dom } g - L(\text{dom } f))$ . By (15.33),  $\mathcal{B} = \overline{\text{span}}(\text{dom } g - L(\text{dom } f))$ . Hence  $\text{dom } f \subset \mathcal{A}$ ,  $\text{dom } g \subset \mathcal{B}$ , and, since  $L(\text{dom } f) \subset \mathcal{B}$ , we have  $L(\mathcal{A}) \subset \mathcal{B}$ . It follows that  $\text{ran } L|_{\mathcal{A}} \subset \mathcal{B}$ , and, in turn, using Fact 2.18(iv), that

$$\mathcal{B}^\perp \subset (\text{ran } L|_{\mathcal{A}})^\perp = \ker((L|_{\mathcal{A}})^*) = \ker P_{\mathcal{A}} L^*. \quad (15.34)$$

Therefore

$$P_{\mathcal{A}} L^* = P_{\mathcal{A}} L^* (P_{\mathcal{B}} + P_{\mathcal{B}^\perp}) = P_{\mathcal{A}} L^* P_{\mathcal{B}} = P_{\mathcal{A}} L^*|_{\mathcal{B}} P_{\mathcal{B}}. \quad (15.35)$$

Next, we observe that condition (15.33) in  $\mathcal{K}$  yields

$$0 \in \text{core}(\text{dom } g|_{\mathcal{B}} - (P_{\mathcal{B}} L|_{\mathcal{A}})(\text{dom } f|_{\mathcal{A}})) \quad (15.36)$$

in  $\mathcal{B}$ . Thus, using the inclusions  $\text{dom } f \subset \mathcal{A}$  and  $L(\mathcal{A}) \subset \mathcal{B}$ , (15.36), Proposition 15.22, (15.35), and Proposition 13.20(vi), we obtain

$$\begin{aligned} \inf_{x \in \mathcal{H}} (f(x) + g(Lx)) &= \inf_{x \in \mathcal{A}} (f|_{\mathcal{A}}(x) + g|_{\mathcal{B}}((P_{\mathcal{B}} L|_{\mathcal{A}})x)) \\ &= -\min_{v \in \mathcal{B}} \left( (f|_{\mathcal{A}})^*((P_{\mathcal{B}} L|_{\mathcal{A}})^*v) + (g|_{\mathcal{B}})^*(-v) \right) \\ &= -\min_{v \in \mathcal{K}} \left( (f|_{\mathcal{A}})^*((P_{\mathcal{A}} L^*|_{\mathcal{B}})(P_{\mathcal{B}}v)) + (g|_{\mathcal{B}})^*(P_{\mathcal{B}}(-v)) \right) \\ &= -\min_{v \in \mathcal{K}} \left( (f|_{\mathcal{A}})^*(P_{\mathcal{A}} L^*v) + (g|_{\mathcal{B}})^*(P_{\mathcal{B}}(-v)) \right) \\ &= -\min_{v \in \mathcal{K}} (f^*(L^*v) + g^*(-v)). \end{aligned} \quad (15.37)$$

We now consider the general case. In view of (15.33), there exist  $b \in \text{dom } g$  and  $a \in \text{dom } f$  such that  $b = La$ . Set  $\varphi: x \mapsto f(x + a)$  and  $\psi: y \mapsto g(y + b)$ . Then  $0 \in \text{dom } \varphi$ ,  $0 \in \text{dom } \psi$ , and  $\text{dom } \psi - L(\text{dom } \varphi) = \text{dom } g - L(\text{dom } f)$ . We therefore apply the above special case with  $\varphi$  and  $\psi$  to obtain

$$\begin{aligned} \inf_{x \in \mathcal{H}} (f(x) + g(Lx)) &= \inf_{x \in \mathcal{H}} (\varphi(x) + \psi(Lx)) \\ &= -\min_{v \in \mathcal{K}} (\varphi^*(L^*v) + \psi^*(-v)) \end{aligned}$$

$$\begin{aligned}
&= -\min_{v \in \mathcal{K}} (f^*(L^*v) - \langle L^*v \mid a \rangle + g^*(-v) - \langle -v \mid b \rangle) \\
&= -\min_{v \in \mathcal{K}} (f^*(L^*v) + g^*(-v)), \tag{15.38}
\end{aligned}$$

where we have used Proposition 13.20(iii) and the identity  $La = b$ .  $\square$

**Proposition 15.24** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$  and that one of the following holds:*

- (i)  $\text{cone}(\text{dom } g - L(\text{dom } f)) = \overline{\text{span}}(\text{dom } g - L(\text{dom } f))$ .
- (ii)  $\text{dom } g - L(\text{dom } f)$  is a closed linear subspace.
- (iii)  $\text{dom } f$  and  $\text{dom } g$  are linear subspaces and  $\text{dom } g + L(\text{dom } f)$  is closed.
- (iv)  $\text{dom } g$  is a cone and  $\text{dom } g - \text{cone } L(\text{dom } f)$  is a closed linear subspace.
- (v)  $0 \in \text{core}(\text{dom } g - L(\text{dom } f))$ .
- (vi)  $0 \in \text{int}(\text{dom } g - L(\text{dom } f))$ .
- (vii)  $\text{cont } g \cap L(\text{dom } f) \neq \emptyset$ .
- (viii)  $\mathcal{K}$  is finite-dimensional and  $(\text{ri } \text{dom } g) \cap (\text{ri } L(\text{dom } f)) \neq \emptyset$ .
- (ix)  $\mathcal{K}$  is finite-dimensional and  $(\text{ri } \text{dom } g) \cap L(\text{qri } \text{dom } f) \neq \emptyset$ .

Then  $0 \in \text{sri}(\text{dom } g - L(\text{dom } f))$ , i.e., (15.33) holds.

*Proof.* The assertions follow from Proposition 8.2, Proposition 3.5, (6.8), Proposition 6.19, and Corollary 8.30.  $\square$

We now turn to a result that is complementary to Theorem 15.23 and that relies on several external results drawn from [219]. To formulate it, we require the notions of polyhedral (convex) set and function.

A subset of  $\mathcal{H}$  is *polyhedral* if it is a finite intersection of closed half-spaces, and a function  $f: \mathcal{H} \rightarrow [-\infty, +\infty]$  is *polyhedral* if  $\text{epi } f$  is a polyhedral set.

**Fact 15.25** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$  be polyhedral, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that  $\mathcal{K}$  is finite-dimensional and that one of the following holds:*

- (i)  $\text{dom } g \cap \text{ri } L(\text{dom } f) \neq \emptyset$ .
- (ii)  $\mathcal{H}$  is finite-dimensional,  $f$  is polyhedral, and  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ .

Then  $\inf(f + g \circ L)(\mathcal{H}) = -\min((f^* \circ L^*) + g^{*\vee})(\mathcal{K})$ .

*Proof.* We set  $\mu = \inf(f + g \circ L)(\mathcal{H})$  and, in view of Proposition 15.21(i), we assume that  $\mu > -\infty$ . Then

$$\begin{aligned}
\mu &= \inf_{(x,y) \in \text{gra } L} (f(x) + g(y)) \\
&= \inf_{y \in \mathcal{K}} \left( g(y) + \inf_{x \in L^{-1}y} f(x) \right) \\
&= \inf_{y \in \mathcal{K}} (g(y) + (L \triangleright f)(y)). \tag{15.39}
\end{aligned}$$

Proposition 12.34(i) and Proposition 13.21(iv) assert that  $\text{dom}(L \triangleright f) = L(\text{dom } f)$  and that  $(L \triangleright f)^* = f^* \circ L^*$ . If (i) holds, then let  $z \in \text{dom } f$  be

such that  $Lz \in \text{dom } g \cap \text{ri } L(\text{dom } f) = \text{dom } g \cap \text{ri } \text{dom}(L \triangleright f)$ ; otherwise, (ii) holds and we let  $z \in \text{dom } f$  be such that  $Lz \in \text{dom } g \cap L(\text{dom } f) = \text{dom } g \cap \text{dom}(L \triangleright f)$ . In both cases,  $\mu \leq g(Lz) + (L \triangleright f)(Lz) \leq g(Lz) + f(z) < +\infty$  and thus

$$\mu \in \mathbb{R}. \quad (15.40)$$

If  $(L \triangleright f)(Lz) = -\infty$ , then there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\mathcal{H}$  such that  $Lx_n \equiv Lz$  and  $f(x_n) \rightarrow -\infty$ , which implies that  $\mu \leq f(x_n) + g(Lx_n) = f(x_n) + g(Lz) \rightarrow -\infty$ , a contradiction to (15.40). Hence

$$(L \triangleright f)(Lz) \in \mathbb{R}. \quad (15.41)$$

If (i) holds, then  $L \triangleright f$  is convex by Proposition 12.34(ii) and proper by [219, Theorem 7.2]. If (ii) holds, then  $L \triangleright f$  is polyhedral by [219, Corollary 19.3.1] and also proper. Therefore, [219, Theorem 31.1] yields

$$\begin{aligned} \inf_{y \in \mathcal{K}} (g(y) + (L \triangleright f)(y)) &= -\min_{v \in \mathcal{K}} (g^{*\vee}(v) + (L \triangleright f)^*(v)) \\ &= -\min_{v \in \mathcal{K}} (g^{*\vee}(v) + f^*(L^*v)). \end{aligned} \quad (15.42)$$

The conclusion follows by combining (15.39) with (15.42).  $\square$

## 15.4 A Conjugation Result

We start with a general conjugation formula.

**Proposition 15.26** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ . Then  $(f + g \circ L)^* = (f^* \square (L^* \triangleright g^*))^{**}$ . Moreover,  $f^* \square (L^* \triangleright g^*)$  is proper and convex, and it possesses a continuous affine minorant.*

*Proof.* It follows from Corollary 13.33 and Corollary 13.22(i) that  $(f + g \circ L)^* = ((f^*)^* + ((g^*)^* \circ (L^*)^*))^* = (f^* \square (L^* \triangleright g^*))^{**}$ . The remaining statements follow as in the proof of Proposition 15.1. Indeed, since  $f + g \circ L \in \Gamma_0(\mathcal{H})$ , we have  $f + g \circ L = (f + g \circ L)^{**} = (f^* \square (L^* \triangleright g^*))^{***} = (f^* \square (L^* \triangleright g^*))^*$ . Hence  $f^* \square (L^* \triangleright g^*)$  is proper by Proposition 13.9(iii), convex by Proposition 12.34(ii) and Proposition 12.11, and it possesses a continuous affine minorant by Proposition 13.10(ii).  $\square$

We now obtain an extension of Theorem 15.3.

**Theorem 15.27** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } g - L(\text{dom } f))$ .
- (ii)  $\mathcal{K}$  is finite-dimensional,  $g$  is polyhedral, and  $\text{dom } g \cap \text{ri } L(\text{dom } f) \neq \emptyset$ .

- (iii)  $\mathcal{H}$  and  $\mathcal{K}$  are finite-dimensional,  $f$  and  $g$  are polyhedral, and  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ .

Then  $(f + g \circ L)^* = f^* \square (L^* \triangleright g^*)$ . In other words,

$$(\forall u \in \mathcal{H}) \quad (f + g \circ L)^*(u) = \min_{v \in \mathcal{K}} (f^*(u - L^*v) + g^*(v)). \quad (15.43)$$

*Proof.* Let  $u \in \mathcal{H}$ . Then it follows from Theorem 15.23 or from Fact 15.25 that

$$\begin{aligned} (f + g \circ L)^*(u) &= \sup_{x \in \mathcal{H}} (\langle x \mid u \rangle - f(x) - g(Lx)) \\ &= - \inf_{x \in \mathcal{H}} (f(x) - \langle x \mid u \rangle + g(Lx)) \\ &= \min_{v \in \mathcal{K}} (f^*(L^*v + u) + g^*(-v)), \end{aligned} \quad (15.44)$$

which yields the result.  $\square$

**Corollary 15.28** *Let  $g \in \Gamma_0(\mathcal{K})$  and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } g - \text{ran } L)$ .
- (ii)  $\mathcal{K}$  is finite-dimensional,  $g$  is polyhedral, and  $\text{dom } g \cap \text{ran } L \neq \emptyset$ .

Then  $(g \circ L)^* = L^* \triangleright g^*$ . In other words,

$$(\forall u \in \mathcal{H}) \quad (g \circ L)^*(u) = \min_{\substack{v \in \mathcal{K} \\ L^*v = u}} g^*(v) = (L^* \triangleright g^*)(u). \quad (15.45)$$

## 15.5 Applications

**Example 15.29** Let  $C \subset \mathcal{H}$  and  $D \subset \mathcal{K}$  be closed convex sets, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $0 \in \text{sri}(\text{bar } D - L(C))$ . Then

$$\inf_{x \in C} \sup_{v \in D} \langle Lx \mid v \rangle = \max_{v \in D} \inf_{x \in C} \langle Lx \mid v \rangle. \quad (15.46)$$

*Proof.* Set  $f = \iota_C$  and  $g = \sigma_D$  in Theorem 15.23. Then Example 13.3(i) and Example 13.37(i) yield  $\inf_{x \in C} \sup_{v \in D} \langle Lx \mid v \rangle = \inf_{x \in \mathcal{H}} \iota_C(x) + \sigma_D(Lx) = \inf_{x \in \mathcal{H}} f(x) + g(Lx) = - \min_{v \in \mathcal{H}} f^*(-L^*v) + g^*(v) = - \min_{v \in D} \sigma_C(-L^*v) = - \min_{v \in D} \sup_{x \in C} \langle x \mid -L^*v \rangle = \max_{v \in D} \inf_{x \in C} \langle Lx \mid v \rangle$ .  $\square$

In Euclidean spaces, the following result is known as von Neumann's min–imax theorem.

**Corollary 15.30 (von Neumann)** *Let  $C \subset \mathcal{H}$  and  $D \subset \mathcal{K}$  be nonempty bounded closed convex sets, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then*

$$\min_{x \in C} \max_{v \in D} \langle Lx \mid v \rangle = \max_{v \in D} \min_{x \in C} \langle Lx \mid v \rangle. \quad (15.47)$$

*Proof.* Since  $D$  is bounded, we have  $\text{bar } D = \mathcal{K}$  and therefore  $0 \in \text{sri}(\text{bar } D - L(C))$ . Moreover, since Proposition 11.14(ii) implies that a continuous linear functional achieves its infimum and its supremum on a nonempty bounded closed convex set, we have  $\sup_{v \in D} \langle Lx \mid v \rangle = \max_{v \in D} \langle Lx \mid v \rangle$  and  $\inf_{x \in C} \langle Lx \mid v \rangle = \min_{x \in C} \langle Lx \mid v \rangle$ . Now set  $\varphi: x \mapsto \max_{v \in D} \langle Lx \mid v \rangle$ . Then  $\varphi(0) = 0$  and, therefore, Proposition 9.3 implies that  $\varphi \in \Gamma_0(\mathcal{H})$ . Thus, invoking Proposition 11.14(ii) again, derive from (15.46) that

$$\begin{aligned} \min_{x \in C} \max_{v \in D} \langle Lx \mid v \rangle &= \min_{x \in C} \varphi(x) \\ &= \inf_{x \in C} \sup_{v \in D} \langle Lx \mid v \rangle \\ &= \max_{v \in D} \inf_{x \in C} \langle Lx \mid v \rangle \\ &= \max_{v \in D} \min_{x \in C} \langle Lx \mid v \rangle, \end{aligned} \quad (15.48)$$

which yields (15.47).  $\square$

**Corollary 15.31** *Let  $C$  be a nonempty closed convex cone in  $\mathcal{H}$ , let  $D$  be a nonempty closed convex cone in  $\mathcal{K}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then the following hold:*

- (i)  $(C \cap L^{-1}(D))^{\ominus} = \overline{C^{\ominus} + L^*(D^{\ominus})}$ .
- (ii) *Suppose that  $D - L(C)$  is a closed linear subspace of  $\mathcal{K}$ . Then the set  $(C \cap L^{-1}(D))^{\ominus} = C^{\ominus} + L^*(D^{\ominus})$  is a nonempty closed convex cone.*

*Proof.* (i): Proposition 6.34 and Proposition 6.36(i) yield  $(C \cap L^{-1}(D))^{\ominus} = \overline{C^{\ominus} + (L^{-1}(D))^{\ominus}} = \overline{C^{\ominus} + \overline{L^*(D^{\ominus})}} = \overline{C^{\ominus} + L^*(D^{\ominus})}$ .

(ii): Set  $f = \iota_C$  and  $g = \iota_D$ , and note that  $0 \in \text{sri}(\text{dom } g - L(\text{dom } f))$  by assumption. Example 13.3(ii) and Theorem 15.27(i) imply that, for every  $u \in \mathcal{H}$ ,

$$\begin{aligned} \iota_{(C \cap L^{-1}(D))^{\ominus}}(u) &= \iota_{C \cap L^{-1}(D)}^*(u) \\ &= (\iota_C + (\iota_D \circ L))^*(u) \\ &= (f + g \circ L)^*(u) \\ &= \min_{v \in \mathcal{K}} (f^*(L^*v + u) + g^*(-v)) \\ &= \min_{v \in \mathcal{K}} (\iota_{C^{\ominus}}(L^*v + u) + \iota_{D^{\ominus}}(-v)) \\ &= \iota_{C^{\ominus} + L^*(D^{\ominus})}(u). \end{aligned} \quad (15.49)$$

Therefore,  $C^{\ominus} \cap L^*(D^{\ominus})$  is the polar cone of  $C \cap L^{-1}(D)$  and hence, by Proposition 6.23(ii), it is closed.  $\square$

**Corollary 15.32** *Let  $D$  be a nonempty closed convex cone in  $\mathcal{K}$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and suppose that  $D - \text{ran } L$  is a closed linear subspace. Then  $(L^{-1}(D))^{\ominus} = L^*(D^{\ominus})$ .*

**Corollary 15.33** *Let  $C$  be a closed linear subspace of  $\mathcal{H}$ , let  $D$  be a closed linear subspace of  $\mathcal{K}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then  $L(C) + D$  is closed if and only if  $C^{\perp} + L^*(D^{\perp})$  is closed.*

**Corollary 15.34** *Let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then  $\text{ran } L$  is closed if and only if  $\text{ran } L^*$  is closed.*

**Corollary 15.35** *Let  $U$  and  $V$  be closed linear subspaces of  $\mathcal{H}$ . Then  $U + V$  is closed if and only if  $U^{\perp} + V^{\perp}$  is closed.*

**Corollary 15.36** *Let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be an operator such that  $\text{ran } L$  is closed, and let  $V$  be a closed linear subspace of  $\mathcal{H}$ . Then  $L(V)$  is closed if and only if  $V + \ker L$  is closed.*

*Proof.* Since  $\text{ran } L$  is closed, Corollary 15.34 yields

$$\text{ran } L^* = \overline{\text{ran } L}^*. \quad (15.50)$$

Using Corollary 15.33, (15.50), Fact 2.18(iii), and Corollary 15.35, we obtain the equivalences  $L(V)$  is closed  $\Leftrightarrow V^{\perp} + \text{ran } L^*$  is closed  $\Leftrightarrow V^{\perp} + \overline{\text{ran } L}^*$  is closed  $\Leftrightarrow V^{\perp} + (\ker L)^{\perp}$  is closed  $\Leftrightarrow V + \ker L$  is closed.  $\square$

## Exercises

**Exercise 15.1** Let  $f \in \Gamma_0(\mathcal{H})$  and  $g \in \Gamma_0(\mathcal{H})$  be such that  $\text{dom } f \cap \text{dom } g \neq \emptyset$ . Show that  $(f + g)^* = f^* \boxplus g^*$  if and only if  $\text{epi } f^* + \text{epi } g^*$  is closed.

**Exercise 15.2** Let  $K_1$  and  $K_2$  be nonempty closed convex cones in  $\mathcal{H}$ . Use Proposition 15.1 to show that  $(K_1 \cap K_2)^{\ominus} = \overline{K_1^{\ominus} + K_2^{\ominus}}$ . Compare with Proposition 6.34.

**Exercise 15.3** Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $\text{dom } f + \text{dom } f^* = \mathcal{H}$ .

**Exercise 15.4** Verify the details in Remark 15.4(i).

**Exercise 15.5** Consider the setting of Definition 15.10, and let  $f$  and  $g$  be as in Remark 15.4(i). Determine  $\mu$ ,  $\mu^*$ , and  $\Delta(f, g)$ , and whether the primal and dual problems admit solutions.

**Exercise 15.6** Consider the setting of Definition 15.19 with  $\mathcal{H} = \mathcal{K} = \mathbb{R}$ ,  $L = \text{Id}$ , and  $f = g = \exp$ . Prove the following:  $\mu = \mu^* = 0$ , the primal problem has no solution, and the dual problem has a unique solution.

**Exercise 15.7** Find two functions  $f$  and  $g$  in  $\Gamma_0(\mathbb{R})$  for which, in the setting of Definition 15.10,  $\Delta(f, g) = 0$ , the primal problem has a unique solution, and the dual problem has no solution.

**Exercise 15.8** Find two functions  $f$  and  $g$  in  $\Gamma_0(\mathbb{R})$  for which  $\Delta(f, g) = 0$  and  $\inf(f + g)(\mathcal{H}) = -\infty$ .

**Exercise 15.9** Find two functions  $f$  and  $g$  in  $\Gamma_0(\mathbb{R})$  for which  $\inf(f + g)(\mathcal{H}) = +\infty$  and  $\inf(f^* + g^{*\vee})(\mathcal{H}) = -\infty$ .

**Exercise 15.10** Suppose that  $\mathcal{H} = \mathbb{R}$ . Set  $f = \iota_{\mathbb{R}_-}$  and

$$g: x \mapsto \begin{cases} x \ln(x) - x, & \text{if } x > 0; \\ 0, & \text{if } x = 0; \\ +\infty, & \text{if } x < 0. \end{cases} \quad (15.51)$$

Show that the primal problem (15.22) admits a minimizer, that  $\Delta(f, g) = 0$ , and that  $0 \notin \text{sri}(\text{dom } f - \text{dom } g)$ .

**Exercise 15.11** Suppose that  $\mathcal{H} = \mathbb{R}$ . Set  $f = \iota_{[-1, 0]}$  and  $g = \iota_{[0, 1]}$ . Show that the primal problem (15.22) and the dual problem (15.23) both admit minimizers, that  $\Delta(f, g) = 0$ , and that  $0 \notin \text{sri}(\text{dom } f - \text{dom } g)$ .

**Exercise 15.12** Find two functions  $f$  and  $g$  in  $\Gamma_0(\mathbb{R})$  such that the following hold: the primal problem (15.22) has a minimizer, the primal optimal value is 0,  $\Delta(f, g) = 0$ , the dual problem (15.23) does not have a minimizer, and  $\text{dom } f^* = \text{dom } g^* = \mathbb{R}$ .

**Exercise 15.13** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$  be proper, and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Show that  $\Delta(f, g \circ L) \leq \Delta(f, g, L)$ .

**Exercise 15.14** Suppose that  $\mathcal{H} = \mathcal{K} = \mathbb{R}$  and set  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto x + \iota_{\mathbb{R}_+}(x)$ ,  $g = \iota_{\{1\}}$ , and  $L = \text{Id}$ . Compute  $\mu$  and  $\mu^*$  as given in Proposition 15.21.

**Exercise 15.15** Suppose that  $\mathcal{H} = \mathcal{K} = \mathbb{R}$  and set  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto -x + \iota_{\mathbb{R}_+}(x)$ ,  $g = \iota_{\{-1\}}$ , and  $L: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto 0$ . Compute  $\mu$  and  $\mu^*$  as given in Proposition 15.21.

**Exercise 15.16** Suppose that  $\mathcal{H} = \mathcal{K} = \mathbb{R}$  and set  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto -x + \iota_{\mathbb{R}_+}(x)$ ,  $g = \iota_{\{0\}}$ , and  $L: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto 0$ . Compute  $\mu$  and  $\mu^*$  as given in Proposition 15.21.

**Exercise 15.17** Suppose that  $\mathcal{H} = \mathcal{K} = \mathbb{R}$  and set  $f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto -x + \iota_{\mathbb{R}_+}(x)$ ,  $g = \iota_{]-\infty, -1]}$ , and  $L: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto x$ . Compute  $\mu$  and  $\mu^*$  as given in Proposition 15.21.

**Exercise 15.18** Suppose that  $\mathcal{H} = \ell^2(\mathbb{N})$ , let  $(\alpha_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1]$  such that  $\sum_{n \in \mathbb{N}} \alpha_n^2 < +\infty$ , and set  $L: \mathcal{H} \rightarrow \mathcal{H}: (\xi_n)_{n \in \mathbb{N}} \mapsto (\alpha_n \xi_n)_{n \in \mathbb{N}}$ . Show that  $L^* = L$ , that  $\overline{\text{ran } L} = \mathcal{H}$ , and that  $\text{ran } L \neq \mathcal{H}$ . Conclude that the closure operation in Corollary 15.31(i) is essential.

**Exercise 15.19** Consider Corollary 15.31(ii). Find an example in which  $D - L(C) = \mathcal{K}$  and  $C^\ominus + L^*(D^\ominus)$  is not a linear subspace.



# Chapter 16

## Subdifferentiability

The subdifferential is a fundamental tool in the analysis of nondifferentiable convex functions. In this chapter we discuss the properties of subdifferentials and the interplay between the subdifferential and the Legendre transform. Moreover, we establish the Brøndsted–Rockafellar theorem, which asserts that the graph of the subdifferential operator is dense in the domain of the separable sum of the function and its conjugate.

### 16.1 Basic Properties

**Definition 16.1** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. The *subdifferential* of  $f$  is the set-valued operator

$$\partial f: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \{u \in \mathcal{H} \mid (\forall y \in \mathcal{H}) \quad \langle y - x \mid u \rangle + f(x) \leq f(y)\}. \quad (16.1)$$

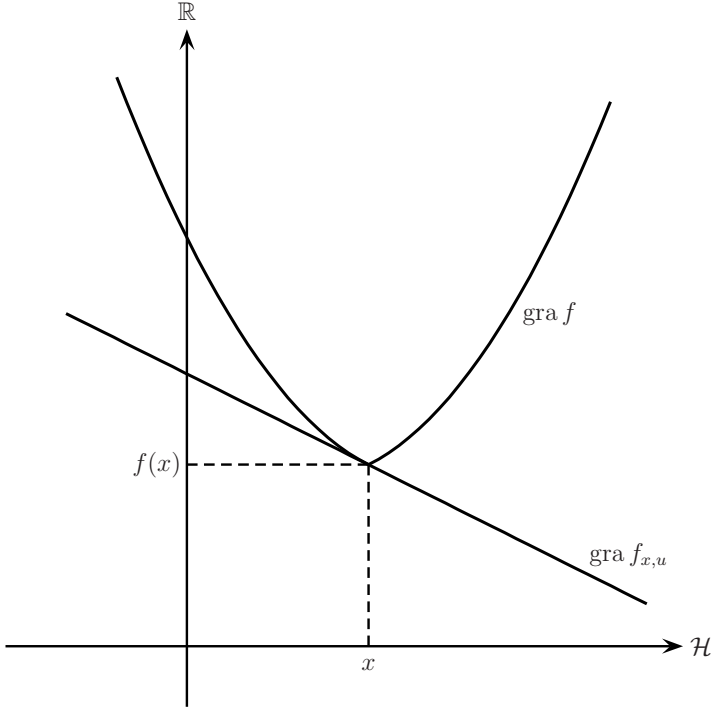
Let  $x \in \mathcal{H}$ . Then  $f$  is *subdifferentiable* at  $x$  if  $\partial f(x) \neq \emptyset$ ; the elements of  $\partial f(x)$  are the *subgradients* of  $f$  at  $x$ .

Graphically (see [Figure 16.1](#)), a vector  $u \in \mathcal{H}$  is a subgradient of a proper function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  at  $x \in \text{dom } f$  if the continuous affine functional  $f_{x,u}: y \mapsto \langle y - x \mid u \rangle + f(x)$ , which coincides with  $f$  at  $x$ , minorizes  $f$ ; in other words,  $u$  is the “slope” of a continuous affine minorant of  $f$  that coincides with  $f$  at  $x$ .

Global minimizers of proper functions can be characterized by a simple but powerful principle which goes back to the seventeenth century and the work of Pierre Fermat.

**Theorem 16.2 (Fermat’s rule)** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then

$$\text{Argmin } f = \text{zer } \partial f = \{x \in \mathcal{H} \mid 0 \in \partial f(x)\}. \quad (16.2)$$



**Fig. 16.1** A vector  $u \in \mathcal{H}$  is a subgradient of  $f$  at  $x$  if it is the “slope” of a continuous affine minorant  $f_{x,u}$  of  $f$  that coincides with  $f$  at  $x$ .

*Proof.* Let  $x \in \mathcal{H}$ . Then  $x \in \operatorname{Argmin} f \Leftrightarrow (\forall y \in \mathcal{H}) \langle y - x \mid 0 \rangle + f(x) \leq f(y) \Leftrightarrow 0 \in \partial f(x)$ .  $\square$

**Proposition 16.3** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and let  $x \in \operatorname{dom} f$ . Then the following hold:

- (i)  $\operatorname{dom} \partial f \subset \operatorname{dom} f$ .
- (ii)  $\partial f(x) = \bigcap_{y \in \operatorname{dom} f} \{u \in \mathcal{H} \mid \langle y - x \mid u \rangle \leq f(y) - f(x)\}$ .
- (iii)  $\partial f(x)$  is closed and convex.
- (iv) Suppose that  $x \in \operatorname{dom} \partial f$ . Then  $f$  is lower semicontinuous at  $x$ .

*Proof.* (i): Since  $f$  is proper,  $f(x) = +\infty \Rightarrow \partial f(x) = \emptyset$ .

(ii): Clear from (16.1).

(iii): Clear from (ii).

(iv): Take  $u \in \partial f(x)$ . Then  $(\forall y \in \mathcal{H}) f(x) \leq f(y) + \langle x - y \mid u \rangle$  and hence  $f(x) \leq \underline{\lim}_{y \rightarrow x} f(y)$ .  $\square$

**Proposition 16.4** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, and let  $x \in \mathcal{H}$ . Suppose that  $x \in \text{dom } \partial f$ . Then  $f^{**}(x) = f(x)$  and  $\partial f^{**}(x) = \partial f(x)$ .*

*Proof.* Take  $u \in \partial f(x)$ . Then  $(\forall y \in \mathcal{H}) \langle y - x \mid u \rangle + f(x) \leq f(y)$ . Hence  $f^{**} = \check{f}$  by Proposition 13.39. In turn, in view of Corollary 9.10, Lemma 1.31(iv), and Proposition 16.3(iv),  $f^{**}(x) = \check{f}(x) = \bar{f}(x) = \varinjlim_{y \rightarrow x} f(y) = f(x)$ . Thus  $\langle y - x \mid u \rangle + f^{**}(x) \leq f^{**}(y)$ , which shows that  $u \in \partial f^{**}(x)$ . Conversely, since  $f^{**}(x) = f(x)$  and since  $f^{**} \leq f$  by Proposition 13.14(i), we obtain  $\partial f^{**}(x) \subset \partial f(x)$ .  $\square$

**Proposition 16.5** *Let  $\mathcal{K}$  be a real Hilbert space, let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  and  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$  be proper, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $\lambda \in \mathbb{R}_{++}$ . Then the following hold:*

- (i)  $\partial(\lambda f) = \lambda \partial f$ .
- (ii) *Suppose that  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ . Then  $\partial f + L^* \circ (\partial g) \circ L \subset \partial(f + g \circ L)$ .*

*Proof.* (i): Clear.

(ii): Take  $x \in \mathcal{H}$ ,  $u \in \partial f(x)$ , and  $v \in \partial g(Lx)$ . Then  $u + L^*v$  is a generic point in  $\partial f(x) + (L^* \circ (\partial g) \circ L)(x)$  and it must be shown that  $u + L^*v \in \partial(f + g \circ L)(x)$ . It follows from (16.1) that, for every  $y \in \mathcal{H}$ , we have  $\langle y - x \mid u \rangle + f(x) \leq f(y)$  and  $\langle Ly - Lx \mid v \rangle + g(Lx) \leq g(Ly)$ , hence  $\langle y - x \mid L^*v \rangle + g(Lx) \leq g(Ly)$ . Adding the first and third inequalities yields

$$(\forall y \in \mathcal{H}) \quad \langle y - x \mid u + L^*v \rangle + (f + g \circ L)(x) \leq (f + g \circ L)(y) \quad (16.3)$$

and, in turn,  $u + L^*v \in \partial(f + g \circ L)(x)$ .  $\square$

**Proposition 16.6** *Let  $(\mathcal{H}_i)_{i \in I}$  be a finite totally ordered family of real Hilbert spaces, set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ , let  $\mathbf{f}: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, and let  $(x_i)_{i \in I} \in \text{dom } \mathbf{f}$ . For every  $i \in I$ , define  $R_i: \mathcal{H}_i \rightarrow \mathcal{H}$  as follows: for every  $y \in \mathcal{H}_i$  and every  $j \in I$ , the  $j$ th component of  $R_i y$  is  $y$  if  $j = i$ , and  $x_j$  otherwise. Then*

$$\partial \mathbf{f}((x_i)_{i \in I}) \subset \bigtimes_{i \in I} \partial(\mathbf{f} \circ R_i)(x_i). \quad (16.4)$$

*Proof.* Set  $\mathbf{x} = (x_i)_{i \in I}$ , and take  $\mathbf{y} = (y_i)_{i \in I} \in \mathcal{H}$  and  $\mathbf{u} = (u_i)_{i \in I} \in \mathcal{H}$ . Then

$$\begin{aligned} \mathbf{u} \in \partial \mathbf{f}(\mathbf{x}) &\Leftrightarrow (\forall \mathbf{y} \in \mathcal{H}) \quad \langle \mathbf{y} - \mathbf{x} \mid \mathbf{u} \rangle + \mathbf{f}(\mathbf{x}) \leq \mathbf{f}(\mathbf{y}) \\ &\Rightarrow (\forall i \in I) (\forall y_i \in \mathcal{H}_i) \quad \langle y_i - x_i \mid u_i \rangle + (\mathbf{f} \circ R_i)(x_i) \leq (\mathbf{f} \circ R_i)(y_i) \\ &\Leftrightarrow (\forall i \in I) \quad u_i \in \partial(\mathbf{f} \circ R_i)(x_i) \\ &\Leftrightarrow \mathbf{u} \in \bigtimes_{i \in I} \partial(\mathbf{f} \circ R_i)(x_i), \end{aligned} \quad (16.5)$$

which completes the proof.  $\square$

**Remark 16.7** The inclusion (16.4) can be strict. Indeed, adopt the notation of Proposition 16.6 and suppose that  $\mathcal{H} = \mathbb{R} \times \mathbb{R}$ . Set  $\mathbf{f} = \iota_{B(0;1)}$  and  $\mathbf{x} = (1, 0)$ . Then  $\mathbf{f} \circ R_1 = \iota_{[-1,1]}$  and  $\mathbf{f} \circ R_2 = \iota_{\{0\}}$ . Therefore,

$$\partial \mathbf{f}(1, 0) = \mathbb{R}_+ \times \{0\} \neq \mathbb{R}_+ \times \mathbb{R} = (\partial(\mathbf{f} \circ R_1)(1)) \times (\partial(\mathbf{f} \circ R_2)(0)). \quad (16.6)$$

**Proposition 16.8** *Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces and, for every  $i \in I$ , let  $f_i: \mathcal{H}_i \rightarrow ]-\infty, +\infty]$  be proper. Then*

$$\partial \bigoplus_{i \in I} f_i = \bigcap_{i \in I} \partial f_i. \quad (16.7)$$

*Proof.* We denote by  $\mathbf{x} = (x_i)_{i \in I}$  a generic element in  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ . Set  $\mathbf{f} = \bigoplus_{i \in I} f_i$ , and take  $\mathbf{x}$  and  $\mathbf{u}$  in  $\mathcal{H}$ . Then Proposition 16.3(i) yields

$$\mathbf{u} \in \bigcap_{i \in I} \partial f_i(x_i) \Leftrightarrow (\forall i \in I)(\forall y_i \in \mathcal{H}_i) \langle y_i - x_i \mid u_i \rangle + f_i(x_i) \leq f_i(y_i) \quad (16.8)$$

$$\Rightarrow (\forall \mathbf{y} \in \mathcal{H}) \sum_{i \in I} \langle y_i - x_i \mid u_i \rangle + \sum_{i \in I} f_i(x_i) \leq \sum_{i \in I} f_i(y_i) \quad (16.9)$$

$$\begin{aligned} &\Leftrightarrow (\forall \mathbf{y} \in \mathcal{H}) \langle \mathbf{y} - \mathbf{x} \mid \mathbf{u} \rangle + \mathbf{f}(\mathbf{x}) \leq \mathbf{f}(\mathbf{y}) \\ &\Leftrightarrow \mathbf{u} \in \partial \mathbf{f}(\mathbf{x}). \end{aligned} \quad (16.10)$$

Finally, to obtain (16.9)  $\Rightarrow$  (16.8), fix  $i \in I$ . By forcing the coordinates of  $\mathbf{y}$  in (16.9) to coincide with those of  $\mathbf{x}$ , except for the  $i$ th, we obtain  $(\forall y_i \in \mathcal{H}_i) \langle y_i - x_i \mid u_i \rangle + f_i(x_i) \leq f_i(y_i)$ .  $\square$

The next result states that the graph of the subdifferential contains precisely those points for which the Fenchel–Young inequality (Proposition 13.13) becomes an equality.

**Proposition 16.9** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $x \in \mathcal{H}$ , and let  $u \in \mathcal{H}$ . Then  $u \in \partial f(x) \Leftrightarrow f(x) + f^*(u) = \langle x \mid u \rangle \Rightarrow x \in \partial f^*(u)$ .*

*Proof.* Using (16.1) and Proposition 13.13, we get

$$\begin{aligned} u \in \partial f(x) &\Leftrightarrow (\forall y \in \text{dom } f) \langle y \mid u \rangle - f(y) \leq \langle x \mid u \rangle - f(x) \leq f^*(u) \\ &\Leftrightarrow f^*(u) = \sup_{y \in \text{dom } f} (\langle y \mid u \rangle - f(y)) \leq \langle x \mid u \rangle - f(x) \leq f^*(u) \\ &\Leftrightarrow f(x) + f^*(u) = \langle x \mid u \rangle. \end{aligned} \quad (16.11)$$

Accordingly, using the Fenchel–Young inequality (Proposition 13.13) and Proposition 13.14(i), we obtain  $u \in \partial f(x) \Rightarrow \langle u \mid x \rangle \leq f^*(u) + f^{**}(x) \leq f^*(u) + f(x) = \langle u \mid x \rangle \Rightarrow f^*(u) + f^{**}(x) = \langle u \mid x \rangle \Rightarrow x \in \partial f^*(u)$ .  $\square$

**Corollary 16.10** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $\text{gra } \partial f \subset \text{dom } f \times \text{dom } f^*$ .*

**Example 16.11** Set  $f = (1/2)\|\cdot\|^2$ . Then  $\partial f = \text{Id}$ .

*Proof.* This follows from Proposition 16.9 and Proposition 13.16.  $\square$

## 16.2 Convex Functions

We start with a key example.

**Example 16.12** Let  $C$  be a nonempty convex subset of  $\mathcal{H}$ . Then  $\partial\iota_C = N_C$ .

*Proof.* Combine (16.1) and (6.31).  $\square$

Here is a refinement of Proposition 16.9.

**Proposition 16.13** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \mathcal{H}$ , and let  $u \in \mathcal{H}$ . Then  $u \in \partial f(x) \Leftrightarrow (u, -1) \in N_{\text{epi } f}(x, f(x)) \Leftrightarrow f(x) + f^*(u) = \langle x | u \rangle \Rightarrow x \in \partial f^*(u)$ .

*Proof.* Note that  $\text{epi } f$  is nonempty and convex. Moreover,  $(u, -1) \in N_{\text{epi } f}(x, f(x)) \Leftrightarrow x \in \text{dom } f$  and  $(\forall (y, \eta) \in \text{epi } f) \langle (y, \eta) - (x, f(x)) | (u, -1) \rangle \leq 0 \Leftrightarrow x \in \text{dom } f$  and  $(\forall (y, \eta) \in \text{epi } f) \langle y - x | u \rangle + (\eta - f(x))(-1) \leq 0 \Leftrightarrow (\forall (y, \eta) \in \text{epi } f) \langle y - x | u \rangle + f(x) \leq \eta \Leftrightarrow (\forall y \in \text{dom } f) \langle y - x | u \rangle + f(x) \leq f(y) \Leftrightarrow u \in \partial f(x)$ . In view of Proposition 16.9, the proof is complete.  $\square$

Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. We have observed in Proposition 16.3(i)&(iii) that  $\text{dom } \partial f \subset \text{dom } f$  and that, for every  $x \in \text{dom } f$ ,  $\partial f(x)$  is closed and convex. Convexity of  $f$  supplies stronger statements.

**Proposition 16.14** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $x \in \text{dom } f$ . Then the following hold:

- (i) Suppose that  $\text{int dom } f \neq \emptyset$  and that  $x \in \text{bdry dom } f$ . Then  $\partial f(x)$  is either empty or unbounded.
- (ii) Suppose that  $x \in \text{cont } f$ . Then  $\partial f(x)$  is nonempty and weakly compact.
- (iii) Suppose that  $x \in \text{cont } f$ . Then there exists  $\delta \in \mathbb{R}_{++}$  such that  $\partial f(B(x; \delta))$  is bounded.
- (iv) Suppose that  $\text{cont } f \neq \emptyset$ . Then  $\text{int dom } f \subset \text{dom } \partial f$ .

*Proof.* (i): By Proposition 7.5,  $x$  is a support point of  $\text{dom } f$ . Consequently, there exists  $u \in \mathcal{H} \setminus \{0\}$  such that  $(\forall y \in \text{dom } f) \langle y - x | u \rangle \leq 0$ . Hence  $(\forall v \in \partial f(x)) (\forall \lambda \in \mathbb{R}_+) v + \lambda u \in \partial f(x)$ .

(ii)&(iii): Observe that  $\text{epi } f$  is nonempty and convex. Also, by Proposition 8.36,  $\text{int epi } f \neq \emptyset$ . For every  $\varepsilon \in \mathbb{R}_{++}$ ,  $(x, f(x) - \varepsilon) \notin \text{epi } f$  and therefore  $(x, f(x)) \in \text{bdry epi } f$ . Using Proposition 7.5, we get  $(u, \nu) \in N_{\text{epi } f}(x, f(x)) \setminus \{(0, 0)\}$ . For every  $y \in \text{dom } f$  and every  $\eta \in \mathbb{R}_+$ , we have  $\langle (y, f(y) + \eta) - (x, f(x)) | (u, \nu) \rangle \leq 0$  and therefore

$$\langle y - x | u \rangle + (f(y) - f(x))\nu + \eta\nu \leq 0. \quad (16.12)$$

We first note that  $\nu \leq 0$  since, otherwise, we get a contradiction in (16.12) by letting  $\eta \rightarrow +\infty$ . To show that  $\nu < 0$ , let us argue by contradiction. If  $\nu = 0$ , then (16.12) yields  $\sup \langle \text{dom } f - x | u \rangle \leq 0$  and, since  $B(x; \varepsilon) \subset \text{dom } f$  for

$\varepsilon \in \mathbb{R}_{++}$  small enough, we would further deduce that  $\sup \langle B(0; \varepsilon) \mid u \rangle \leq 0$ . This would imply that  $u = 0$  and, in turn, that  $(u, \nu) = (0, 0)$ . Hence  $\nu < 0$ . Since  $N_{\text{epi } f}(x, f(x))$  is a cone, we also have

$$(u/|\nu|, -1) = (1/|\nu|)(u, \nu) \in N_{\text{epi } f}(x, f(x)). \quad (16.13)$$

Using Proposition 16.13, we obtain  $u/|\nu| \in \partial f(x)$ . Hence  $\partial f(x) \neq \emptyset$ . By Theorem 8.29, there exist  $\beta \in \mathbb{R}_{++}$  and  $\delta \in \mathbb{R}_{++}$  such that  $f$  is Lipschitz continuous with constant  $\beta$  relative to  $B(x; 2\delta)$ . Now take  $y \in B(x; \delta)$  and  $v \in \partial f(y)$ . Then  $(\forall z \in B(0; \delta)) \langle z \mid v \rangle \leq f(y + z) - f(y) \leq \beta \|z\|$  and hence  $\|v\| \leq \beta$ . It follows that  $\partial f(x) \subset \partial f(B(x; \delta)) \subset B(0; \beta)$ . Thus,  $\partial f(x)$  is bounded and, by Proposition 16.3(iii), closed and convex. It is therefore weakly compact by Theorem 3.33.

(iv): A consequence of (ii) and Corollary 8.30(i).  $\square$

**Corollary 16.15** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Suppose that  $\mathcal{H}$  is finite-dimensional and let  $x \in \text{ri dom } f$ . Then  $\partial f(x)$  is nonempty.*

*Proof.* This follows from Corollary 8.32 and Proposition 16.14(ii).  $\square$

**Corollary 16.16** *Let  $h: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and suppose that  $D$  is a nonempty open convex subset of  $\text{cont } h$ . Then  $(h + \iota_D)^{**}$  is the unique function  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  that satisfies*

$$f \in \Gamma_0(\mathcal{H}), \quad \text{dom } f \subset \overline{D}, \quad \text{and} \quad f|_D = h|_D. \quad (16.14)$$

Moreover,

$$(\forall x \in \mathcal{H})(\forall y \in D) \quad f(x) = \begin{cases} \lim_{\alpha \downarrow 0} h((1 - \alpha)x + \alpha y), & \text{if } x \in \overline{D}; \\ +\infty, & \text{if } x \notin \overline{D}. \end{cases} \quad (16.15)$$

*Proof.* Set  $g = h + \iota_D$ . Then  $g$  is proper and convex, and  $D \subset \text{cont } g$ . Using Proposition 16.14(ii) and Proposition 16.4, we see that  $D \subset \text{dom } \partial g$  and that

$$h|_D = g|_D = g^{**}|_D. \quad (16.16)$$

Thus  $g^{**} \in \Gamma_0(\mathcal{H})$ . Since  $\text{dom } \partial g \neq \emptyset$ , Proposition 13.40(i) implies that  $\text{dom } g^{**} \subset \overline{\text{dom } g} \subset \overline{D}$ . Hence  $g^{**}$  satisfies (16.14). Now assume that  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  satisfies (16.14), fix  $y \in D = \text{int dom } f$ , and take  $x \in \mathcal{H}$ . If  $x \notin \overline{D}$ , then  $f(x) = +\infty$ . So assume that  $x \in \overline{D}$ . Proposition 3.35 implies that  $[x, y] \subset D$ . In view of Proposition 9.14 and (16.14), we deduce that

$$f(x) = \lim_{\alpha \downarrow 0} f((1 - \alpha)x + \alpha y) = \lim_{\alpha \downarrow 0} h((1 - \alpha)x + \alpha y). \quad (16.17)$$

This verifies the uniqueness of  $f$  as well as (16.15).  $\square$

**Proposition 16.17** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be lower semicontinuous and convex. Then the following are equivalent:*

- (i)  $f$  is bounded on every bounded subset of  $\mathcal{H}$ .
- (ii)  $f$  is Lipschitz continuous relative to every bounded subset of  $\mathcal{H}$ .
- (iii)  $\text{dom } \partial f = \mathcal{H}$  and  $\partial f$  maps every bounded subset of  $\mathcal{H}$  to a bounded set.
- (iv)  $f^*$  is supercoercive.

If  $\mathcal{H}$  is finite-dimensional, then  $f$  satisfies these properties.

*Proof.* (i) $\Rightarrow$ (ii): Take a bounded subset  $C$  of  $\mathcal{H}$ . Let  $x_0 \in \mathcal{H}$  and  $\rho \in \mathbb{R}_{++}$  be such that  $C \subset B(x_0; \rho)$ . By assumption,  $f$  is bounded on  $B(x_0; 2\rho)$ . Proposition 8.28(ii) implies that  $f$  is Lipschitz continuous relative to  $B(x_0; \rho)$ , and hence relative to  $C$ .

(ii) $\Rightarrow$ (iii): Proposition 16.14(ii) asserts that  $\text{dom } \partial f = \mathcal{H}$ . It suffices to show that the subgradients of  $f$  are uniformly bounded on every open ball centered at 0. To this end, fix  $\rho \in \mathbb{R}_{++}$ , let  $\lambda \in \mathbb{R}_+$  be a Lipschitz constant of  $f$  relative to  $\text{int } B(0; \rho)$ , take  $x \in \text{int } B(0; \rho)$ , and let  $\alpha \in \mathbb{R}_{++}$  be such that  $B(x; \alpha) \subset \text{int } B(0; \rho)$ . Now suppose that  $u \in \partial f(x)$ . Then

$$(\forall y \in B(0; 1)) \quad \langle \alpha y \mid u \rangle \leq f(x + \alpha y) - f(x) \leq \lambda \|\alpha y\| \leq \lambda \alpha. \quad (16.18)$$

Taking the supremum over  $y \in B(0; 1)$  in (16.18), we obtain  $\alpha \|u\| \leq \lambda \alpha$ , i.e.,  $\|u\| \leq \lambda$ . Thus,  $\sup \|\partial f(\text{int } B(0; \rho))\| \leq \lambda$ .

(iii) $\Rightarrow$ (i): It suffices to show that  $f$  is bounded on every closed ball centered at 0. Fix  $\rho \in \mathbb{R}_{++}$ . Then  $\beta = \sup \|\partial f(B(0; \rho))\| < +\infty$ . Now take  $x \in B(0; \rho)$  and  $u \in \partial f(x)$ . Then  $\langle 0 - x \mid u \rangle + f(x) \leq f(0)$  and hence  $f(x) \leq f(0) + \langle x \mid u \rangle \leq f(0) + \rho \beta$ . It follows that

$$\sup f(B(0; \rho)) \leq f(0) + \rho \beta. \quad (16.19)$$

Now take  $v \in \partial f(0)$ . Then  $\langle x - 0 \mid v \rangle + f(0) \leq f(x)$  and thus  $f(x) \geq f(0) - \|x\| \|v\| \geq f(0) - \rho \beta$ . We deduce that

$$\inf f(B(0; \rho)) \geq f(0) - \rho \beta. \quad (16.20)$$

Altogether, (16.19) and (16.20) imply that  $f$  is bounded on  $B(0; \rho)$ .

(i) $\Leftrightarrow$ (iv): This is an immediate consequence of Proposition 14.15 and Corollary 13.33.

Finally, suppose that  $\mathcal{H}$  is finite-dimensional. By Corollary 8.32,  $f$  satisfies (ii) and, in turn, the equivalent properties (i), (iii), and (iv).  $\square$

## 16.3 Lower Semicontinuous Convex Functions

Let us first revisit the properties of positively homogeneous convex functions.

**Proposition 16.18** *Let  $f \in \Gamma_0(\mathcal{H})$  be positively homogeneous. Then  $f = \sigma_C$ , where  $C = \partial f(0)$ .*

*Proof.* Since  $f(0) = 0$ , the result follows from Proposition 14.11 and Definition 16.1.  $\square$

A consequence of Corollary 16.16 is the following result, which shows that certain functions in  $\Gamma_0(\mathcal{H})$  are uniquely determined by their behavior on the interior of their domain.

**Proposition 16.19** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{int dom } f \neq \emptyset$ . Then*

$$f = (f + \iota_{\text{int dom } f})^{**}. \quad (16.21)$$

The next example shows that in Proposition 16.19  $\text{int dom } f$  cannot be replaced in general by a dense convex subset of  $\text{dom } f$ . (However, as will be seen in Corollary 16.31,  $\text{dom } \partial f$  is large enough to reconstruct  $f$ .)

**Example 16.20** Suppose that  $\mathcal{H}$  is infinite-dimensional and separable. Let  $(e_n)_{n \in \mathbb{N}}$  be an orthonormal basis of  $\mathcal{H}$ , and let  $(\alpha_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1[$  such that  $(\alpha_n)_{n \in \mathbb{N}} \in \ell^2(\mathbb{N})$ . Define

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sup_{n \in \mathbb{N}} \frac{\langle x | e_n \rangle}{\alpha_n} \quad (16.22)$$

and set  $C = \text{span}\{e_n\}_{n \in \mathbb{N}}$ . Then  $f \in \Gamma_0(\mathcal{H})$  and  $C \subset \text{dom } f$ . In fact,  $C$  is a convex and dense subset of  $\mathcal{H}$  and  $0 \leq f + \iota_C$ . By Proposition 13.14(ii),  $0 = 0^{**} \leq (f + \iota_C)^{**}$ . On the other hand, set  $z = -\sum_{n \in \mathbb{N}} \alpha_n e_n$ . Then  $f(z) = -1$  and  $z \in (\text{dom } f) \setminus C$ . Altogether,  $(f + \iota_C)^{**} \neq f$ .

**Proposition 16.21** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{int dom } f = \text{cont } f \subset \text{dom } \partial f \subset \text{dom } f$ .*

*Proof.* The equality follows from Corollary 8.30(ii) and the first inclusion from Proposition 16.14(ii). The second inclusion was observed in Proposition 16.3(i).  $\square$

**Remark 16.22** Let  $f \in \Gamma_0(\mathcal{H})$ . Then Proposition 16.21 asserts that  $\text{dom } \partial f$  is sandwiched between the convex sets  $\text{int dom } f$  and  $\text{dom } f$ . However it may not be convex itself. For instance, set  $\mathcal{H} = \mathbb{R}^2$  and  $f: (\xi_1, \xi_2) \mapsto \max\{g(\xi_1), |\xi_2|\}$ , where  $g(\xi_1) = 1 - \sqrt{\xi_1}$  if  $\xi_1 \geq 0$ ;  $g(\xi_1) = +\infty$  if  $\xi_1 < 0$ . Then  $\text{dom } \partial f = (\mathbb{R}_+ \times \mathbb{R}) \setminus (\{0\} \times ]-1, 1])$ .

The following theorem provides characterizations of the subdifferential of a function in  $\Gamma_0(\mathcal{H})$ .

**Theorem 16.23** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \mathcal{H}$ , and let  $u \in \mathcal{H}$ . Then the following are equivalent:*

- (i)  $(x, u) \in \text{gra } \partial f$ .
- (ii)  $(u, -1) \in N_{\text{epi } f}(x, f(x))$ .
- (iii)  $f(x) + f^*(u) = \langle x | u \rangle$ .



(iv)  $(u, x) \in \text{gra } \partial f^*$ .

*Proof.* (i) $\Leftrightarrow$ (ii) $\Leftrightarrow$ (iii) $\Rightarrow$ (iv): Proposition 16.13.

(iv) $\Rightarrow$ (iii): The implication (i) $\Rightarrow$ (iii) and Corollary 13.33 yield  $(u, x) \in \text{gra } \partial f^* \Rightarrow \langle u \mid x \rangle = f^*(u) + f^{**}(x) = f^*(u) + f(x)$ .  $\square$

**Corollary 16.24** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $(\partial f)^{-1} = \partial f^*$ .*

**Example 16.25** Let  $x \in \mathcal{H}$ . Then

$$\partial \|\cdot\|(x) = \begin{cases} \{x/\|x\|\}, & \text{if } x \neq 0; \\ B(0; 1), & \text{if } x = 0. \end{cases} \quad (16.23)$$

*Proof.* Let  $f = \|\cdot\|$ . Then  $f \in \Gamma_0(\mathcal{H})$  and, as seen in Example 13.3(v),  $f^* = \iota_{B(0;1)}$ . Now let  $u \in \mathcal{H}$ . If  $x \neq 0$ , the equivalence (i) $\Leftrightarrow$ (iii) in Theorem 16.23, together with Cauchy–Schwarz, yields  $u \in \partial f(x) \Leftrightarrow [\|u\| \leq 1 \text{ and } \|x\| = \langle x \mid u \rangle] \Leftrightarrow [\|u\| \leq 1 \text{ and } \|x\| = \langle x \mid u \rangle \leq \|x\| \|u\| \leq \|x\|] \Leftrightarrow [\|u\| = 1 \text{ and } (\exists \alpha \in \mathbb{R}_{++}) u = \alpha x] \Leftrightarrow u = x/\|x\|$ . Finally, if  $x = 0$ , the equivalence (i) $\Leftrightarrow$ (iii) in Theorem 16.23 becomes  $u \in \partial f(x) \Leftrightarrow \iota_{B(0;1)}(u) = 0 \Leftrightarrow u \in B(0; 1)$ .  $\square$

**Proposition 16.26** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{gra } \partial f$  is sequentially closed in  $\mathcal{H}^{\text{weak}} \times \mathcal{H}^{\text{strong}}$  and in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ .*

*Proof.* Let  $(x_n, u_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } \partial f$  such that  $x_n \rightharpoonup x$  and  $u_n \rightarrow u$ . Then it follows from Theorem 16.23 that  $(\forall n \in \mathbb{N}) f(x_n) + f^*(u_n) = \langle x_n \mid u_n \rangle$ . Hence, we derive from Proposition 13.13, Theorem 9.1, and Lemma 2.41(iii) that

$$\begin{aligned} \langle x \mid u \rangle &\leq f(x) + f(u) \\ &\leq \underline{\lim} f(x_n) + \underline{\lim} f^*(u_n) \\ &\leq \underline{\lim} (f(x_n) + f^*(u_n)) \\ &= \lim \langle x_n \mid u_n \rangle \\ &= \langle x \mid u \rangle. \end{aligned} \quad (16.24)$$

Invoking Theorem 16.23 once more, we deduce that  $(x, u) \in \text{gra } \partial f$ , which proves the first assertion. Applying this result to  $f^*$  and then appealing to Corollary 16.24 yields the second assertion.  $\square$

**Proposition 16.27** *Let  $f \in \Gamma_0(\mathcal{H})$ , and let  $x, u_0$ , and  $u_1$  be in  $\mathcal{H}$ . Then the following hold:*

- (i) *Suppose that  $[u_0, u_1] \subset \partial f(x)$ . Then  $f^*$  is affine on  $[u_0, u_1]$ .*
- (ii) *Suppose that  $f^*$  is affine on  $[u_0, u_1]$  and that  $x \in \partial f^*([u_0, u_1])$ . Then  $[u_0, u_1] \subset \partial f(x)$ .*

*Proof.* Set  $(\forall \alpha \in ]0, 1[) u_\alpha = (1 - \alpha)u_0 + \alpha u_1$ .

(i): Theorem 16.23 yields  $f(x) + f^*(u_0) = \langle x \mid u_0 \rangle$  and  $f(x) + f^*(u_1) = \langle x \mid u_1 \rangle$ . Now take  $\alpha \in ]0, 1[$ . Then

$$\begin{aligned}
 f(x) + (1 - \alpha)f^*(u_0) + \alpha f^*(u_1) &= (1 - \alpha) \langle x \mid u_0 \rangle + \alpha \langle x \mid u_1 \rangle \\
 &= \langle x \mid u_\alpha \rangle \\
 &= f(x) + f^*(u_\alpha) \\
 &= f(x) + f^*((1 - \alpha)u_0 + \alpha u_1) \\
 &\leq f(x) + (1 - \alpha)f^*(u_0) + \alpha f^*(u_1),
 \end{aligned} \tag{16.25}$$

and we deduce that  $f^*$  is affine on  $[u_0, u_1]$ .

(ii): We have  $x \in \partial f^*(u_\alpha)$  for some  $\alpha \in ]0, 1[$ . Hence, Theorem 16.23 implies that

$$\begin{aligned}
 0 &= f(x) + f^*(u_\alpha) - \langle x \mid u_\alpha \rangle \\
 &= (1 - \alpha)(f(x) + f^*(u_0) - \langle x \mid u_0 \rangle) \\
 &\quad + \alpha(f(x) + f^*(u_1) - \langle x \mid u_1 \rangle).
 \end{aligned} \tag{16.26}$$

Since the terms  $f(x) + f^*(u_0) - \langle x \mid u_0 \rangle$  and  $f(x) + f^*(u_1) - \langle x \mid u_1 \rangle$  are positive by the Fenchel–Young inequality (Proposition 13.13), they must therefore be zero. Hence,  $\{u_0, u_1\} \subset \partial f(x)$  by Theorem 16.23, and we deduce from Proposition 16.3(iii) that  $[u_0, u_1] \subset \partial f(x)$ .  $\square$

**Proposition 16.28** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{gra}(f + \iota_{\text{dom } \partial f})$  is a dense subset of  $\text{gra } f$ ; in other words, for every  $x \in \text{dom } f$ , there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\text{dom } \partial f$  such that  $x_n \rightarrow x$  and  $f(x_n) \rightarrow f(x)$ .*

*Proof.* Take  $x \in \text{dom } f$  and  $\varepsilon \in \mathbb{R}_{++}$ , and set  $(p, \pi) = P_{\text{epi } f}(x, f(x) - \varepsilon)$ . Proposition 9.18 implies that  $\pi = f(p) > f(x) - \varepsilon$  and

$$(\forall y \in \text{dom } f) \quad \langle y - p \mid x - p \rangle \leq (f(p) - f(x) + \varepsilon)(f(y) - f(p)). \tag{16.27}$$

In view of (16.1), we deduce that  $(x - p)/(f(p) - f(x) + \varepsilon) \in \partial f(p)$ , hence  $p \in \text{dom } \partial f$ . In addition, (16.27) with  $y = x$  yields  $\|(x, f(x)) - (p, f(p))\|^2 = \|x - p\|^2 + |f(x) - f(p)|^2 \leq \varepsilon(f(x) - f(p)) < \varepsilon^2$ .  $\square$

**Corollary 16.29** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{dom } \partial f$  is a dense subset of  $\text{dom } f$ .*

**Corollary 16.30** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f^* = (f + \iota_{\text{dom } \partial f})^*$ .*

*Proof.* This follows from Proposition 16.28, Proposition 13.14(iv), and Proposition 13.9(v).  $\square$

In the light of Proposition 16.21, the following result can be viewed as a variant of Proposition 16.19.

**Corollary 16.31** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $f = (f + \iota_{\text{dom } \partial f})^{**}$ .*

*Proof.* Combine Corollary 16.30 and Corollary 13.33.  $\square$

## 16.4 Subdifferential Calculus

In this section we establish several rules for computing subdifferentials of transformations of convex functions in terms of the subdifferentials of these functions.

**Proposition 16.32** *Let  $\mathcal{K}$  be a real Hilbert space, let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $L(\text{dom } f) \cap \text{dom } g \neq \emptyset$ . Suppose that  $(f + g \circ L)^* = f^* \square (L^* \triangleright g^*)$ . Then  $\partial(f + g \circ L) = \partial f + L^* \circ (\partial g) \circ L$ .*

*Proof.* In view of Proposition 16.5(ii), it remains to establish the inclusion  $\text{gra } \partial(f + g \circ L) \subset \text{gra } (\partial f + L^* \circ (\partial g) \circ L)$ . Take  $(x, u) \in \text{gra } \partial(f + g \circ L)$ . On the one hand, Proposition 16.9 forces

$$(f + g \circ L)(x) + (f + g \circ L)^*(u) = \langle x \mid u \rangle. \quad (16.28)$$

On the other hand, our assumption implies that there exists  $v \in \mathcal{K}$  such that  $(f + g \circ L)^*(u) = f^*(u - L^*v) + g^*(v)$ . Altogether,

$$(f(x) + f^*(u - L^*v) - \langle x \mid u - L^*v \rangle) + (g(Lx) + g^*(v) - \langle x \mid L^*v \rangle) = 0. \quad (16.29)$$

In view of Proposition 13.13, we obtain  $f(x) + f^*(u - L^*v) = \langle x \mid u - L^*v \rangle$  and  $g(Lx) + g^*(v) = \langle Lx \mid v \rangle$ . In turn, Proposition 16.9 yields  $u - L^*v \in \partial f(x)$  and  $v \in \partial g(Lx)$ , hence  $u \in \partial f(x) + L^*(\partial g(Lx))$ .  $\square$

**Example 16.33** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\partial(f + (\gamma/2)\|\cdot\|^2) = \partial f + \gamma \text{Id}$ .

*Proof.* Combine Proposition 16.32, Proposition 14.1, and Example 16.11.  $\square$

**Proposition 16.34** *Let  $f \in \Gamma_0(\mathcal{H})$ , and let  $x$  and  $p$  be in  $\mathcal{H}$ . Then*

$$p = \text{Prox}_f x \quad \Leftrightarrow \quad x - p \in \partial f(p). \quad (16.30)$$

*In other words,*

$$\text{Prox}_f = (\text{Id} + \partial f)^{-1}. \quad (16.31)$$

*Proof.* We derive (16.30) from Proposition 12.26 and (16.1). Alternatively, it follows from Definition 12.23, Theorem 16.2, and Example 16.33 that  $p = \text{Prox}_f x \Leftrightarrow 0 \in \partial(f + (1/2)\|x - \cdot\|^2)(p) = \partial f(p) + p - x$ .  $\square$

**Proposition 16.35** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\text{ran}(\text{Id} + \partial f) = \mathcal{H}$ .*

*Proof.* We deduce from (16.31) and Definition 12.23 that  $\text{ran}(\text{Id} + \partial f) = \text{dom } \text{Prox}_f = \mathcal{H}$ .  $\square$

**Remark 16.36** Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$  and suppose that  $\text{dom } f \cap \text{dom } g \neq \emptyset$ . Proposition 16.32 yields  $(f + g)^* = f^* \square g^* \Rightarrow \partial(f + g) = \partial f + \partial g$ , i.e.,

$$f^* \square g^* \text{ is exact on } \text{dom}(f + g)^* \Rightarrow \partial(f + g) = \partial f + \partial g. \quad (16.32)$$

Almost conversely, one has (see Exercise 16.8)

$$\partial(f + g) = \partial f + \partial g \Rightarrow f^* \square g^* \text{ is exact on } \text{dom } \partial(f + g)^*, \quad (16.33)$$

which raises the question whether the implication  $\partial(f + g) = \partial f + \partial g \Rightarrow (f + g)^* = f^* \square g^*$  holds. An example constructed in [116] shows that the answer is negative.

**Theorem 16.37** *Let  $\mathcal{K}$  be a real Hilbert space, let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } g - L(\text{dom } f))$  (see Proposition 6.19 for special cases).
- (ii)  $\mathcal{K}$  is finite-dimensional,  $g$  is polyhedral, and  $\text{dom } g \cap \text{ri } L(\text{dom } f) \neq \emptyset$ .
- (iii)  $\mathcal{H}$  and  $\mathcal{K}$  are finite-dimensional,  $f$  and  $g$  are polyhedral, and  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ .

Then  $\partial(f + g \circ L) = \partial f + L^* \circ (\partial g) \circ L$ .

*Proof.* Combine Theorem 15.27 and Proposition 16.32.  $\square$

**Corollary 16.38** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$ .
- (ii)  $\text{dom } f \cap \text{int dom } g \neq \emptyset$ .
- (iii)  $\text{dom } g = \mathcal{H}$ .
- (iv)  $\mathcal{H}$  is finite-dimensional and  $\text{ri dom } f \cap \text{ri dom } g \neq \emptyset$ .

Then  $\partial(f + g) = \partial f + \partial g$ .

*Proof.* (i): Clear from Theorem 16.37(i).

(ii) $\Rightarrow$ (i): Proposition 6.19(vii).

(iii) $\Rightarrow$ (ii): Clear.

(iv) $\Rightarrow$ (i): Proposition 6.19(viii).  $\square$

**Corollary 16.39** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , and let  $(f_i)_{i \in I}$  be functions in  $\Gamma_0(\mathcal{H})$  such that one of the following holds:*

- (i) We have

$$0 \in \bigcap_{i=2}^m \text{sri} \left( \text{dom } f_i - \bigcap_{j=1}^{i-1} \text{dom } f_j \right). \quad (16.34)$$

- (ii) For every  $i \in \{2, \dots, m\}$ ,  $\text{dom } f_i - \bigcap_{j=1}^{i-1} \text{dom } f_j$  is a closed linear subspace.
- (iii) The sets  $(\text{dom } f_i)_{i \in I}$  are linear subspaces and, for every  $i \in \{2, \dots, m\}$ ,  $\text{dom } f_i + \bigcap_{j=1}^{i-1} \text{dom } f_j$  is closed.
- (iv)  $\text{dom } f_m \cap \bigcap_{i=1}^{m-1} \text{int dom } f_i \neq \emptyset$ .

(v)  $\mathcal{H}$  is finite-dimensional and  $\bigcap_{i \in I} \text{ri dom } f_i \neq \emptyset$ .

Then  $\partial(\sum_{i \in I} f_i) = \sum_{i \in I} \partial f_i$ .

*Proof.* (i): We proceed by induction. For  $m = 2$  and functions  $f_1$  and  $f_2$  in  $\Gamma_0(\mathcal{H})$ , (16.34) becomes  $0 \in \text{sri}(\text{dom } f_2 - \text{dom } f_1)$  and we derive from Corollary 16.38(i) that  $\partial(f_2 + f_1) = \partial f_2 + \partial f_1$ . Now suppose that the result is true for  $m \geq 2$  functions  $(f_i)_{1 \leq i \leq m}$  in  $\Gamma_0(\mathcal{H})$ , and let  $f_{m+1}$  be a function in  $\Gamma_0(\mathcal{H})$  such that

$$0 \in \bigcap_{i=2}^{m+1} \text{sri} \left( \text{dom } f_i - \bigcap_{j=1}^{i-1} \text{dom } f_j \right). \quad (16.35)$$

Then (16.34) holds and hence the induction hypothesis yields  $\partial(\sum_{i=1}^m f_i) = \sum_{i=1}^m \partial f_i$ . Moreover, it follows from (16.35) that

$$0 \in \text{sri} \left( \text{dom } f_{m+1} - \bigcap_{i=1}^m \text{dom } f_i \right) = \text{sri} \left( \text{dom } f_{m+1} - \text{dom } \sum_{i=1}^m f_i \right), \quad (16.36)$$

where  $\sum_{i=1}^m f_i \in \Gamma_0(\mathcal{H})$ . We therefore derive from Corollary 16.38(i) that

$$\begin{aligned} \partial \left( \sum_{i=1}^{m+1} f_i \right) &= \partial \left( f_{m+1} + \sum_{i=1}^m f_i \right) \\ &= \partial f_{m+1} + \partial \left( \sum_{i=1}^m f_i \right) \\ &= \partial f_{m+1} + \sum_{i=1}^m \partial f_i \\ &= \sum_{i=1}^{m+1} \partial f_i, \end{aligned} \quad (16.37)$$

which concludes the proof.

(ii) $\Rightarrow$ (i): Proposition 8.2 and Proposition 6.20(i).

(iii) $\Rightarrow$ (i): Proposition 6.20(ii).

(iv) $\Rightarrow$ (i): Proposition 8.2 and Proposition 6.20(iii).

(v) $\Rightarrow$ (i): Proposition 8.2 and Proposition 6.20(iv).  $\square$

The next example shows that the sum rule for subdifferentials in Corollary 16.38 fails if the domains of the functions merely intersect.

**Example 16.40** Suppose that  $\mathcal{H} = \mathbb{R}^2$ , and set  $C = B((-1, 0); 1)$  and  $D = B((1, 0); 1)$ . Then  $C \cap D = \text{dom } \iota_C \cap \text{dom } \iota_D = \{(0, 0)\}$  and  $\partial(\iota_C + \iota_D)(0, 0) = \mathbb{R}^2 \neq \mathbb{R} \times \{0\} = \partial \iota_C(0, 0) + \partial \iota_D(0, 0)$ .

It follows from Fermat's rule (Theorem 16.2) that a function that admits a minimizer is bounded below and 0 belongs to the range of its subdifferential

operator. In contrast, a function without a minimizer that is unbounded below—such as a nonzero continuous linear functional—cannot have 0 in the closure of the range of its subdifferential operator. However, the following result implies that functions in  $\Gamma_0(\mathcal{H})$  that are bounded below possess arbitrarily small subgradients.

**Corollary 16.41** *Let  $f \in \Gamma_0(\mathcal{H})$  be bounded below. Then  $0 \in \overline{\text{ran}} \partial f$ .*

*Proof.* Fix  $\varepsilon \in \mathbb{R}_{++}$ . Since  $f$  is bounded below and  $\|\cdot\|$  is coercive, Corollary 11.15(ii) implies that  $f + \varepsilon\|\cdot\|$  has a minimizer, say  $z$ . Using Theorem 16.2, Corollary 16.38(iii), and Example 16.25, we obtain  $0 \in \partial f(z) + \varepsilon B(0; 1)$ .  $\square$

Here is another consequence of Theorem 16.37.

**Corollary 16.42** *Let  $g \in \Gamma_0(\mathcal{K})$  and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that one of the following holds:*

- (i)  $0 \in \text{sri}(\text{dom } g - \text{ran } L)$ .
- (ii)  $\mathcal{K}$  is finite-dimensional,  $g$  is polyhedral, and  $\text{dom } g \cap \text{ran } L \neq \emptyset$ .

*Then  $\partial(g \circ L) = L^* \circ (\partial g) \circ L$ .*

**Example 16.43** Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\partial(f^\vee) = -(\partial f)^\vee$ .

**Example 16.44** Let  $N$  be a strictly positive integer, set  $I = \{1, \dots, N\}$ , let  $(u_i)_{i \in I} \in \mathcal{H}^N$ , let  $(\beta_i)_{i \in I} \in \mathbb{R}^N$ , and set  $C = \bigcap_{i \in I} \{x \in \mathcal{H} \mid \langle x \mid u_i \rangle \leq \beta_i\}$ . Suppose that  $z \in C$  and set  $J = \{i \in I \mid \langle z \mid u_i \rangle = \beta_i\}$ . Then  $N_C z = \sum_{j \in J} \mathbb{R}_+ u_j$ .

*Proof.* Set  $b = (\beta_i)_{i \in I}$ , set  $L: \mathcal{H} \rightarrow \mathbb{R}^N: x \mapsto (\langle x \mid u_i \rangle)_{i \in I}$ , and set  $g: \mathbb{R}^N \rightarrow ]-\infty, +\infty]: y \mapsto \iota_{\mathbb{R}_-^N}(y - b)$ . Then  $g$  is polyhedral,  $\iota_C = g \circ L$ , and  $Lz \in \text{dom } g \cap \text{ran } L$ . By Corollary 16.42(ii),  $N_C z = \partial \iota_C(z) = \partial(g \circ L)(z) = L^*(\partial g(Lz)) = L^*(N_{\mathbb{R}_-^N}(Lz - b))$ . Hence the result follows from Example 6.41(ii).  $\square$

The following central result can be viewed as another consequence of Corollary 16.38.

**Theorem 16.45 (Brøndsted–Rockafellar)** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $(y, v) \in \text{dom}(f \oplus f^*)$ , and let  $\lambda$  and  $\mu$  in  $\mathbb{R}_+$  satisfy  $f(y) + f^*(v) \leq \langle y \mid v \rangle + \lambda\mu$ . Then there exists  $(z, w) \in \text{gra } \partial f$  such that  $\|z - y\| \leq \lambda$  and  $\|w - v\| \leq \mu$ .*

*Proof.* Set  $\alpha = \lambda\mu$ . If  $\alpha = 0$ , then  $(y, v)$  has the required properties by Proposition 16.9. So assume that  $\alpha > 0$ , set  $\beta = \lambda$ , and set  $h = f - \langle \cdot \mid v \rangle$ . Then for every  $x \in \mathcal{H}$ , we have

$$h(x) = -(\langle x \mid v \rangle - f(x)) \geq -f^*(v) \geq f(y) - \langle y \mid v \rangle - \lambda\mu = h(y) - \alpha. \quad (16.38)$$

Hence  $h$  is bounded below and  $\alpha \geq h(y) - \inf h(\mathcal{H})$ . Theorem 1.45 yields a point  $z \in \mathcal{H}$  such that  $\|z - y\| \leq \beta = \lambda$  and  $z \in \text{Argmin}(h + (\alpha/\beta)\|\cdot - z\|)$ . In turn, Theorem 16.2, Corollary 16.38(iii), and Example 16.25 imply that

$$0 \in \partial(h + (\alpha/\beta)\|\cdot - z\|)(z) = \partial f(z) - v + \mu B(0; 1). \quad (16.39)$$

Thus, there exists  $w \in \partial f(z)$  such that  $v - w \in \mu B(0; 1)$ , i.e.,  $\|w - v\| \leq \mu$ .  $\square$

**Proposition 16.46** *Let  $\mathcal{K}$  be a real Hilbert space, let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ , and set*

$$f: \mathcal{H} \rightarrow [-\infty, +\infty] : x \mapsto \inf F(x, \mathcal{K}). \quad (16.40)$$

*Suppose that  $f$  is proper and that  $(x, y) \in \mathcal{H} \times \mathcal{K}$  satisfies  $f(x) = F(x, y)$ , and let  $u \in \mathcal{H}$ . Then  $u \in \partial f(x) \Leftrightarrow (u, 0) \in \partial F(x, y)$ .*

*Proof.* As seen in Proposition 13.28,  $f^*(u) = F^*(u, 0)$ . Hence, by Proposition 16.9,  $u \in \partial f(x) \Leftrightarrow f^*(u) = \langle x \mid u \rangle - f(x) \Leftrightarrow F^*(u, 0) = \langle x \mid u \rangle - F(x, y) \Leftrightarrow F^*(u, 0) = \langle (x, y) \mid (u, 0) \rangle - F(x, y) \Leftrightarrow (u, 0) \in \partial F(x, y)$ .  $\square$

**Proposition 16.47** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , suppose that  $y \in \text{dom}(L \triangleright f)$ , and let  $x \in \mathcal{H}$ . Suppose that  $Lx = y$ . Then the following hold:*

(i) *Suppose that  $(L \triangleright f)(y) = f(x)$ . Then*

$$\partial(L \triangleright f)(y) = (L^*)^{-1}(\partial f(x)). \quad (16.41)$$

(ii) *Suppose that  $(L^*)^{-1}(\partial f(x)) \neq \emptyset$ . Then  $(L \triangleright f)(y) = f(x)$ .*

*Proof.* Let  $v \in \mathcal{K}$ . It follows from Proposition 13.21(iv), Proposition 13.13, and Proposition 16.9 that

$$\begin{aligned} f(x) + (L \triangleright f)^*(v) &= \langle y \mid v \rangle \Leftrightarrow f(x) + f^*(L^*v) = \langle Lx \mid v \rangle \\ &\Leftrightarrow f(x) + f^*(L^*v) = \langle x \mid L^*v \rangle \\ &\Leftrightarrow L^*v \in \partial f(x). \end{aligned} \quad (16.42)$$

(i): Proposition 16.9 and Proposition 13.21(iv) imply that

$$\begin{aligned} v \in \partial(L \triangleright f)(y) &\Leftrightarrow (L \triangleright f)(y) + (L \triangleright f)^*(v) = \langle y \mid v \rangle \\ &\Leftrightarrow f(x) + f^*(L^*v) = \langle Lx \mid v \rangle. \end{aligned} \quad (16.43)$$

To obtain (16.41), combine (16.43) with (16.42).

(ii): Suppose that  $v \in (L^*)^{-1}(\partial f(x))$ . Proposition 13.13 and (16.42) yield  $\langle y \mid v \rangle \leq (L \triangleright f)(y) + (L \triangleright f)^*(v) \leq f(x) + (L \triangleright f)^*(v) = \langle y \mid v \rangle$ .  $\square$

We now derive the following important rule for the subdifferential of the infimal convolution.

**Proposition 16.48** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , let  $x \in \text{dom}(f \square g)$ , and let  $y \in \mathcal{H}$ . Then the following hold:*

(i) Suppose that  $(f \square g)(x) = f(y) + g(x - y)$ . Then

$$\partial(f \square g)(x) = \partial f(y) \cap \partial g(x - y). \quad (16.44)$$

(ii) Suppose that  $\partial f(y) \cap \partial g(x - y) \neq \emptyset$ . Then  $(f \square g)(x) = f(y) + g(x - y)$ .

*Proof.* Set  $L = \text{Id} \oplus \text{Id}$ . Then  $L \in \mathcal{B}(\mathcal{H} \times \mathcal{H}, \mathcal{H})$  and  $L^*: \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: u \mapsto (u, u)$ . Proposition 12.35 implies that  $L \triangleright (f \oplus g) = f \square g$  and hence that  $\text{dom}(L \triangleright (f \oplus g)) = \text{dom}(f \square g)$ . Thus,  $L(y, x - y) = x \in \text{dom}(L \triangleright (f \oplus g))$ .

(i): Since  $(L \triangleright (f \oplus g))(x) = (f \oplus g)(y, x - y)$ , Proposition 16.47(i) and Proposition 16.8 imply that  $\partial(f \square g)(x) = (L^*)^{-1}(\partial(f \oplus g)(y, x - y)) = (L^*)^{-1}(\partial f(y) \times \partial g(x - y)) = \partial f(y) \cap \partial g(x - y)$ .

(ii): The assumption that  $\partial f(y) \cap \partial g(x - y) \neq \emptyset$  and Proposition 16.47(ii) yield  $(f \square g)(x) = (L \triangleright (f \oplus g))(x) = (f \oplus g)(y, x - y) = f(y) + g(x - y)$ .  $\square$

**Example 16.49** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then

$$(\forall x \in \mathcal{H}) \quad \partial d_C(x) = \begin{cases} \left\{ \frac{x - P_C x}{d_C(x)} \right\}, & \text{if } x \notin C; \\ N_C x \cap B(0; 1), & \text{if } x \in \text{bdry } C; \\ \{0\}, & \text{if } x \in \text{int } C. \end{cases} \quad (16.45)$$

*Proof.* Set  $f = \iota_C$  and  $g = \|\cdot\|$ . Then it follows from Theorem 3.14 that  $d_C(x) = f(P_C x) + g(x - P_C x)$ . Therefore, Proposition 16.48(i) yields  $\partial d_C(x) = \partial(f \square g)(x) = \partial f(P_C x) \cap \partial g(x - P_C x) = N_C(P_C x) \cap (\partial \|\cdot\|)(x - P_C x)$ . Since  $x - P_C x \in N_C(P_C x)$  by Proposition 6.46, (16.45) follows from Example 16.25, Proposition 6.43, and Proposition 6.12(ii).  $\square$

Our next result concerns the subdifferential of integral functions.

**Proposition 16.50** Let  $(\Omega, \mathcal{F}, \mu)$  be a complete  $\sigma$ -finite measure space, let  $(H, \langle \cdot | \cdot \rangle_H)$  be a separable real Hilbert space, and let  $\varphi \in \Gamma_0(H)$ . Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); H)$  and that one of the following holds:

- (i)  $\mu(\Omega) < +\infty$ .
- (ii)  $\varphi \geq \varphi(0) = 0$ .

Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty] \\ x \mapsto \begin{cases} \int_{\Omega} \varphi(x(\omega)) \mu(d\omega), & \text{if } \varphi \circ x \in L^1((\Omega, \mathcal{F}, \mu); \mathbb{R}); \\ +\infty, & \text{otherwise.} \end{cases} \quad (16.46)$$

Then  $f \in \Gamma_0(\mathcal{H})$  and, for every  $x \in \text{dom } f$ ,

$$\partial f(x) = \{u \in \mathcal{H} \mid u(\omega) \in \partial \varphi(x(\omega)) \text{ } \mu\text{-a.e.}\}. \quad (16.47)$$



*Proof.* Take  $x$  and  $u$  in  $\mathcal{H}$ . It follows from Proposition 13.13 that

$$\varphi(x(\omega)) + \varphi^*(u(\omega)) - \langle x(\omega) | u(\omega) \rangle_{\mathbf{H}} \geq 0 \quad \mu\text{-a.e.} \quad (16.48)$$

On the other hand, Proposition 13.43(i) yields  $f \in \Gamma_0(\mathcal{H})$ . Hence, we derive from Proposition 13.43(ii) and Theorem 16.23 that

$$\begin{aligned} u \in \partial f(x) &\Leftrightarrow f(x) + f^*(u) - \langle x | u \rangle = 0 \\ &\Leftrightarrow \int_{\Omega} \left( \varphi(x(\omega)) + \varphi^*(u(\omega)) - \langle x(\omega) | u(\omega) \rangle_{\mathbf{H}} \right) \mu(d\omega) = 0 \\ &\Leftrightarrow \varphi(x(\omega)) + \varphi^*(u(\omega)) - \langle x(\omega) | u(\omega) \rangle_{\mathbf{H}} = 0 \quad \mu\text{-a.e.} \\ &\Leftrightarrow u(\omega) \in \partial\varphi(x(\omega)) \quad \mu\text{-a.e.}, \end{aligned} \quad (16.49)$$

which provides (16.47).  $\square$

**Example 16.51** Let  $(\Omega, \mathcal{F}, \mu)$  be a complete  $\sigma$ -finite measure space and let  $(\mathbf{H}, \|\cdot\|_{\mathbf{H}})$  be a separable real Hilbert space. Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); \mathbf{H})$ . Set

$$\begin{aligned} f: \mathcal{H} &\rightarrow ]-\infty, +\infty] \\ x &\mapsto \begin{cases} \int_{\Omega} \|x(\omega)\|_{\mathbf{H}} \mu(d\omega), & \text{if } x \in L^1((\Omega, \mathcal{F}, \mu); \mathbf{H}); \\ +\infty, & \text{otherwise,} \end{cases} \end{aligned} \quad (16.50)$$

and define

$$(\forall \omega \in \Omega) \quad u(\omega) = \begin{cases} x(\omega)/\|x(\omega)\|_{\mathbf{H}}, & \text{if } x(\omega) \neq 0; \\ 0, & \text{otherwise.} \end{cases} \quad (16.51)$$

Then  $u \in \partial f(x)$ .

*Proof.* Apply Proposition 16.50 with  $\varphi = \|\cdot\|_{\mathbf{H}}$  and use Example 16.25.  $\square$

**Proposition 16.52** Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$  be autoconjugate and let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then the following are equivalent:

- (i)  $F(x, u) = \langle x | u \rangle$ .
- (ii)  $F^*(u, x) = \langle x | u \rangle$ .
- (iii)  $(u, x) \in \partial F(x, u)$ .
- (iv)  $(x, u) = \text{Prox}_F(x + u, x + u)$ .

*Proof.* (i)  $\Leftrightarrow F^{\top}(u, x) = \langle x | u \rangle \Leftrightarrow F^*(u, x) = \langle x | u \rangle \Leftrightarrow$  (ii). Hence, Proposition 13.31, (i), Proposition 16.9, and (16.31) yield (ii)  $\Leftrightarrow F(x, u) + F^*(u, x) = 2\langle x | u \rangle \Leftrightarrow F(x, u) + F^*(u, x) = \langle (x, u) | (u, x) \rangle \Leftrightarrow (u, x) \in \partial F(x, u) \Leftrightarrow$  (iii)  $\Leftrightarrow (x, u) + (u, x) \in (\text{Id} + \partial F)(x, u) \Leftrightarrow$  (iv).  $\square$

**Proposition 16.53** Set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: (x, u) \mapsto (u, x)$  and let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ . Then  $\text{Prox}_{F^{\top}} = \text{Id} - L \text{Prox}_F L$ .

*Proof.* Since  $L^* = L = L^{-1}$ , it follows that  $\text{Id} + \partial F^{*\top} = LL + L(\partial F^*)L = L(\text{Id} + \partial F^*)L$ . Hence  $\text{Prox}_{F^{*\top}} = (L(\text{Id} + \partial F^*)L)^{-1} = L^{-1}(\text{Id} + \partial F^*)^{-1}L^{-1} = L \text{Prox}_{F^*} L = L(\text{Id} - \text{Prox}_F)L = LL - L \text{Prox}_F L = \text{Id} - L \text{Prox}_F L$ .  $\square$

## Exercises

**Exercise 16.1** Provide a function  $f \in \Gamma_0(\mathbb{R})$  and a point  $x \in \text{dom } f$  such that  $\partial f(x) = \emptyset$ . Compare to Proposition 16.3(iv).

**Exercise 16.2** Define  $f: \mathbb{R} \times \mathbb{R} \rightarrow ]-\infty, +\infty]$  by

$$(x_1, x_2) \mapsto \begin{cases} x_1 x_2, & \text{if } x_1 > 0 \text{ and } x_2 > 0; \\ +\infty, & \text{otherwise.} \end{cases} \quad (16.52)$$

Show that (16.4) fails at every point  $(x_1, x_2) \in \mathbb{R}_{++} \times \mathbb{R}_{++}$ .

**Exercise 16.3** Provide  $f$  and  $g$  in  $\Gamma_0(\mathcal{H})$ ,  $x \in \text{dom}(f \square g)$ , and  $y \in \mathcal{H}$  such that  $(f \square g)(x) = f(y) + g(x - y)$  and  $\partial(f \square g)(x) = \emptyset$ .

**Exercise 16.4** Suppose that  $\mathcal{H} = \ell^2(\mathbb{N})$  and set  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x = (\xi_n)_{n \in \mathbb{N}} \mapsto \sum_{n \in \mathbb{N}} n \xi_n^{2n}$ . Prove the following:  $f \in \Gamma_0(\mathcal{H})$ ,  $\text{dom } f = \mathcal{H}$ , and  $f^*$  is not supercoercive.

**Exercise 16.5** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Suppose that  $\text{dom } f$  is convex and that  $\text{dom } \partial f = \text{dom } f$ . Show that  $f$  is convex.

**Exercise 16.6** Provide an example in which  $\partial f(x)$  is unbounded in Corollary 16.15.

**Exercise 16.7** Prove Proposition 16.19.

**Exercise 16.8** Prove the implication (16.33).

**Exercise 16.9** Use Corollary 16.29 to provide a different proof of Corollary 16.41.

**Exercise 16.10** Is the converse of Corollary 16.41 true, i.e., if  $f \in \Gamma_0(\mathcal{H})$  and  $0 \in \overline{\text{ran}} \partial f$ , must  $f$  be bounded below?

**Exercise 16.11** Use Theorem 16.45 to prove Corollary 16.29.

**Exercise 16.12** Let  $C$  be a convex subset of  $\mathcal{H}$  such that  $0 \in C$  and let  $x \in \mathcal{H}$ . Set  $y = x/m_C(x)$  and assume that  $y \in C$ . Let  $u \in \mathcal{H}$ . Show that  $u \in \partial m_C(x)$  if and only if  $u \in N_C y$  and  $\langle y | u \rangle = 1$ .

**Exercise 16.13** Let  $f \in \Gamma_0(\mathcal{H})$  be sublinear and set  $C = \partial f(0)$ . Use Proposition 14.11 to show that  $f = \sigma_C$  and that  $\text{epi } f^* = C \times \mathbb{R}_+$ .

**Exercise 16.14** Let  $f$  and  $g$  be two sublinear functions in  $\Gamma_0(\mathcal{H})$ . Use Exercise 12.2 and Exercise 15.1 to prove the equivalence  $(f + g)^* = f^* \square g^* \Leftrightarrow \partial(f + g)(0) = \partial f(0) + \partial g(0)$ .

# Chapter 17

## Differentiability of Convex Functions

Fréchet differentiability, Gâteaux differentiability, directional differentiability, subdifferentiability, and continuity are notions that are closely related to each other. In this chapter, we provide fundamental results on these relationships, as well as basic results on the steepest descent direction, the Chebyshev center, and the max formula that relates the directional derivative to the support function of the subdifferential at a given point.

### 17.1 Directional Derivatives

**Definition 17.1** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper, let  $x \in \text{dom } f$ , and let  $y \in \mathcal{H}$ . The *directional derivative* of  $f$  at  $x$  in the direction  $y$  is

$$f'(x; y) = \lim_{\alpha \downarrow 0} \frac{f(x + \alpha y) - f(x)}{\alpha}, \quad (17.1)$$

provided that this limit exists in  $[-\infty, +\infty]$ .

**Proposition 17.2** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \text{dom } f$ , and let  $y \in \mathcal{H}$ . Then the following hold:

- (i)  $\phi: \mathbb{R}_{++} \rightarrow ]-\infty, +\infty]: \alpha \mapsto (f(x + \alpha y) - f(x))/\alpha$  is increasing.
- (ii)  $f'(x; y)$  exists in  $[-\infty, +\infty]$  and

$$f'(x; y) = \inf_{\alpha \in \mathbb{R}_{++}} \frac{f(x + \alpha y) - f(x)}{\alpha}. \quad (17.2)$$

- (iii)  $f'(x; y - x) + f(x) \leq f(y)$ .
- (iv)  $f'(x; \cdot)$  is sublinear and  $f'(x; 0) = 0$ .
- (v)  $f'(x; \cdot)$  is proper, convex, and  $\text{dom } f'(x; \cdot) = \text{cone}(\text{dom } f - x)$ .
- (vi) Suppose that  $x \in \text{core dom } f$ . Then  $f'(x; \cdot)$  is real-valued and sublinear.

*Proof.* (i): Fix  $\alpha$  and  $\beta$  in  $\mathbb{R}_{++}$  such that  $\alpha < \beta$ , and set  $\lambda = \alpha/\beta$  and  $z = x + \beta y$ . If  $f(z) = +\infty$ , then certainly  $\phi(\alpha) \leq \phi(\beta) = +\infty$ . Otherwise, by (8.1),  $f(x + \alpha y) = f(\lambda z + (1 - \lambda)x) \leq \lambda f(z) + (1 - \lambda)f(x) = f(x) + \lambda(f(z) - f(x))$ ; hence  $\phi(\alpha) \leq \phi(\beta)$ .

(ii): An easy consequence of (i).

(iii): If  $y \notin \text{dom } f$ , the inequality is trivial. Otherwise, it follows from (8.1) that  $(\forall \alpha \in ]0, 1[) f((1 - \alpha)x + \alpha y) - f(x) \leq \alpha(f(y) - f(x))$ . Dividing by  $\alpha$  and letting  $\alpha \downarrow 0$ , we obtain  $f'(x; y - x) \leq f(y) - f(x)$ .

(iv): It is clear that  $f'(x; 0) = 0$  and that  $f'(x; \cdot)$  is positively homogeneous. Now take  $(y, \eta)$  and  $(z, \zeta)$  in  $\text{epi } f'(x; \cdot)$ ,  $\lambda \in ]0, 1[$ , and  $\varepsilon \in \mathbb{R}_{++}$ . Then, for all  $\alpha \in \mathbb{R}_{++}$  sufficiently small, we have  $(f(x + \alpha y) - f(x))/\alpha \leq \eta + \varepsilon$  and  $(f(x + \alpha z) - f(x))/\alpha \leq \zeta + \varepsilon$ . For such small  $\alpha$ , Corollary 8.10 yields

$$\begin{aligned} f(x + \alpha((1 - \lambda)y + \lambda z)) - f(x) &= f((1 - \lambda)(x + \alpha y) + \lambda(x + \alpha z)) - f(x) \\ &\leq (1 - \lambda)(f(x + \alpha y) - f(x)) + \lambda(f(x + \alpha z) - f(x)). \end{aligned} \quad (17.3)$$

Consequently,

$$\begin{aligned} \frac{f(x + \alpha((1 - \lambda)y + \lambda z)) - f(x)}{\alpha} &\leq (1 - \lambda) \frac{f(x + \alpha y) - f(x)}{\alpha} + \lambda \frac{f(x + \alpha z) - f(x)}{\alpha} \\ &\leq (1 - \lambda)(\eta + \varepsilon) + \lambda(\zeta + \varepsilon). \end{aligned} \quad (17.4)$$

Letting  $\alpha \downarrow 0$  and then  $\varepsilon \downarrow 0$ , we deduce that  $f'(x; (1 - \lambda)y + \lambda z) \leq (1 - \lambda)\eta + \lambda\zeta$ . Therefore,  $f'(x; \cdot)$  is convex.

(v): This follows from (ii) and (iv).

(vi): There exists  $\beta \in \mathbb{R}_{++}$  such that  $[x - \beta y, x + \beta y] \subset \text{dom } f$ . Now, take  $\alpha \in ]0, \beta]$ . Then, by (8.1),  $f(x) \leq (f(x - \alpha y) + f(x + \alpha y))/2$  and thus, appealing to (i), we obtain

$$\begin{aligned} -\left(\frac{f(x - \beta y) - f(x)}{\beta}\right) &\leq -\left(\frac{f(x - \alpha y) - f(x)}{\alpha}\right) \\ &= \frac{f(x) - f(x - \alpha y)}{\alpha} \\ &\leq \frac{f(x + \alpha y) - f(x)}{\alpha} \\ &\leq \frac{f(x + \beta y) - f(x)}{\beta}. \end{aligned} \quad (17.5)$$

Letting  $\alpha \downarrow 0$ , we deduce that

$$\frac{f(x) - f(x - \beta y)}{\beta} \leq -f'(x; -y) \leq f'(x; y) \leq \frac{f(x + \beta y) - f(x)}{\beta}. \quad (17.6)$$

Since the leftmost and the rightmost terms in (17.6) are in  $\mathbb{R}$ , so are the middle terms. Therefore,  $f'(x; \cdot)$  is real-valued on  $\mathcal{H}$ .  $\square$

**Proposition 17.3** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $x \in \text{dom } f$ . Then  $x \in \text{Argmin } f \Leftrightarrow f'(x; \cdot) \geq 0$ .*

*Proof.* Let  $y \in \mathcal{H}$ . We have  $x \in \text{Argmin } f \Rightarrow (\forall \alpha \in \mathbb{R}_{++}) (f(x + \alpha y) - f(x))/\alpha \geq 0 \Rightarrow f'(x; y) \geq 0$ . Conversely, suppose that  $f'(x; \cdot) \geq 0$ . Then, by Proposition 17.2(iii),  $f(x) \leq f'(x; y - x) + f(x) \leq f(y)$  and, therefore,  $x \in \text{Argmin } f$ .  $\square$

Let  $x \in \text{dom } f$  and suppose that  $f'(x; \cdot)$  is linear and continuous on  $\mathcal{H}$ . Then, by analogy with Definition 2.43 and Remark 2.44 (which concern real-valued functions, while  $f$  maps to  $]-\infty, +\infty]$ ),  $f$  is said to be Gâteaux differentiable at  $x$  and, by Riesz–Fréchet representation (Fact 2.17), there exists a unique vector  $\nabla f(x) \in \mathcal{H}$  such that

$$(\forall y \in \mathcal{H}) \quad f'(x; y) = \langle y \mid \nabla f(x) \rangle, \quad (17.7)$$

namely the Gâteaux gradient of  $f$  at  $x$ . Alternatively, as in (2.33),

$$(\forall y \in \mathcal{H}) \quad \langle y \mid \nabla f(x) \rangle = \lim_{\alpha \rightarrow 0} \frac{f(x + \alpha y) - f(x)}{\alpha}. \quad (17.8)$$

In a similar fashion, the Hessian of  $f$  at  $x$ , if it exists, is defined as in Remark 2.44. Furthermore, by analogy with Definition 2.45, if the convergence in (17.8) is uniform with respect to  $y$  on bounded sets, then  $\nabla f(x)$  is called the Fréchet gradient of  $f$  at  $x$ ; equivalently,

$$\lim_{0 \neq y \rightarrow 0} \frac{f(x + y) - f(x) - \langle y \mid \nabla f(x) \rangle}{\|y\|} = 0. \quad (17.9)$$

Lemma 2.49(ii) implies that, if  $f$  is Fréchet differentiable at  $x \in \text{int dom } f$ , then it is continuous at  $x$ .

**Proposition 17.4** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex, let  $x \in \mathcal{H}$ , and suppose that  $f$  is Gâteaux differentiable at  $x$ . Then  $x \in \text{Argmin } f \Leftrightarrow \nabla f(x) = 0$ .*

*Proof.* Proposition 17.3 and (17.7) yield  $x \in \text{Argmin } f \Leftrightarrow (\forall y \in \mathcal{H}) \langle y \mid \nabla f(x) \rangle = f'(x; y) \geq 0 \Leftrightarrow \nabla f(x) = 0$ .  $\square$

**Corollary 17.5** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Then the following are equivalent:*

- (i)  $x \in \text{Argmin } f$ .
- (ii)  $x \in \text{Argmin } \gamma f$ .
- (iii)  $x = \text{Prox}_{\gamma f} x$ .
- (iv)  $f(x) = \gamma f(x)$ .

*Proof.* (i) $\Leftrightarrow$ (iii): Proposition 12.28.

(ii) $\Leftrightarrow$ (iii): Proposition 12.29 and Proposition 17.4.

(iii) $\Leftrightarrow$ (iv): Remark 12.24 and Definition 12.23.  $\square$

**Corollary 17.6** *Let  $f \in \Gamma_0(\mathcal{H})$ , and let  $x$  and  $p$  be in  $\mathcal{H}$ . Suppose that  $\text{dom } f$  is open and that  $f$  is Gâteaux differentiable on  $\text{dom } f$ . Then  $p = \text{Prox}_f x \Leftrightarrow \nabla f(p) + p - x = 0$ .*

*Proof.* Set  $g: y \mapsto f(y) + (1/2)\|x - y\|^2$ . Then  $g$  is convex and Gâteaux differentiable on  $\text{dom } f$  with  $(\forall y \in \text{dom } f) \nabla g(y) = \nabla f(y) + y - x$ . Hence, the equivalence follows from Definition 12.23 and Proposition 17.4.  $\square$

**Example 17.7** Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $r \in \mathcal{K}$ , let  $\gamma \in \mathbb{R}_{++}$ , and set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \frac{1}{2}\|Lx - r\|^2. \quad (17.10)$$

Then it follows from Corollary 17.6 and Example 2.48 that  $(\forall x \in \mathcal{H}) \text{Prox}_{\gamma f} x = (\text{Id} + \gamma L^* L)^{-1}(x + \gamma L^* r)$ .

**Example 17.8** Let  $\gamma \in \mathbb{R}_{++}$  and set

$$f: \mathbb{R} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} -\gamma \ln(x), & \text{if } x > 0; \\ +\infty, & \text{if } x \leq 0. \end{cases} \quad (17.11)$$

Then Corollary 17.6 yields  $(\forall x \in \mathbb{R}) \text{Prox}_f x = (x + \sqrt{x^2 + 4\gamma})/2$ .

**Proposition 17.9** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex, let  $x \in \mathcal{H}$ , and suppose that  $f$  is Gâteaux differentiable at  $x$ . Then  $(\forall y \in \mathcal{H}) \langle y - x \mid \nabla f(x) \rangle + f(x) \leq f(y)$ .*

*Proof.* Combine Proposition 17.2(iii) and (17.7).  $\square$

## 17.2 Characterizations of Convexity

Convexity can be characterized in terms of first- and second-order differentiability properties.

**Proposition 17.10** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Suppose that  $\text{dom } f$  is open and convex, and that  $f$  is Gâteaux differentiable on  $\text{dom } f$ . Then the following are equivalent:*

- (i)  $f$  is convex.
- (ii)  $(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \langle x - y \mid \nabla f(y) \rangle + f(y) \leq f(x)$ .
- (iii)  $(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \langle x - y \mid \nabla f(x) - \nabla f(y) \rangle \geq 0$ . In other words,  $\nabla f$  is monotone.

Moreover, if  $f$  is twice Gâteaux differentiable on  $\text{dom } f$ , then each of the above is equivalent to

$$(iv) (\forall x \in \text{dom } f)(\forall z \in \mathcal{H}) \langle z \mid \nabla^2 f(x)z \rangle \geq 0.$$

*Proof.* Fix  $x \in \text{dom } f$ ,  $y \in \text{dom } f$ , and  $z \in \mathcal{H}$ . Since  $\text{dom } f$  is open, there exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $x + \varepsilon(x - y) \in \text{dom } f$  and  $y + \varepsilon(y - x) \in \text{dom } f$ . Furthermore, set  $C = ]-\varepsilon, 1 + \varepsilon[$  and

$$\phi: \mathbb{R} \rightarrow ]-\infty, +\infty]: \alpha \mapsto f(y + \alpha(x - y)) + \iota_C(\alpha). \quad (17.12)$$

Then  $\phi$  is Gâteaux differentiable on  $C$  and

$$(\forall \alpha \in C) \quad \phi'(\alpha) = \langle x - y \mid \nabla f(y + \alpha(x - y)) \rangle. \quad (17.13)$$

(i) $\Rightarrow$ (ii): Proposition 17.9.

(ii) $\Rightarrow$ (iii): It follows from (ii) that  $\langle x - y \mid \nabla f(y) \rangle + f(y) \leq f(x)$  and  $\langle y - x \mid \nabla f(x) \rangle + f(x) \leq f(y)$ . Adding up these two inequalities, we obtain  $\langle x - y \mid \nabla f(x) - \nabla f(y) \rangle \geq 0$ .

(iii) $\Rightarrow$ (i): Take  $\alpha$  and  $\beta$  in  $C$  such that  $\alpha < \beta$ , and set  $y_\alpha = y + \alpha(x - y)$  and  $y_\beta = y + \beta(x - y)$ . Then (iii) and (17.13) imply that  $\phi'(\beta) - \phi'(\alpha) = \langle y_\beta - y_\alpha \mid \nabla f(y_\beta) - \nabla f(y_\alpha) \rangle / (\beta - \alpha) \geq 0$ . Consequently,  $\phi'$  is increasing on  $C$  and  $\phi$  is therefore convex by Proposition 8.12(i). In particular,

$$f(\alpha x + (1 - \alpha)y) = \phi(\alpha) \leq \alpha\phi(1) + (1 - \alpha)\phi(0) = \alpha f(x) + (1 - \alpha)f(y). \quad (17.14)$$

(iii) $\Rightarrow$ (iv): Let  $z \in \mathcal{H}$ . Since  $\text{dom } f$  is open, for  $\alpha \in \mathbb{R}_{++}$  small enough,  $x + \alpha z \in \text{dom } f$ , and (iii) yields

$$\begin{aligned} \langle z \mid \nabla f(x + \alpha z) - \nabla f(x) \rangle &= \frac{1}{\alpha} \langle (x + \alpha z) - x \mid \nabla f(x + \alpha z) - \nabla f(x) \rangle \\ &\geq 0. \end{aligned} \quad (17.15)$$

In view of (2.32), dividing by  $\alpha$  and letting  $\alpha \downarrow 0$ , we obtain  $\langle z \mid \nabla^2 f(x)z \rangle \geq 0$ .

(iv) $\Rightarrow$ (i): Note that  $\phi$  is twice Gâteaux differentiable on  $C$  with  $(\forall \alpha \in C) \phi''(\alpha) = \langle x - y \mid \nabla^2 f(y + \alpha(x - y))(x - y) \rangle \geq 0$ . Hence,  $\phi'$  is increasing on  $C$  and, by Proposition 8.12(i),  $\phi$  is convex. We conclude with (17.14).  $\square$

**Example 17.11** Let  $A \in \mathcal{B}(\mathcal{H})$  and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \langle x \mid Ax \rangle$ . Then  $f$  is convex if and only if  $(\forall x \in \mathcal{H}) \langle x \mid (A + A^*)x \rangle \geq 0$ .

*Proof.* Combine Example 2.46 with Proposition 17.10(iv).  $\square$

**Proposition 17.12** Let  $h: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be such that  $D = \text{dom } h$  is nonempty, open, and convex. Suppose that  $h$  is Fréchet differentiable on  $D$  and that one of the following holds:

$$(i) (\forall x \in D)(\forall y \in D) \langle x - y \mid \nabla h(x) - \nabla h(y) \rangle \geq 0.$$

- (ii)  $h$  is twice Fréchet differentiable on  $D$  and  $(\forall x \in D)(\forall z \in \mathcal{H})$   
 $\langle z \mid \nabla^2 h(x)z \rangle \geq 0$ .

Take  $y \in D$  and set

$$f: x \mapsto \begin{cases} h(x), & \text{if } x \in D; \\ \lim_{\alpha \downarrow 0} h((1-\alpha)x + \alpha y), & \text{if } x \in \text{bdry } D; \\ +\infty, & \text{if } x \notin \overline{D}. \end{cases} \quad (17.16)$$

Then  $f \in \Gamma_0(\mathcal{H})$ ,  $D \subset \text{dom } f \subset \overline{D}$ , and  $f|_D = h|_D$ .

*Proof.* Proposition 17.10 guarantees that  $h$  is convex. Moreover, since  $h$  is Fréchet differentiable on  $D$ , it is continuous on  $D$  by Lemma 2.49(ii). Hence, the result follows from Corollary 16.16.  $\square$

### 17.3 Characterizations of Strict Convexity

Some of the results of Section 17.2 have counterparts for strictly convex functions. However, as Example 17.16 will illustrate, subtle differences exist. The proofs of the following results are left as Exercise 17.3, Exercise 17.6, and Exercise 17.9.

**Proposition 17.13** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Suppose that  $\text{dom } f$  is open and convex, and that  $f$  is Gâteaux differentiable on  $\text{dom } f$ . Then the following are equivalent:*

- (i)  $f$  is strictly convex.
- (ii)  $(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \ x \neq y \Rightarrow \langle x - y \mid \nabla f(y) \rangle + f(y) < f(x)$ .
- (iii)  $(\forall x \in \text{dom } f)(\forall y \in \text{dom } f) \ x \neq y \Rightarrow \langle x - y \mid \nabla f(x) - \nabla f(y) \rangle > 0$ .

Moreover, if  $f$  is twice Gâteaux differentiable on  $\text{dom } f$ , then each of the above is implied by

- (iv)  $(\forall x \in \text{dom } f)(\forall z \in \mathcal{H} \setminus \{0\}) \ \langle z \mid \nabla^2 f(x)z \rangle > 0$ .

**Example 17.14** Let  $A \in \mathcal{B}(\mathcal{H})$  and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \langle x \mid Ax \rangle$ . Then  $f$  is strictly convex if and only if  $A + A^*$  is strictly positive, i.e.,  $(\forall x \in \mathcal{H} \setminus \{0\}) \ \langle x \mid (A + A^*)x \rangle > 0$ .

*Proof.* Combine Example 2.46 with Proposition 17.13.  $\square$

Corollary 17.15, which complements Proposition 9.28, provides a convenient sufficient condition for strict convexity (see Exercise 17.8); extensions to  $\mathbb{R}^N$  can be obtained as described in Remark 9.31.



**Corollary 17.15** *Let  $h: \mathbb{R} \rightarrow ]-\infty, +\infty]$  be such that  $D = \text{dom } h$  is a nonempty open interval. Suppose that  $h$  is twice differentiable on  $D$  and such that, for every  $x \in D$ ,  $h''(x) > 0$ . Take  $y \in D$  and set*

$$f: x \mapsto \begin{cases} h(x), & \text{if } x \in D; \\ \lim_{\alpha \downarrow 0} h((1-\alpha)x + \alpha y), & \text{if } x \in \text{bdry } D; \\ +\infty, & \text{if } x \notin \overline{D}. \end{cases} \quad (17.17)$$

*Then  $f \in \Gamma_0(\mathbb{R})$ ,  $f|_D = h|_D$ ,  $D \subset \text{dom } f \subset \overline{D}$ , and  $f$  is strictly convex.*

Corollary 17.15 and the following example illustrate the absence of a strictly convex counterpart to Proposition 17.12, even in the Euclidean plane.

**Example 17.16** The function

$$h: \mathbb{R}^2 \rightarrow ]-\infty, +\infty]: (\xi, \eta) \mapsto \begin{cases} \eta^2 + \eta^2/\xi, & \text{if } \xi > 0 \text{ and } \eta > 0; \\ +\infty, & \text{otherwise,} \end{cases} \quad (17.18)$$

is twice Fréchet differentiable on its domain  $\mathbb{R}_{++}^2$ . Now let  $f$  be as in (17.16). Then

$$f: \mathbb{R}^2 \rightarrow ]-\infty, +\infty]: (\xi, \eta) \mapsto \begin{cases} \eta^2 + \eta^2/\xi, & \text{if } \xi > 0 \text{ and } \eta \geq 0; \\ 0, & \text{if } \xi = \eta = 0; \\ +\infty, & \text{otherwise,} \end{cases} \quad (17.19)$$

belongs to  $\Gamma_0(\mathbb{R}^2)$  but  $f$  is not strictly convex.

## 17.4 Directional Derivatives and Subgradients

**Proposition 17.17** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, suppose that  $x \in \text{dom } \partial f$ , and let  $u \in \mathcal{H}$ . Then the following hold:*

- (i)  $u \in \partial f(x) \Leftrightarrow \langle \cdot | u \rangle \leq f'(x; \cdot)$ .
- (ii)  $f'(x; \cdot)$  is proper and sublinear.

*Proof.* (i): Let  $\alpha \in \mathbb{R}_{++}$ . Then (16.1) yields  $u \in \partial f(x) \Rightarrow (\forall y \in \mathcal{H}) \langle y | u \rangle = \langle (x + \alpha y) - x | u \rangle / \alpha \leq (f(x + \alpha y) - f(x)) / \alpha$ . Taking the limit as  $\alpha \downarrow 0$  yields  $(\forall y \in \mathcal{H}) \langle y | u \rangle \leq f'(x; y)$ . Conversely, it follows from Proposition 17.2(iii) and (16.1) that  $(\forall y \in \mathcal{H}) \langle y - x | u \rangle \leq f'(x; y - x) \Rightarrow (\forall y \in \mathcal{H}) \langle y - x | u \rangle \leq f(y) - f(x) \Rightarrow u \in \partial f(x)$ .

(ii): Take  $u \in \partial f(x)$ . Then (i) yields  $f'(x; \cdot) \geq \langle \cdot | u \rangle$  and therefore  $-\infty \notin f'(x; \mathcal{H})$ . Hence, in view of Proposition 17.2(iv),  $f'(x; \cdot)$  is proper and sublinear.  $\square$

**Proposition 17.18** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $x \in \text{dom } f$ . Then  $(f'(x; \cdot))^* = \iota_{\partial f(x)}$ .*

*Proof.* Define  $\varphi: \mathcal{H} \rightarrow [-\infty, +\infty]: y \mapsto f'(x; y)$  and let  $u \in \mathcal{H}$ . Then, using Proposition 17.2(ii) and (13.1), we obtain

$$\begin{aligned} \varphi^*(u) &= \sup_{\alpha \in \mathbb{R}_{++}} \sup_{y \in \mathcal{H}} \left( \langle y | u \rangle - \frac{f(x + \alpha y) - f(x)}{\alpha} \right) \\ &= \sup_{\alpha \in \mathbb{R}_{++}} \frac{f(x) + \sup_{y \in \mathcal{H}} (\langle x + \alpha y | u \rangle - f(x + \alpha y)) - \langle x | u \rangle}{\alpha} \\ &= \sup_{\alpha \in \mathbb{R}_{++}} \frac{f(x) + f^*(u) - \langle x | u \rangle}{\alpha}. \end{aligned} \quad (17.20)$$

However, Proposition 16.9 asserts that  $u \in \partial f(x) \Leftrightarrow f(x) + f^*(u) - \langle x | u \rangle = 0$ . Hence,  $u \in \partial f(x) \Rightarrow \varphi^*(u) = 0$  and, moreover, Fenchel–Young (Proposition 13.13) yields  $u \notin \partial f(x) \Rightarrow f(x) + f^*(u) - \langle x | u \rangle > 0 \Rightarrow \varphi^*(u) = +\infty$ . Altogether,  $\varphi^* = \iota_{\partial f(x)}$ .  $\square$

**Theorem 17.19 (max formula)** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and suppose that  $x \in \text{cont } f$ . Then  $f'(x; \cdot) = \max \langle \cdot | \partial f(x) \rangle$ .*

*Proof.* Define  $\varphi: \mathcal{H} \rightarrow ]-\infty, +\infty]: y \mapsto f'(x; y)$  and fix  $y \in \mathcal{H}$ . Since  $x \in \text{cont } f$ , Theorem 8.29 asserts that there exist  $\rho \in \mathbb{R}_{++}$  and  $\beta \in \mathbb{R}_{++}$  such that  $f$  is  $\beta$ -Lipschitz continuous relative to  $B(x; \rho)$ . Hence, for  $\alpha \in \mathbb{R}_{++}$  small enough to ensure that  $x + \alpha y \in B(x; \rho)$ , we have  $|f(x + \alpha y) - f(x)|/\alpha \leq \beta \|y\|$ . Taking the limit as  $\alpha \downarrow 0$ , we obtain  $|\varphi(y)| \leq \beta \|y\|$ . Consequently,  $\varphi$  is locally bounded at every point in  $\mathcal{H}$  and  $\text{dom } \varphi = \mathcal{H}$ . Moreover,  $\varphi$  is convex by Proposition 17.2(v). We thus deduce from Corollary 8.30(i) that  $\varphi$  is continuous. Altogether,  $\varphi \in \Gamma_0(\mathcal{H})$  and it follows from Theorem 13.32 and Proposition 17.18 that

$$\varphi(y) = \varphi^{**}(y) = \sup_{u \in \text{dom } \varphi^*} (\langle u | y \rangle - \varphi^*(u)) = \sup_{u \in \partial f(x)} \langle y | u \rangle. \quad (17.21)$$

However,  $\partial f(x)$  is nonempty and weakly compact by Proposition 16.14(ii). On the other hand,  $\langle y | \cdot \rangle$  is weakly continuous. Hence, we derive from Theorem 1.28 that  $\varphi(y) = \max_{u \in \partial f(x)} \langle y | u \rangle$ .  $\square$

**Definition 17.20** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \text{dom } f$ , and let  $y \in \mathcal{H}$ . Then  $y$  is a *descent direction* of  $f$  at  $x$  if there exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $(\forall \alpha \in ]0, \varepsilon]) f(x + \alpha y) < f(x)$ .

**Proposition 17.21** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \text{dom } f$ , and let  $y \in \mathcal{H}$ . Then the following hold:*

- (i)  $y$  is a descent direction of  $f$  at  $x$  if and only if  $f'(x; y) < 0$ .

- (ii) Suppose that  $f$  is differentiable at  $x$  and that  $x \notin \operatorname{Argmin} f$ . Then  $-\nabla f(x)$  is a descent direction of  $f$  at  $x$ .

*Proof.* (i): By Proposition 17.2(i)&(ii),  $(f(x + \alpha y) - f(x))/\alpha \downarrow f'(x; y)$  as  $\alpha \downarrow 0$ .

(ii): By (17.7) and Proposition 17.4,  $f'(x; -\nabla f(x)) = -\|\nabla f(x)\|^2 < 0$ .  $\square$

**Proposition 17.22 (steepest descent direction)** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and suppose that  $x \in (\operatorname{cont} f) \setminus (\operatorname{Argmin} f)$ . Set  $u = P_{\partial f(x)} 0$  and  $z = -u/\|u\|$ . Then  $z$  is the unique minimizer of  $f'(x; \cdot)$  over  $B(0; 1)$ .*

*Proof.* Proposition 16.14(ii), Proposition 16.3(iii), and Theorem 16.2 imply that  $\partial f(x)$  is a nonempty closed convex set that does not contain 0. Hence  $u \neq 0$  and Theorem 3.14 yields  $\max \langle -u \mid \partial f(x) - u \rangle = 0$ , i.e.,  $\max \langle -u \mid \partial f(x) \rangle = -\|u\|^2$ . Using Theorem 17.19, we deduce that

$$f'(x; z) = \max \langle z \mid \partial f(x) \rangle = -\|u\| \quad (17.22)$$

and that

$$(\forall y \in B(0; 1)) \quad f'(x; y) = \max \langle y \mid \partial f(x) \rangle \geq \langle y \mid u \rangle \geq -\|u\|. \quad (17.23)$$

Therefore,  $z$  is a minimizer of  $f'(x; \cdot)$  over  $B(0; 1)$  and uniqueness follows from Fact 2.10.  $\square$

**Example 17.23** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and suppose that  $x \in \operatorname{dom} f \setminus \operatorname{Argmin} f$  and that  $u \in \partial f(x)$ . If  $f$  is not differentiable at  $x$ ,  $-u$  may not be a descent direction. For instance, take  $\mathcal{H} = \mathbb{R}^2$ , set  $f: (\xi_1, \xi_2) \mapsto |\xi_1| + 2|\xi_2|$ , set  $x = (1, 0)$ , and set  $u = (1, \pm\delta)$ , where  $\delta \in [1/2, 3/2]$ . Then one easily checks that  $u \in \partial f(x)$  and  $(\forall \alpha \in \mathbb{R}_{++}) f(x - \alpha u) \geq f(x)$ .

**Proposition 17.24** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Suppose that  $x \in \operatorname{cont} f$  and that  $C$  is a nonempty subset of  $\mathcal{H}$  such that  $f'(x; \cdot) = \sigma_C$ . Then  $\partial f(x) = \overline{\operatorname{conv}} C$ .*

*Proof.* In view of Theorem 17.19, we have  $\sigma_C = \sigma_{\partial f(x)}$ . Taking conjugates and using Example 13.37(i) and Proposition 16.3(iii), we obtain  $\iota_{\overline{\operatorname{conv}} C} = \sigma_C^* = \sigma_{\partial f(x)}^* = \iota_{\overline{\operatorname{conv}} \partial f(x)} = \iota_{\partial f(x)}$ .  $\square$

**Proposition 17.25 (Chebyshev center)** *Let  $C$  be a nonempty compact subset of  $\mathcal{H}$ , set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \max_{y \in C} \|x - y\|^2$ , and set*

$$\Phi_C: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \left\{ r \in C \mid \|x - r\| = \max_{y \in C} \|x - y\| \right\}. \quad (17.24)$$

*Then the following hold:*

- (i)  $f$  is continuous, strongly convex, and supercoercive.
- (ii)  $\text{dom } \Phi_C = \mathcal{H}$ ,  $\text{gra } \Phi_C$  is closed, and the sets  $(\Phi_C(x))_{x \in \mathcal{H}}$  are compact.
- (iii)  $(\forall x \in \mathcal{H})(\forall z \in \mathcal{H}) f'(x; z) = 2 \max \langle z \mid x - \Phi_C(x) \rangle$ .
- (iv)  $(\forall x \in \mathcal{H}) \partial f(x) = 2(x - \overline{\text{conv}} \Phi_C(x))$ .
- (v) The function  $f$  has a unique minimizer  $r$ , called the Chebyshev center of  $C$ , and characterized by

$$r \in \overline{\text{conv}} \Phi_C(r). \quad (17.25)$$

*Proof.* (i): Since the functions  $(\|\cdot - y\|^2)_{y \in C}$  are convex and lower semicontinuous, it follows from Proposition 9.3 that  $f$  is likewise. It follows from Corollary 8.30(ii) that  $f$  is continuous. Arguing similarly, we note that

$$g: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \max_{y \in C} (\|y\|^2 - 2 \langle x \mid y \rangle) \quad (17.26)$$

is convex and continuous, and that  $f = g + \|\cdot\|^2$  is strongly convex, hence supercoercive by Corollary 11.16.

(ii): It is clear that  $\text{dom } \Phi_C = \mathcal{H}$ . Now let  $(x_n, y_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } \Phi_C$  converging to  $(x, y) \in \mathcal{H} \times \mathcal{H}$ . Then  $y \in C$  by closedness of  $C$ . Since  $f$  is continuous, we have  $f(x) \leftarrow f(x_n) = \|x_n - y_n\|^2 \rightarrow \|x - y\|^2$ . Thus,  $y \in \Phi_C(x)$  and hence  $\text{gra } \Phi_C$  is closed. Therefore,  $\Phi_C(x)$  is compact.

(iii): Let  $x$  and  $z$  be in  $\mathcal{H}$ , let  $y \in \Phi_C(x)$ , and let  $t \in \mathbb{R}_{++}$ . Then  $f(x) = \|x - y\|^2$  and  $f(x + tz) \geq \|x + tz - y\|^2$ . Hence,  $(f(x + tz) - f(x))/t \geq t\|x - y\|^2 + 2 \langle z \mid x - y \rangle$ . This implies that  $f'(x; z) \geq 2 \langle z \mid x - y \rangle$  and, furthermore, that

$$f'(x; z) \geq 2 \max \langle z \mid x - \Phi_C(x) \rangle. \quad (17.27)$$

To establish the reverse inequality, let  $(t_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  such that  $t_n \rightarrow 0$ . For every  $n \in \mathbb{N}$ , set  $x_n = x + t_n z$ , and let  $r_n \in \Phi_C(x_n)$ . Then  $x_n \rightarrow x$ . Due to the compactness of  $C$ , after passing to a subsequence and relabeling if necessary, we assume that there exists  $r \in C$  such that  $r_n \rightarrow r$ . For every  $n \in \mathbb{N}$ , since  $f(x_n) = \|x_n - r_n\|^2$  and  $f(x) \geq \|x - r_n\|^2$ , we have  $(f(x_n) - f(x))/t_n \leq (\|x_n\|^2 - \|x\|^2)/t_n - 2 \langle z \mid r_n \rangle$ . Taking the limit as  $n \rightarrow +\infty$ , we deduce that

$$f'(x; z) \leq 2 \langle z \mid x - r \rangle \leq 2 \max \langle z \mid x - \Phi_C(x) \rangle. \quad (17.28)$$

Combining (17.27) and (17.28), we obtain (iii).

(iv): Combine (iii) and Proposition 17.24.

(v): In view of (i) and Corollary 11.16,  $f$  has a unique minimizer over  $\mathcal{H}$ , say  $r$ . Finally, Theorem 16.2 and (iv) yield the characterization (17.25).  $\square$

## 17.5 Gâteaux and Fréchet Differentiability

In this section, we explore the relationships between Gâteaux derivatives, Fréchet derivatives, and single-valued subdifferentials.

**Proposition 17.26** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, and let  $x \in \text{dom } f$ . Then the following hold:*

- (i) *Suppose that  $f$  is Gâteaux differentiable at  $x$ . Then  $\partial f(x) = \{\nabla f(x)\}$ .*
- (ii) *Suppose that  $x \in \text{cont } f$  and that  $\partial f(x)$  consists of a single element  $u$ . Then  $f$  is Gâteaux differentiable at  $x$  and  $u = \nabla f(x)$ .*

*Proof.* (i): It follows from (16.1) and Proposition 17.9 that  $\nabla f(x) \in \partial f(x)$ . Now let  $u \in \partial f(x)$ . By Proposition 17.17(i) and (17.7),  $\langle u - \nabla f(x) | u \rangle \leq f'(x; u - \nabla f(x)) = \langle u - \nabla f(x) | \nabla f(x) \rangle$ ; hence  $\|u - \nabla f(x)\|^2 \leq 0$ . Therefore  $u = \nabla f(x)$ . Altogether,  $\partial f(x) = \{\nabla f(x)\}$ .

(ii): Theorem 17.19 and (17.7) yield  $(\forall y \in \mathcal{H}) f'(x; y) = \langle y | u \rangle = \langle y | \nabla f(x) \rangle$ .  $\square$

**Proposition 17.27** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \mathcal{H}$ , and suppose that  $f$  is Gâteaux differentiable at  $x$ . Then  $f^*(\nabla f(x)) = \langle x | \nabla f(x) \rangle - f(x)$ .*

*Proof.* This follows from Proposition 16.9 and Proposition 17.26(i).  $\square$

**Proposition 17.28** *Let  $A \in \mathcal{B}(\mathcal{H})$  be positive and self-adjoint. Suppose that  $\text{ran } A$  is closed, set  $q_A: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2) \langle x | Ax \rangle$ , and define  $q_{A^\dagger}$  likewise. Then the following hold:*

- (i)  $q_A$  is convex, continuous, Fréchet differentiable, and  $\nabla q_A = A$ .
- (ii)  $q_{A^\dagger} = q_A \circ A^\dagger = \iota_{\ker A} \square q_A^*$ .
- (iii)  $q_A^* = \iota_{\text{ran } A} + q_{A^\dagger}$ .
- (iv)  $q_A^* \circ A = q_A$ .

*Proof.* (i): Clearly,  $\text{dom } q_A = \mathcal{H}$  and  $q_A$  is continuous. The convexity of  $q_A$  follows from Example 17.11. The Fréchet differentiability and the gradient formula were already observed in Example 2.46.

(ii)&(iii): Take  $u \in \mathcal{H}$ . Using Exercise 3.11 and Corollary 3.30(ii), we obtain  $2q_{A^\dagger}(u) = \langle u | A^\dagger u \rangle = \langle u | A^\dagger A A^\dagger u \rangle = \langle A^\dagger u | A A^\dagger u \rangle = \langle A^* u | A A^\dagger u \rangle = \langle A^\dagger u | A A^\dagger u \rangle = 2q_A(A^\dagger u)$ . Hence  $q_{A^\dagger} = q_A \circ A^\dagger$  is convex and continuous, and the first identity in (ii) holds.

Let us now verify (iii) since it will be utilized in the proof of the second identity in (ii). Set  $V = \ker A$ . Then it follows from Fact 2.18(iii) that  $V^\perp = \overline{\text{ran } A^*} = \text{ran } A$ . We assume first that  $u \notin \text{ran } A$ , i.e.,  $P_V u \neq 0$ . Then  $q_A^*(u) \geq \sup_{n \in \mathbb{N}} (\langle n P_V u | u \rangle - (1/2) \langle n P_V u | A(n P_V u) \rangle) = \sup_{n \in \mathbb{N}} n \|P_V u\|^2 = +\infty = (\iota_{\text{ran } A} + q_{A^\dagger})(u)$ , as required. Next, assume that  $u \in \text{ran } A$ , say  $u = Az = \nabla q_A(z)$ , where  $z \in \mathcal{H}$ . Since  $AA^\dagger = P_{\text{ran } A}$  by Proposition 3.28(ii), it follows from Proposition 17.27 that  $q_A^*(u) = \langle z | u \rangle - (1/2) \langle z | Az \rangle = (1/2) \langle z | u \rangle =$

$(1/2) \langle z \mid AA^\dagger u \rangle = (1/2) \langle Az \mid A^\dagger u \rangle = (1/2) \langle u \mid A^\dagger u \rangle = q_{A^\dagger}(u)$ . This completes the proof of (iii).

The fact that  $\text{dom}(q_A^*)^* = \text{dom } q_A = \mathcal{H}$ , Theorem 15.3, Proposition 3.28(v), Corollary 3.30(i), (i), and Corollary 13.33 imply that  $\iota_{\ker A} \square q_A^* = (\iota_{\text{ran } A} + q_A)^* = (\iota_{\text{ran } A^\dagger} + q_{A^\dagger}^*)^* = (q_{A^\dagger}^*)^* = q_{A^\dagger}$ .

(iv): In view of (iii), (ii), and Corollary 3.30(i), we have  $q_A^* \circ A = \iota_{\text{ran } A} \circ A + q_{A^\dagger} \circ A = q_{A^\dagger} \circ A^{\dagger\dagger} = q_{A^\dagger} = q_A$ .  $\square$

The following example extends Example 16.25, which corresponds to the case  $\phi = |\cdot|$ .

**Example 17.29** Set  $f = \phi \circ \|\cdot\|$ , where  $\phi: \mathbb{R} \rightarrow \mathbb{R}$  is convex, even, and differentiable on  $\mathbb{R} \setminus \{0\}$ . Then  $\partial\phi(0) = [-\rho, \rho]$  for some  $\rho \in \mathbb{R}_+$  and

$$(\forall x \in \mathcal{H}) \quad \partial f(x) = \begin{cases} \left\{ \phi'(\|x\|) \frac{x}{\|x\|} \right\}, & \text{if } x \neq 0; \\ B(0; \rho), & \text{if } x = 0. \end{cases} \quad (17.29)$$

*Proof.* Let  $x \in \mathcal{H}$ . We first suppose that  $x \neq 0$ . Then  $\phi$  is Fréchet differentiable at  $\|x\|$  (Exercise 2.10) and it follows from Fact 2.51 and Example 2.52 that  $f$  is Fréchet differentiable at  $x$  with  $\nabla f(x) = \phi'(\|x\|)(\|\cdot\|)'(x) = (\phi'(\|x\|)/\|x\|)x$ . Hence, Proposition 17.26(i) yields  $\partial f(x) = \{(\phi'(\|x\|)/\|x\|)x\}$ . Now suppose that  $x = 0$  and let  $u \in \mathcal{H}$ . Then we derive from Theorem 16.23 and Example 13.7 that  $u \in \partial f(x) \Leftrightarrow 0 \|u\| = \langle x \mid u \rangle = f(x) + f^*(u) = \phi(0) + \phi^*(\|u\|) \Leftrightarrow \|u\| \in \partial\phi(0)$ . However, since  $\phi$  is continuous by Corollary 8.31, Proposition 16.14(ii) implies that  $\partial\phi(0)$  is a closed bounded interval. Hence, since  $\phi$  is even,  $\partial\phi(0) = [-\rho, \rho]$  for some  $\rho \in \mathbb{R}_+$ . Thus,  $u \in \partial f(x) \Leftrightarrow \|u\| \leq \rho$ .  $\square$

The implication in Proposition 17.26(i) is not reversible.

**Example 17.30** Let  $C$  be the nonempty closed convex set of Example 6.11(iii) and let  $u \in \mathcal{H}$ . Then

$$0 \in C \subset \text{cone } C = \text{span } C \neq \overline{\text{cone } C} = \overline{\text{span } C} = \mathcal{H} \quad (17.30)$$

and  $u \in \partial\iota_C(0) \Leftrightarrow \sup \langle C \mid u \rangle \leq 0 \Leftrightarrow \sup \langle \overline{\text{cone } C} \mid u \rangle \leq 0 \Leftrightarrow \sup \langle \mathcal{H} \mid u \rangle \leq 0 \Leftrightarrow u = 0$ . Therefore,  $\partial\iota_C(0) = \{0\}$  and hence  $\iota_C$  possesses a unique subgradient at 0. On the other hand,  $0 \notin \text{core } C$  and thus  $\iota_C$  is not Gâteaux differentiable at 0.

**Proposition 17.31** Let  $f \in \Gamma_0(\mathcal{H})$  and suppose that  $x \in \text{int dom } f$ . Then the following are equivalent:

- (i)  $f$  is Gâteaux differentiable at  $x$ .
- (ii) Every selection of  $\partial f$  is strong-to-weak continuous at  $x$ .

*Proof.* (i) $\Rightarrow$ (ii): Let  $G$  be a selection of  $\partial f$  and let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{int dom } f$  converging to  $x$ . In view of Fact 1.37 and Lemma 2.23, it is

enough to show that  $Gx_n \rightarrow Gx$ . Since Proposition 16.14(iii) implies that  $(Gx_n)_{n \in \mathbb{N}}$  is bounded, let  $(Gx_{k_n})_{n \in \mathbb{N}}$  be a weakly convergent subsequence of  $(Gx_n)_{n \in \mathbb{N}}$ , say  $Gx_{k_n} \rightarrow u$ . Then  $(x_{k_n}, Gx_{k_n})_{n \in \mathbb{N}}$  lies in  $\text{gra } \partial f$  and we derive from Proposition 16.26 that  $(x, u) \in \text{gra } \partial f$ . By Proposition 17.26(i),  $u = \nabla f(x)$  and, using Lemma 2.38, we deduce that  $Gx_n \rightarrow \nabla f(x)$ .

(ii) $\Rightarrow$ (i): Let  $G$  be a strong-to-weak continuous selection of  $\partial f$  and fix  $y \in \mathcal{H}$ . Then there exists  $\beta \in \mathbb{R}_{++}$  such that  $[x, x + \beta y] \subset \text{int dom } f$ . Take  $\alpha \in ]0, \beta]$ . Then  $\alpha \langle y \mid Gx \rangle = \langle (x + \alpha y) - x \mid Gx \rangle \leq f(x + \alpha y) - f(x)$  and  $-\alpha \langle y \mid G(x + \alpha y) \rangle = \langle x - (x + \alpha y) \mid G(x + \alpha y) \rangle \leq f(x) - f(x + \alpha y)$ . Thus

$$0 \leq f(x + \alpha y) - f(x) - \alpha \langle y \mid Gx \rangle \leq \alpha \langle y \mid G(x + \alpha y) - Gx \rangle \quad (17.31)$$

and hence

$$0 \leq \frac{f(x + \alpha y) - f(x)}{\alpha} - \langle y \mid Gx \rangle \leq \langle y \mid G(x + \alpha y) - Gx \rangle. \quad (17.32)$$

This implies that

$$\lim_{\alpha \downarrow 0} \frac{f(x + \alpha y) - f(x)}{\alpha} = \langle y \mid Gx \rangle, \quad (17.33)$$

and therefore that  $f$  is Gâteaux differentiable at  $x$ .  $\square$

**Proposition 17.32** *Let  $f \in \Gamma_0(\mathcal{H})$  and suppose that  $x \in \text{int dom } f$ . Then the following are equivalent:*

- (i)  $f$  is Fréchet differentiable at  $x$ .
- (ii) Every selection of  $\partial f$  is continuous at  $x$ .

*Proof.* (i) $\Rightarrow$ (ii): Assume that this implication is false and set  $u = \nabla f(x)$ . Then there exist a sequence  $(x_n, u_n)_{n \in \mathbb{N}}$  in  $\text{gra } \partial f$  and  $\varepsilon \in \mathbb{R}_{++}$  such that  $x_n \rightarrow x$  and

$$(\forall n \in \mathbb{N}) \quad \|u_n - u\| > 2\varepsilon. \quad (17.34)$$

Note that

$$(\forall n \in \mathbb{N})(\forall y \in \mathcal{H}) \quad \langle y \mid u_n \rangle \leq \langle x_n - x \mid u_n \rangle + f(x + y) - f(x_n). \quad (17.35)$$

The Fréchet differentiability of  $f$  at  $x$  ensures the existence of  $\delta \in \mathbb{R}_{++}$  such that

$$(\forall y \in B(0; \delta)) \quad f(x + y) - f(x) - \langle y \mid u \rangle \leq \varepsilon \|y\|. \quad (17.36)$$

In view of (17.34), there exists a sequence  $(z_n)_{n \in \mathbb{N}}$  in  $B(0; 1)$  such that  $(\forall n \in \mathbb{N}) \langle z_n \mid u_n - u \rangle > 2\varepsilon$ . Hence, using (17.35), (17.36), the boundedness of  $(u_n)_{n \in \mathbb{N}}$ , which is guaranteed by Proposition 16.14(iii), and the continuity of  $f$  at  $x$ , which is guaranteed by Corollary 8.30(ii), we obtain

$$2\varepsilon\delta < \delta \langle z_n \mid u_n - u \rangle$$

$$\begin{aligned}
&= \langle \delta z_n \mid u_n \rangle - \langle \delta z_n \mid u \rangle \\
&\leq \langle x_n - x \mid u_n \rangle + f(x + \delta z_n) - f(x_n) - \langle \delta z_n \mid u \rangle \\
&= (f(x + \delta z_n) - f(x) - \langle \delta z_n \mid u \rangle) + \langle x_n - x \mid u_n \rangle + f(x) - f(x_n) \\
&\leq \varepsilon \delta + \|x_n - x\| \|u_n\| + f(x) - f(x_n) \\
&\rightarrow \varepsilon \delta,
\end{aligned} \tag{17.37}$$

which is absurd.

(ii) $\Rightarrow$ (i): Let  $G$  be a selection of  $\partial f$ . There exists  $\delta \in \mathbb{R}_{++}$  such that  $B(x; \delta) \subset \text{dom } G = \text{dom } \partial f$ . Now take  $y \in B(0; \delta)$ . Then  $\langle y \mid Gx \rangle \leq f(x + y) - f(x)$  and  $\langle -y \mid G(x + y) \rangle \leq f(x) - f(x + y)$ . Thus

$$\begin{aligned}
0 &\leq f(x + y) - f(x) - \langle y \mid Gx \rangle \\
&\leq \langle y \mid G(x + y) - Gx \rangle \\
&\leq \|y\| \|G(x + y) - Gx\|.
\end{aligned} \tag{17.38}$$

Since  $G$  is continuous at  $x$ , this implies that

$$\lim_{0 \neq \|y\| \rightarrow 0} \frac{f(x + y) - f(x) - \langle Gx \mid y \rangle}{\|y\|} = 0. \tag{17.39}$$

Therefore,  $f$  is Fréchet differentiable at  $x$ .  $\square$

Item (ii) of the next result shows that the sufficient condition for Fréchet differentiability discussed in Fact 2.50 is also necessary for functions in  $\Gamma_0(\mathcal{H})$ .

**Corollary 17.33** *Let  $f \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable in a neighborhood  $D$  of a point  $x \in \text{dom } f$ . Then the following hold:*

- (i)  $\nabla f$  is strong-to-weak continuous on  $D$ .
- (ii)  $f$  is Fréchet differentiable at  $x$  if and only if  $\nabla f$  is continuous at  $x$ .

*Proof.* This is clear from Proposition 17.31 and Proposition 17.32.  $\square$

**Corollary 17.34** *Let  $f \in \Gamma_0(\mathcal{H})$  be Fréchet differentiable on  $\text{int dom } f$ . Then  $\nabla f$  is continuous on  $\text{int dom } f$ .*

**Corollary 17.35** *Suppose that  $\mathcal{H}$  is finite-dimensional. Then Gâteaux and Fréchet differentiability are the same notions for functions in  $\Gamma_0(\mathcal{H})$ .*

*Proof.* Combine Proposition 17.31 and Proposition 17.32.  $\square$

The finite-dimensional setting allows for the following useful variant of Proposition 17.26.

**Proposition 17.36** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \mathcal{H}$ , and let  $u \in \mathcal{H}$ . Then  $\partial f(x) = \{u\}$  if and only if  $f$  is Fréchet differentiable at  $x$  and  $u = \nabla f(x)$ .*



*Proof.* Suppose first that  $\partial f(x) = \{u\}$  and set  $C = \text{dom } f$ . By Corollary 16.29,  $\overline{C} = \text{dom } \partial f$ . Hence, using Theorem 20.40 and Proposition 21.14, we deduce that  $N_C x = N_{\overline{C}} x = \text{rec}(\partial f(x)) = \text{rec}\{u\} = \{0\}$ . By Corollary 6.44 and Corollary 8.30(iii),  $x \in \text{int } C = \text{cont } f$ . It thus follows from Proposition 17.26(ii) that  $f$  is Gâteaux differentiable at  $x$  and that  $\nabla f(x) = u$ . Invoking Corollary 17.35, we obtain that  $f$  is Fréchet differentiable at  $x$ . The reverse implication is a consequence of Proposition 17.26(i).  $\square$

**Example 17.37** Let  $(u_n)_{n \in \mathbb{N}}$  be a sequence in  $B(0; 1)$  that converges weakly to 0 and let  $(\alpha_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  such that  $\alpha_n \downarrow 0$ . Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty] : x \mapsto \sup_{n \in \mathbb{N}} (\langle x | u_n \rangle - \alpha_n). \quad (17.40)$$

Then the following hold:

- (i)  $f \in \Gamma_0(\mathcal{H})$ ,  $\text{dom } f = \mathcal{H}$ , and  $f$  is Lipschitz continuous with constant 1.
- (ii)  $f \geq 0$ ,  $f(0) = 0$ , and  $f$  is Gâteaux differentiable at 0 with  $\nabla f(0) = 0$ .
- (iii)  $f$  is Fréchet differentiable at 0 if and only if  $u_n \rightarrow 0$ .

*Proof.* Fix  $x \in \mathcal{H}$ .

(i): As a supremum of lower semicontinuous and convex functions, the function  $f$  is also lower semicontinuous and convex by Proposition 9.3. Since  $u_n \rightharpoonup 0$ , it follows that  $\langle x | u_n \rangle - \alpha_n \rightarrow 0$ . Thus,  $\text{dom } f = \mathcal{H}$ . Now let  $y \in \mathcal{H}$ . Then  $(\forall n \in \mathbb{N}) \langle x | u_n \rangle - \alpha_n = \langle x - y | u_n \rangle + \langle y | u_n \rangle - \alpha_n \leq \|x - y\| + f(y)$ . Hence  $f(x) \leq \|x - y\| + f(y)$ , which supplies the Lipschitz continuity of  $f$  with constant 1.

(ii): Since  $0 = \lim(\langle x | u_n \rangle - \alpha_n) \leq \sup_{n \in \mathbb{N}} (\langle x | u_n \rangle - \alpha_n)$ , it is clear that  $f \geq 0$ , that  $f(0) = 0$ , and that  $0 \in \partial f(0)$ . Hence  $f'(0; \cdot) \geq 0$  by Proposition 17.3. Now fix  $y \in \mathcal{H}$  and let  $m \in \mathbb{N}$ . Then there exists  $\beta \in \mathbb{R}_{++}$  such that  $(\forall n \in \{0, \dots, m-1\}) (\forall \alpha \in ]0, \beta]) \alpha \langle y | u_n \rangle < \alpha_n$ . For every  $\alpha \in ]0, \beta]$ ,  $\max_{0 \leq n \leq m-1} (\langle y | u_n \rangle - \alpha_n / \alpha) < 0$  and hence

$$\begin{aligned} 0 &\leq f(\alpha y) / \alpha \\ &= \sup_{n \in \mathbb{N}} (\langle y | u_n \rangle - \alpha_n / \alpha) \\ &= \max \left\{ \max_{0 \leq n \leq m-1} (\langle y | u_n \rangle - \alpha_n / \alpha), \sup_{n \geq m} (\langle y | u_n \rangle - \alpha_n / \alpha) \right\} \\ &= \sup_{n \geq m} (\langle y | u_n \rangle - \alpha_n / \alpha) \\ &\leq \sup_{n \geq m} \langle y | u_n \rangle. \end{aligned} \quad (17.41)$$

In view of Proposition 17.2(ii), we deduce that

$$(\forall m \in \mathbb{N}) \quad 0 \leq f'(0; y) = \lim_{\alpha \downarrow 0} \frac{f(\alpha y)}{\alpha} \leq \sup_{n \geq m} \langle y | u_n \rangle. \quad (17.42)$$

Hence

$$0 \leq f'(0; y) \leq \inf_{m \in \mathbb{N}} \sup_{n \geq m} \langle y \mid u_n \rangle = \overline{\lim} \langle y \mid u_n \rangle = 0. \quad (17.43)$$

Therefore,  $f'(0; \cdot) \equiv 0$ , i.e.,  $f$  is Gâteaux differentiable at 0 with  $\nabla f(0) = 0$ .

(iii): Since  $f \geq 0 = f(0)$ , it suffices to show that

$$u_n \rightarrow 0 \quad \Leftrightarrow \quad \lim_{0 \neq \|y\| \rightarrow 0} \frac{f(y)}{\|y\|} = 0. \quad (17.44)$$

Assume first that  $u_n \not\rightarrow 0$ . Then there exist  $\delta \in \mathbb{R}_{++}$  and a subsequence  $(u_{k_n})_{n \in \mathbb{N}}$  of  $(u_n)_{n \in \mathbb{N}}$  such that  $\inf_{n \in \mathbb{N}} \|u_{k_n}\| \geq \delta$ . Set

$$(\forall n \in \mathbb{N}) \quad y_n = \sqrt{\alpha_{k_n}} \frac{u_{k_n}}{\|u_{k_n}\|}. \quad (17.45)$$

Then  $(\forall n \in \mathbb{N}) \quad \|y_n\| = \sqrt{\alpha_{k_n}}$  and  $f(y_n) \geq \langle y_n \mid u_{k_n} \rangle - \alpha_{k_n} = \sqrt{\alpha_{k_n}} \|u_{k_n}\| - \alpha_{k_n} \geq \sqrt{\alpha_{k_n}} \delta - \alpha_{k_n}$ . Hence

$$y_n \rightarrow 0 \quad \text{and} \quad \underline{\lim} \frac{f(y_n)}{\|y_n\|} \geq \underline{\lim} (\delta - \sqrt{\alpha_{k_n}}) = \delta > 0. \quad (17.46)$$

Now assume that  $u_n \rightarrow 0$  and fix  $\varepsilon \in ]0, 1[$ . There exists  $m \in \mathbb{N}$  such that for every integer  $n \geq m$  we have  $\|u_n\| \leq \varepsilon$ , and hence  $(\forall y \in \mathcal{H} \setminus \{0\})$   $\langle y \mid u_n \rangle - \alpha_n \leq \langle y \mid u_n \rangle \leq \|y\| \|u_n\| \leq \|y\| \varepsilon$ . Thus

$$(\forall y \in \mathcal{H} \setminus \{0\}) (\forall n \in \mathbb{N}) \quad n \geq m \quad \Rightarrow \quad \frac{\langle y \mid u_n \rangle - \alpha_n}{\|y\|} \leq \varepsilon. \quad (17.47)$$

Set  $\delta = \min_{0 \leq n \leq m-1} \alpha_n / (1 - \varepsilon)$ . Then

$$(\forall y \in B(0; \delta)) (\forall n \in \{0, \dots, m-1\}) \quad \langle y \mid u_n \rangle - \alpha_n \leq \|y\| - \alpha_n \leq \varepsilon \|y\|. \quad (17.48)$$

Hence

$$(\forall y \in B(0; \delta) \setminus \{0\}) (\forall n \in \{0, \dots, m-1\}) \quad \frac{\langle y \mid u_n \rangle - \alpha_n}{\|y\|} \leq \varepsilon. \quad (17.49)$$

Altogether,

$$(\forall y \in B(0; \delta) \setminus \{0\}) \quad \frac{f(y)}{\|y\|} = \sup_{n \in \mathbb{N}} \frac{\langle y \mid u_n \rangle - \alpha_n}{\|y\|} \leq \varepsilon, \quad (17.50)$$

which completes the proof of (17.44).  $\square$

## 17.6 Differentiability and Continuity

We first illustrate the fact that, at a point, a convex function may be Gâteaux differentiable but not continuous.

**Example 17.38** Let  $C$  be as in Example 8.33 and set  $f = \iota_C$ . Then  $0 \in (\text{core } C) \setminus (\text{int } C)$ ,  $\partial f(0) = \{0\}$  is a singleton, and  $f$  is Gâteaux differentiable at 0 with  $\nabla f(0) = 0$ . Moreover,  $f$  is lower semicontinuous at 0 by Proposition 16.3(iv). However, since  $0 \notin \text{int dom } f = \text{int } C$ ,  $f$  is not continuous at 0.

**Proposition 17.39** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and Gâteaux differentiable at  $x \in \text{dom } f$ . Then the following hold:*

- (i)  $f$  is lower semicontinuous at  $x$ .
- (ii) Suppose that  $\mathcal{H}$  is finite-dimensional. Then  $x \in \text{int dom } f$  and  $f$  is continuous on  $\text{int dom } f$ .

*Proof.* (i): It follows from Proposition 17.26(i) that  $x \in \text{dom } \partial f$  and hence from Proposition 16.3(iv) that  $f$  is lower semicontinuous at  $x$ .

(ii): Note that  $x \in \text{core}(\text{dom } f)$  by definition of Gâteaux differentiability. Proposition 6.12(iii) implies that  $x \in \text{int dom } f$ . By Corollary 8.30(iii),  $f$  is continuous on  $\text{int dom } f$ .  $\square$

The next example shows that Proposition 17.39 is sharp and that the implication in Proposition 17.26(ii) cannot be reversed even when  $x \in \text{int dom } f$ .

**Example 17.40** Let  $f$  be the discontinuous linear functional of Example 2.20 and set  $g = f^2$ . Then  $g$  is convex,  $\text{dom } g = \mathcal{H}$ , and  $g$  is Gâteaux differentiable at 0. By Proposition 17.39,  $g$  is lower semicontinuous at 0. However,  $g$  is neither continuous nor Fréchet differentiable at 0.

We conclude with two additional continuity results.

**Proposition 17.41** *Let  $f \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable at  $x \in \text{dom } f$ . Then  $x \in \text{int dom } f$  and  $f$  is continuous on  $\text{int dom } f$ .*

*Proof.* Since  $x$  is a point of Gâteaux differentiability, we have  $x \in \text{core dom } f$ . By Corollary 8.30(ii) and Fact 9.16,  $\text{cont } f = \text{int dom } f = \text{core dom } f$ .  $\square$

**Proposition 17.42** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex, and suppose that  $f$  is Gâteaux differentiable on some open subset  $U$  of  $\text{dom } f$ . Then  $f$  is continuous on  $\text{int dom } f$ .*

*Proof.* Observe that  $U \subset \text{int dom } f$ . Let  $x \in U$ , take  $\rho \in \mathbb{R}_{++}$  such that  $C = B(x; \rho) \subset U$ , and set  $g = f + \iota_C$ . Proposition 17.39 implies that  $f$  is lower semicontinuous and real-valued on  $C$ . Consequently,  $g \in \Gamma_0(\mathcal{H})$ . Hence, Corollary 8.30(ii) asserts that  $g$  is continuous and real-valued on  $\text{int dom } g = \text{int } C$ . Since  $g|_C = f|_C$  and since  $x \in \text{int } C$ , we deduce that  $x \in \text{cont } f$ . Thus, the conclusion follows from Theorem 8.29.  $\square$

## Exercises

**Exercise 17.1** Let  $A \in \mathcal{B}(\mathcal{H})$  be self-adjoint and positive, and set  $q_A: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2) \langle x | Ax \rangle$ . Suppose that  $\text{ran } A$  is closed. Show that the following are equivalent:

- (i)  $q_A$  is supercoercive.
- (ii)  $\text{dom } q_A^* = \mathcal{H}$ .
- (iii)  $A$  is surjective.

**Exercise 17.2** Let  $A \in \mathcal{B}(\mathcal{H})$  be self-adjoint, and set  $q_A: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2) \langle x | Ax \rangle$ . Show that  $A$  is positive if and only if  $A^\dagger$  is.

**Exercise 17.3** Prove Proposition 17.13.

**Exercise 17.4** Set  $f: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto x^4$ . Use condition (ii) of Proposition 17.13 to show that  $f$  is strictly convex. Since  $f''(0) = 0$ , this demonstrates that strict convexity does not imply condition (iv) of Proposition 17.13.

**Exercise 17.5 (Bregman distance)** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be strictly convex, proper, and Gâteaux differentiable on  $\text{int dom } f \neq \emptyset$ . Show that the *Bregman distance*

$$D: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty]$$

$$(x, y) \mapsto \begin{cases} f(x) - f(y) - \langle x - y | \nabla f(y) \rangle, & \text{if } y \in \text{int dom } f; \\ +\infty, & \text{otherwise,} \end{cases} \quad (17.51)$$

satisfies  $(\forall (x, y) \in \text{int dom } f \times \text{int dom } f) \ D(x, y) \geq 0$  and  $[D(x, y) = 0 \Leftrightarrow x = y]$ .

**Exercise 17.6** Use Proposition 17.12, Proposition 17.13, and Exercise 8.7 to prove Corollary 17.15.

**Exercise 17.7** Explain why the function  $f$  constructed in Proposition 17.12 and Corollary 17.15 is independent of the choice of the vector  $y$ .

**Exercise 17.8** Revisit Example 9.29 and Example 9.30 via Corollary 17.15.

**Exercise 17.9** Verify the details of Example 17.16.

**Exercise 17.10** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $x \in \text{dom } f$ , and suppose that  $f'(x; \cdot)$  is continuous at  $y \in \mathcal{H}$ . Use Theorem 17.19 to show that  $f'(x; y) = \max \langle y | \partial f(x) \rangle$ .

**Exercise 17.11** Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and suppose that  $f+g$  is Gâteaux differentiable on  $\mathcal{H}$ . Show that both  $f$  and  $g$  are Gâteaux differentiable.

**Exercise 17.12 (partial derivatives)** Consider Proposition 16.6 and suppose in addition that  $\mathbf{f}$  is proper and convex, that  $\mathbf{x} = (x_i)_{i \in I} \in \text{cont } \mathbf{f}$ , and that for every  $i \in I$ ,  $\mathbf{f} \circ Q_i$  is Gâteaux differentiable at  $x_i$ . Show that  $\mathbf{f}$  is Gâteaux differentiable at  $\mathbf{x}$  and that

$$\nabla \mathbf{f}(\mathbf{x}) = \bigtimes_{i \in I} \nabla (\mathbf{f} \circ Q_i)(x_i). \quad (17.52)$$

**Exercise 17.13** Suppose that  $\mathcal{H}$  is finite-dimensional, and let  $C$  be a bounded closed convex subset of  $\mathcal{H}$  such that  $0 \in \text{int } C$ . Suppose further that, for every  $y \in \text{bdry } C$ ,  $N_C y$  is a ray. Show that  $m_C$  is Gâteaux differentiable on  $\mathcal{H} \setminus \{0\}$ .

**Exercise 17.14** Consider Example 17.30, which deals with a nonempty closed convex set  $C$  and a point  $x \in C$  at which  $\iota_C$  is not Gâteaux differentiable, and yet  $N_C x = \{0\}$ . Demonstrate the impossibility of the existence of a nonempty closed convex subset  $C$  of  $\mathcal{H}$  such that  $\text{ran } N_D = \{0\}$  and  $\iota_D$  is nowhere Gâteaux differentiable.



# Chapter 18

## Further Differentiability Results

Further results concerning derivatives and subgradients are collected in this chapter. The Ekeland–Lebourg theorem gives conditions under which the set of points of Fréchet differentiability is a dense  $G_\delta$  subset of the domain of the function. Formulas for the subdifferential of a maximum and of an infimal convolution are provided, and the basic duality between differentiability and strict convexity is presented. Another highlight of this chapter is the Baillon–Haddad theorem, which states that nonexpansiveness and firm nonexpansiveness are identical properties for gradients of convex functions. Finally, the subdifferential operator of the distance to a convex set is analyzed in detail.

### 18.1 The Ekeland–Lebourg Theorem

**Proposition 18.1** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex and suppose that  $x \in \text{cont } f$ . Then  $f$  is Fréchet differentiable at  $x$  if and only if*

$$(\forall \varepsilon \in \mathbb{R}_{++})(\exists \eta \in \mathbb{R}_{++})(\forall y \in \mathcal{H}) \\ \|y\| = 1 \Rightarrow f(x + \eta y) + f(x - \eta y) - 2f(x) < \eta \varepsilon. \quad (18.1)$$

*Proof.* We assume that  $\mathcal{H} \neq \{0\}$ . Let  $\varepsilon \in \mathbb{R}_{++}$  and let  $y \in \mathcal{H}$  be such that  $\|y\| = 1$ . Assume that  $f$  is Fréchet differentiable at  $x$ . Then there exists  $\eta \in \mathbb{R}_{++}$  such that

$$(\forall z \in \mathcal{H}) \quad \|z\| = 1 \Rightarrow f(x + \eta z) - f(x) - \langle \eta z \mid \nabla f(x) \rangle < (\varepsilon/2)\eta. \quad (18.2)$$

Letting  $z$  be successively  $y$  and then  $-y$  in (18.2) and adding the two resulting inequalities yields (18.1). Conversely, let  $\eta \in \mathbb{R}_{++}$  be such that (18.1) holds. Using Proposition 16.14(ii), we take  $u \in \partial f(x)$ . Then

$$\begin{aligned}
0 &\leq f(x + \eta y) - f(x) - \langle \eta y \mid u \rangle \\
&< \eta \varepsilon + f(x) - f(x - \eta y) + \langle -\eta y \mid u \rangle \\
&\leq \eta \varepsilon.
\end{aligned} \tag{18.3}$$

Let  $\alpha \in ]0, \eta]$ . Then Proposition 17.2(i) implies that  $(f(x + \alpha y) - f(x))/\alpha \leq (f(x + \eta y) - f(x))/\eta$ . Hence, since  $u \in \partial f(x)$ , (18.3) yields  $(\forall z \in B(0; 1))$   $0 \leq f(x + \eta z) - f(x) - \langle \eta z \mid u \rangle \leq \eta \|z\| \varepsilon$ . It follows that  $f$  is Fréchet differentiable at  $x$  and that  $\nabla f(x) = u$ .  $\square$

**Proposition 18.2** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $\varepsilon \in \mathbb{R}_{++}$ , and set*

$$S_\varepsilon = \bigcup_{\eta \in \mathbb{R}_{++}} \left\{ x \in \text{cont } f \mid \sup_{y \in \mathcal{H}, \|y\|=1} \frac{f(x + \eta y) + f(x - \eta y) - 2f(x)}{\eta} < \varepsilon \right\}. \tag{18.4}$$

Then  $S_\varepsilon$  is open.

*Proof.* If  $\text{cont } f = \emptyset$  or  $\mathcal{H} = \{0\}$ , the result is clear. We therefore assume otherwise. Take  $x \in S_\varepsilon$ . Since  $f$  is continuous at  $x$ , Theorem 8.29 implies the existence of  $\eta_1 \in \mathbb{R}_{++}$  and  $\beta \in \mathbb{R}_{++}$  such that

$$(\forall y \in B(x; \eta_1)) (\forall z \in B(x; \eta_1)) \quad |f(y) - f(z)| \leq \beta \|y - z\|. \tag{18.5}$$

Furthermore, using the definition of  $S_\varepsilon$  and Proposition 17.2(i), we deduce the existence of  $\eta \in ]0, \eta_1[$  such that

$$\sigma = \sup_{y \in \mathcal{H}, \|y\|=1} \frac{f(x + \eta y) + f(x - \eta y) - 2f(x)}{\eta} < \varepsilon. \tag{18.6}$$

Take  $\eta_2 \in [0, \min\{\eta_1 - \eta, \eta(\varepsilon - \sigma)/(4\beta)\}]$ . Then for every  $z \in B(x; \eta_2)$  and every  $y \in \mathcal{H}$  such that  $\|y\| = 1$ , we have

$$\begin{aligned}
\frac{f(z + \eta y) + f(z - \eta y) - 2f(z)}{\eta} &\leq \frac{f(x + \eta y) + f(x - \eta y) - 2f(x)}{\eta} \\
&\quad + \frac{f(z + \eta y) - f(x + \eta y)}{\eta} \\
&\quad + \frac{f(z - \eta y) - f(x - \eta y)}{\eta} \\
&\quad + 2 \frac{f(x) - f(z)}{\eta} \\
&\leq \sigma + \frac{4\beta}{\eta} \|x - z\| \\
&\leq \sigma + \frac{4\beta}{\eta} \eta_2 \\
&< \varepsilon.
\end{aligned} \tag{18.7}$$



Consequently,  $B(x; \eta_2) \subset S_\varepsilon$ .  $\square$

**Theorem 18.3 (Ekeland–Lebourg)** *Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be convex, and suppose that  $\text{cont } f \neq \emptyset$ . Then the set of points at which  $f$  is Fréchet differentiable is a dense  $G_\delta$  subset of  $\overline{\text{dom } f}$ .*

*Proof.* We assume that  $\mathcal{H} \neq \{0\}$ . Set  $\mathcal{X} = \overline{\text{dom } f}$ . It follows from Theorem 8.29 that  $\text{cont } f = \text{int dom } f$  and from Proposition 3.36(iii) that  $\overline{\text{int dom } f} = \mathcal{X}$ . Take  $y \in \text{int dom } f$ . Then there exists  $\rho \in \mathbb{R}_{++}$  such that  $B(y; \rho) \subset \text{int dom } f$  and  $f$  is bounded on  $B(y; \rho)$ . Now let  $\psi: \mathbb{R} \rightarrow \mathbb{R}$  be such that  $\psi$  is differentiable on  $[0, \rho]$ ,  $\psi|_{]0, \rho[} > 0$ , and  $\psi(0) = \psi(\rho) = 0$  (for instance,  $\psi: t \mapsto \sin(t\pi/\rho)$ ). Set

$$\varphi: B(y; \rho) \rightarrow [0, +\infty]: x \mapsto \begin{cases} 1/\psi(\|x - y\|), & \text{if } 0 < \|x - y\| < \rho; \\ +\infty, & \text{otherwise,} \end{cases} \quad (18.8)$$

and  $U = (\text{int } B(y; \rho)) \setminus \{y\}$ . Then  $\varphi$  is proper and lower semicontinuous. Combining Example 2.52 and Fact 2.51, we see that  $\varphi$  is Fréchet differentiable on  $U$ . Fix  $\varepsilon \in \mathbb{R}_{++}$ . Applying Theorem 1.45(i)&(iii) in the metric space  $B(y; \rho)$  to the function  $\varphi - f$ , we deduce the existence of  $z_\varepsilon \in B(y; \rho)$  such that

$$(\varphi - f)(z_\varepsilon) \leq \varepsilon + \inf(\varphi - f)(B(y; \rho)) \quad (18.9)$$

and

$$(\forall x \in B(y; \rho)) \quad (\varphi - f)(z_\varepsilon) - (\varphi - f)(x) \leq \varepsilon \|x - z_\varepsilon\|. \quad (18.10)$$

Since  $\varphi - f$  is bounded below and takes on the value  $+\infty$  only on  $B(y; \rho) \setminus U$ , (18.9) implies that  $z_\varepsilon \in U$ . Hence,  $\varphi$  is Fréchet differentiable at  $z_\varepsilon$  and there exists  $\eta \in \mathbb{R}_{++}$  such that  $B(z_\varepsilon; \eta) \subset U$  and

$$(\forall x \in B(z_\varepsilon; \eta)) \quad \varphi(x) - \varphi(z_\varepsilon) - \langle x - z_\varepsilon | \nabla \varphi(z_\varepsilon) \rangle \leq \varepsilon \|x - z_\varepsilon\|. \quad (18.11)$$

Adding (18.10) and (18.11), we obtain

$$(\forall x \in B(z_\varepsilon; \eta)) \quad f(x) - f(z_\varepsilon) - \langle x - z_\varepsilon | \nabla \varphi(z_\varepsilon) \rangle \leq 2\varepsilon \|x - z_\varepsilon\|. \quad (18.12)$$

Hence

$$(\forall r \in \mathcal{H}) \quad \|r\| = 1 \Rightarrow f(z_\varepsilon + \eta r) - f(z_\varepsilon) - \langle \eta r | \nabla \varphi(z_\varepsilon) \rangle \leq 2\varepsilon \eta. \quad (18.13)$$

Invoking the convexity of  $f$ , and considering (18.13) for  $r$  and  $-r$ , we deduce that for every  $r \in \mathcal{H}$  such that  $\|r\| = 1$ , we have

$$0 \leq \frac{f(z_\varepsilon + \eta r) + f(z_\varepsilon - \eta r) - 2f(z_\varepsilon)}{\eta} \leq 4\varepsilon. \quad (18.14)$$

It follows that

$$\sup_{\substack{r \in \mathcal{H} \\ \|r\|=1}} (f(z_\varepsilon + \eta r) + f(z_\varepsilon - \eta r) - 2f(z_\varepsilon)) < 5\eta\varepsilon. \quad (18.15)$$

For the remainder of this proof, we adopt the notation (18.4). Since  $z_\varepsilon \in S_{5\varepsilon}$ , it follows from Proposition 18.2 that  $S_\varepsilon$  is dense and open in  $\mathcal{X}$ . Thus, by Corollary 1.44, the set  $S = \bigcap_{n \in \mathbb{N} \setminus \{0\}} S_{1/n}$  is a dense  $G_\delta$  subset of  $\mathcal{X}$ . Using (18.15) and Proposition 18.1, we therefore conclude that  $f$  is Fréchet differentiable on  $S$ .  $\square$

**Proposition 18.4** *Let  $C$  be a nonempty bounded subset of  $\mathcal{H}$ , and let  $\varepsilon \in \mathbb{R}_{++}$ . Then there exist  $x \in \mathcal{H}$  and  $\alpha \in \mathbb{R}_{++}$  such that*

$$\text{diam}(\{u \in C \mid \sigma_C(x) - \alpha < \langle x \mid u \rangle\}) \leq \varepsilon. \quad (18.16)$$

*Proof.* The support function  $\sigma_C$  is real-valued, convex, and continuous on  $\mathcal{H}$  by Example 11.2. In view of Theorem 18.3, there exists a point  $x \in \mathcal{H}$  at which  $\sigma_C$  is Fréchet differentiable. In turn, by Proposition 18.1, there exists  $\delta \in \mathbb{R}_{++}$  such that

$$(\forall y \in \mathcal{H}) \quad \|y\| = 1 \Rightarrow \sigma_C(x + \delta y) + \sigma_C(x - \delta y) - 2\sigma_C(x) < \delta\varepsilon/3. \quad (18.17)$$

Now set  $\alpha = \delta\varepsilon/3$  and assume that  $\text{diam}(\{u \in C \mid \sigma_C(x) - \alpha < \langle x \mid u \rangle\}) > \varepsilon$ . Then there exist  $u$  and  $v$  in  $C$  such that  $\langle x \mid u \rangle > \sigma_C(x) - \alpha$ ,  $\langle x \mid v \rangle > \sigma_C(x) - \alpha$ , and  $\|u - v\| > \varepsilon$ . Set  $z = (u - v)/\|u - v\|$ . Then  $\|z\| = 1$  and  $\langle z \mid u - v \rangle > \varepsilon$ . Hence,

$$\begin{aligned} \sigma_C(x + \delta z) + \sigma_C(x - \delta z) &\geq \langle x + \delta z \mid u \rangle + \langle x - \delta z \mid v \rangle \\ &> 2\sigma_C(x) - 2\alpha + \delta\varepsilon \\ &= 2\sigma_C(x) + \delta\varepsilon/3, \end{aligned} \quad (18.18)$$

which contradicts (18.17).  $\square$

## 18.2 The Subdifferential of a Maximum

**Theorem 18.5** *Let  $(f_i)_{i \in I}$  be a finite family of convex functions from  $\mathcal{H}$  to  $]-\infty, +\infty]$ , and suppose that  $x \in \bigcap_{i \in I} \text{cont } f_i$ . Set  $f = \max_{i \in I} f_i$  and let  $I(x) = \{i \in I \mid f_i(x) = f(x)\}$ . Then*

$$\partial f(x) = \overline{\text{conv}} \bigcup_{i \in I(x)} \partial f_i(x). \quad (18.19)$$

*Proof.* Let  $i \in I(x)$  and  $u \in \partial f_i(x)$ . Then  $(\forall y \in \mathcal{H}) \langle y - x \mid u \rangle \leq f_i(y) - f_i(x) \leq f(y) - f(x)$  and hence  $u \in \partial f(x)$ . This and Proposition 16.3(iii) imply that  $\overline{\text{conv}} \bigcup_{i \in I(x)} \partial f_i(x) \subset \partial f(x)$ . We now argue by contradiction and

assume that this inclusion is strict, i.e., that there exists

$$u \in \partial f(x) \setminus \overline{\text{conv}} \bigcup_{i \in I(x)} \partial f_i(x). \quad (18.20)$$

By Theorem 3.38 and Theorem 17.19, there exist  $y \in \mathcal{H} \setminus \{0\}$  and  $\varepsilon \in \mathbb{R}_{++}$  such that

$$\langle y \mid u \rangle \geq \varepsilon + \max_{i \in I(x)} \sup \langle y \mid \partial f_i(x) \rangle = \varepsilon + \max_{i \in I(x)} f'_i(x; y). \quad (18.21)$$

Since  $y$  and  $\varepsilon$  can be rescaled, we assume that

$$x + y \in \bigcap_{i \in I} \text{dom } f_i = \text{dom } f. \quad (18.22)$$

Now let  $(\alpha_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1[$  such that  $\alpha_n \downarrow 0$ . Since  $I$  is finite, there exist a subsequence of  $(\alpha_n)_{n \in \mathbb{N}}$ , which we still denote by  $(\alpha_n)_{n \in \mathbb{N}}$ , and  $j \in I$  such that

$$(\forall n \in \mathbb{N}) \quad f(x + \alpha_n y) = f_j(x + \alpha_n y). \quad (18.23)$$

Now let  $n \in \mathbb{N}$ . Then  $f_j(x + \alpha_n y) \leq (1 - \alpha_n)f_j(x) + \alpha_n f_j(x + y)$  and thus, by (18.23) and (18.20),

$$\begin{aligned} (1 - \alpha_n)f_j(x) &\geq f_j(x + \alpha_n y) - \alpha_n f_j(x + y) \\ &\geq f(x + \alpha_n y) - \alpha_n f(x + y) \\ &\geq f(x) + \langle \alpha_n y \mid u \rangle - \alpha_n f(x + y) \\ &\geq f_j(x) + \alpha_n \langle y \mid u \rangle - \alpha_n f(x + y). \end{aligned} \quad (18.24)$$

Letting  $n \rightarrow +\infty$  and using (18.22), we deduce that

$$f_j(x) = f(x). \quad (18.25)$$

In view of (18.23), (18.25), (18.20), and (18.21), we obtain

$$\begin{aligned} f'_j(x; y) &\leftarrow \frac{f_j(x + \alpha_n y) - f_j(x)}{\alpha_n} \\ &= \frac{f(x + \alpha_n y) - f(x)}{\alpha_n} \\ &\geq \langle y \mid u \rangle \\ &\geq \varepsilon + f'_j(x; y), \end{aligned} \quad (18.26)$$

which is the desired contradiction.  $\square$

### 18.3 Differentiability of Infimal Convolutions

**Proposition 18.6** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and let  $x$  and  $y$  be in  $\mathcal{H}$ . Suppose that  $x \in \text{dom } \partial(f \square g)$ , that  $(f \square g)(x) = f(y) + g(x - y)$ , and that  $f$  is Gâteaux differentiable at  $y$ . Then  $\partial(f \square g)(x) = \{\nabla f(y)\}$ .*

*Proof.* Proposition 16.48(i) and Proposition 17.26(i) imply that

$$\partial(f \square g)(x) = \partial f(y) \cap \partial g(x - y) = \{\nabla f(y)\} \cap \partial g(x - y) \subset \{\nabla f(y)\}. \quad (18.27)$$

Since  $\partial(f \square g)(x) \neq \emptyset$ , we conclude that  $\partial(f \square g)(x) = \{\nabla f(y)\}$ .  $\square$

**Proposition 18.7** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ , and let  $x$  and  $y$  be in  $\mathcal{H}$ . Suppose that  $f \square g \in \Gamma_0(\mathcal{H})$ , that  $(f \square g)(x) = f(y) + g(x - y)$ , and that  $f$  is Gâteaux differentiable at  $y$ . Then  $x \in \text{cont}(f \square g)$ ,  $f \square g$  is Gâteaux differentiable at  $x$ , and*

$$\nabla(f \square g)(x) = \nabla f(y). \quad (18.28)$$

*In addition, if  $f$  is Fréchet differentiable at  $y$ , then  $f \square g$  is Fréchet differentiable at  $x$ .*

*Proof.* Proposition 17.41 and Proposition 12.6(ii) imply that  $x = y + (x - y) \in (\text{int dom } f) + \text{dom } g \subset \text{dom } f + \text{dom } g = \text{dom}(f \square g)$ . However,  $\text{dom } g + \text{int dom } f = \bigcup_{z \in \text{dom } g} (z + \text{int dom } f)$  is open as a union of open sets. Thus,  $x \in \text{int dom}(f \square g)$  and, since  $f \square g \in \Gamma_0(\mathcal{H})$ , Proposition 16.21 implies that  $x \in \text{cont}(f \square g) \subset \text{dom } \partial(f \square g)$ . Next, we combine Proposition 18.6 and Proposition 17.26(ii) to deduce the Gâteaux differentiability of  $f \square g$  at  $x$  and (18.28). Now assume that  $f$  is Fréchet differentiable at  $y$ . Then, for every  $z \in \mathcal{H} \setminus \{0\}$ ,

$$\begin{aligned} 0 &\leq (f \square g)(x + z) - (f \square g)(x) - \langle z \mid \nabla f(y) \rangle \\ &\leq f(y + z) + g(x - y) - (f(y) + g(x - y)) - \langle z \mid \nabla f(y) \rangle \\ &= f(y + z) - f(y) - \langle z \mid \nabla f(y) \rangle \end{aligned} \quad (18.29)$$

and hence

$$\begin{aligned} 0 &\leq \lim_{0 \neq \|z\| \rightarrow 0} \frac{(f \square g)(x + z) - (f \square g)(x) - \langle z \mid \nabla f(y) \rangle}{\|z\|} \\ &\leq \lim_{0 \neq \|z\| \rightarrow 0} \frac{f(y + z) - f(y) - \langle z \mid \nabla f(y) \rangle}{\|z\|} \\ &= 0. \end{aligned} \quad (18.30)$$

Therefore,  $f \square g$  is Fréchet differentiable at  $x$ .  $\square$

**Corollary 18.8** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Suppose that  $f$  is real-valued, supercoercive, and Fréchet differentiable on  $\mathcal{H}$ . Then  $f \square g$  is Fréchet differentiable on  $\mathcal{H}$ .*

*Proof.* By Proposition 12.14(i),  $f \square g = f \square g \in \Gamma_0(\mathcal{H})$ . Now let  $x \in \mathcal{H}$ . Since  $\text{dom}(f \square g) = \mathcal{H}$  by Proposition 12.6(ii), there exists  $y \in \mathcal{H}$  such that  $(f \square g)(x) = f(y) + g(x - y)$ . Now apply Proposition 18.7.  $\square$

## 18.4 Differentiability and Strict Convexity

In this section, we examine the interplay between Gâteaux differentiability and strict convexity via duality.

**Proposition 18.9** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $f^*$  is strictly convex on every nonempty convex subset of  $\text{dom } \partial f^*$ . Then  $f$  is Gâteaux differentiable on  $\text{int dom } f$ .*

*Proof.* Suppose that  $x \in \text{int dom } f$  and that  $[u_1, u_2] \subset \partial f(x)$ . Then it follows from Corollary 16.24 that  $[u_1, u_2] \subset \text{ran } \partial f = \text{dom}(\partial f)^{-1} = \text{dom } \partial f^*$ . Hence, Proposition 16.27(i) implies that  $f^*$  is affine on  $[u_1, u_2]$ . Consequently,  $u_1 = u_2$  and  $\partial f(x)$  is a singleton. Furthermore,  $x \in \text{cont } f$  by Corollary 8.30(ii) and the conclusion thus follows from Proposition 17.26(ii).  $\square$

**Proposition 18.10** *Let  $f \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable on  $\text{int dom } f$ . Then  $f^*$  is strictly convex on every nonempty convex subset of  $\nabla f(\text{int dom } f)$ .*

*Proof.* Assume to the contrary that there exist two distinct points  $u_1$  and  $u_2$  such that  $f^*$  is affine on  $[u_1, u_2] \subset \nabla f(\text{int dom } f)$ . Choose  $x \in \text{int dom } f$  such that  $\nabla f(x) \in ]u_1, u_2[$ . Then Proposition 16.27(ii) implies that  $[u_1, u_2] \subset \partial f(x)$ , which contradicts the Gâteaux differentiability of  $f$  at  $x$ .  $\square$

**Corollary 18.11** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{dom } \partial f = \text{int dom } f$ . Then  $f$  is Gâteaux differentiable on  $\text{int dom } f$  if and only if  $f^*$  is strictly convex on every nonempty convex subset of  $\text{dom } \partial f^*$ , in which case  $\text{dom } \partial f^* = \nabla f(\text{int dom } f)$ .*

*Proof.* If  $f^*$  is strictly convex on every nonempty convex subset of  $\text{dom } \partial f^*$ , then  $f$  is Gâteaux differentiable on  $\text{int dom } f$  by Proposition 18.9. Let us now assume that  $f$  is Gâteaux differentiable on  $\text{int dom } f$ . Using Proposition 17.26(i) and Corollary 16.24, we see that

$$\nabla f(\text{int dom } f) = \partial f(\text{int dom } f) = \partial f(\text{dom } \partial f) = \text{ran } \partial f = \text{dom } \partial f^*. \quad (18.31)$$

The result thus follows from Proposition 18.10.  $\square$

**Corollary 18.12** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{dom } \partial f = \text{int dom } f$  and  $\text{dom } \partial f^* = \text{int dom } f^*$ . Then the following hold:*

- (i)  *$f$  is Gâteaux differentiable on  $\text{int dom } f$  if and only if  $f^*$  is strictly convex on  $\text{int dom } f^*$ .*

- (ii)  $f$  is strictly convex on  $\text{int dom } f$  if and only if  $f^*$  is Gâteaux differentiable on  $\text{int dom } f^*$ .

*Proof.* (i): This follows from Corollary 18.11.

(ii): Apply (i) to  $f^*$  and use Corollary 13.33.  $\square$

## 18.5 Stronger Notions of Differentiability

In this section, we investigate certain properties of the gradient of convex functions.

**Theorem 18.13** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be a Fréchet differentiable convex function, and let  $\phi: \mathbb{R} \rightarrow \mathbb{R}_+$  be an even convex function that vanishes only at 0. For every  $s \in \mathbb{R}$ , define*

$$\psi(s) = \begin{cases} \phi(s)/|s|, & \text{if } s \neq 0; \\ 0, & \text{if } s = 0, \end{cases} \quad \theta(s) = \int_0^1 \frac{\phi(st)}{t} dt,$$

$$\text{and } \varrho(s) = \sup \{ \nu \in \mathbb{R}_+ \mid 2\theta^*(\nu) \leq \nu s \}. \quad (18.32)$$

Now consider the following statements:

- (i)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \|\nabla f(x) - \nabla f(y)\| \leq \psi(\|x - y\|).$
- (ii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \langle x - y \mid \nabla f(x) - \nabla f(y) \rangle \leq \phi(\|x - y\|).$
- (iii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) f(y) \leq f(x) + \langle y - x \mid \nabla f(x) \rangle + \theta(\|x - y\|).$
- (iv)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) f^*(\nabla f(y)) \geq f^*(\nabla f(x)) + \langle x \mid \nabla f(y) - \nabla f(x) \rangle + \theta^*(\|\nabla f(x) - \nabla f(y)\|).$
- (v)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \langle x - y \mid \nabla f(x) - \nabla f(y) \rangle \geq 2\theta^*(\|\nabla f(x) - \nabla f(y)\|).$
- (vi)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \|\nabla f(x) - \nabla f(y)\| \leq \varrho(\|x - y\|).$

Then (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (iv) $\Rightarrow$ (v) $\Rightarrow$ (vi).

*Proof.* Let  $x$  and  $y$  be in  $\mathcal{H}$ , and set  $u = \nabla f(x)$  and  $v = \nabla f(y)$ .

(i) $\Rightarrow$ (ii): Cauchy–Schwarz.

(ii) $\Rightarrow$ (iii): Since the Gâteaux derivative of  $t \mapsto f(x + t(y - x))$  is  $t \mapsto \langle y - x \mid \nabla f(x + t(y - x)) \rangle$ , we have

$$\begin{aligned} f(y) - f(x) - \langle y - x \mid u \rangle &= \int_0^1 \langle y - x \mid \nabla f(x + t(y - x)) - \nabla f(x) \rangle dt \\ &= \int_0^1 \frac{1}{t} \langle t(y - x) \mid \nabla f(x + t(y - x)) - \nabla f(x) \rangle dt \\ &\leq \int_0^1 \frac{1}{t} \phi(\|y - x\|t) dt \\ &= \theta(\|y - x\|). \end{aligned} \quad (18.33)$$

(iii) $\Rightarrow$ (iv): For every  $z \in \mathcal{H}$ , we derive from Proposition 17.27 that

$$\begin{aligned} -f(z) &\geq -f(x) + \langle x | u \rangle - \langle z | u \rangle - \theta(\|z - x\|) \\ &= f^*(u) - \langle z | u \rangle - \theta(\|z - x\|) \end{aligned} \quad (18.34)$$

and, in turn, we obtain

$$\begin{aligned} f^*(v) &\geq \langle z | v \rangle - f(z) \\ &\geq \langle z | v \rangle + f^*(u) - \langle z | u \rangle - \theta(\|z - x\|) \\ &= f^*(u) + \langle x | v - u \rangle + \langle z - x | v - u \rangle - \theta(\|z - x\|). \end{aligned} \quad (18.35)$$

However, since  $\phi$  is even, so is  $\theta$  and, therefore, it follows from Example 13.7 that  $(\theta \circ \|\cdot\|)^* = \theta^* \circ \|\cdot\|$ . Hence, we derive from (18.35) that

$$\begin{aligned} f^*(v) &\geq f^*(u) + \langle x | v - u \rangle + \sup_{z \in \mathcal{H}} (\langle z - x | v - u \rangle - \theta(\|z - x\|)) \\ &= f^*(u) + \langle x | v - u \rangle + \theta^*(\|v - u\|). \end{aligned} \quad (18.36)$$

(iv) $\Rightarrow$ (v): We have

$$f^*(u) \geq f^*(v) + \langle y | u - v \rangle + \theta^*(\|u - v\|). \quad (18.37)$$

Adding (18.36) and (18.37) yields (v).

(v) $\Rightarrow$ (vi): Cauchy–Schwarz.  $\square$

The following special case concerns convex functions with Hölder continuous gradients.

**Corollary 18.14** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be Fréchet differentiable and convex, let  $\beta \in \mathbb{R}_{++}$ , and let  $p \in ]0, 1]$ . Consider the following statements:*

- (i)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \|\nabla f(x) - \nabla f(y)\| \leq \beta \|x - y\|^p$ .
- (ii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \langle x - y | \nabla f(x) - \nabla f(y) \rangle \leq \beta \|x - y\|^{p+1}$ .
- (iii)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) f(y) \leq f(x) + \langle y - x | \nabla f(x) \rangle + \beta(p+1)^{-1} \|x - y\|^{p+1}$ .
- (iv)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) f^*(\nabla f(y)) \geq f^*(\nabla f(x)) + \langle x | \nabla f(y) - \nabla f(x) \rangle + \beta^{-1/p} p(p+1)^{-1} \|\nabla f(x) - \nabla f(y)\|^{1+1/p}$ .
- (v)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \langle x - y | \nabla f(x) - \nabla f(y) \rangle \geq \frac{2\beta^{-1/p} p(p+1)^{-1} \|\nabla f(x) - \nabla f(y)\|^{1+1/p}}{2\beta^{-1/p} p(p+1)^{-1} \|\nabla f(x) - \nabla f(y)\|^{1+1/p}}$ .
- (vi)  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \|\nabla f(x) - \nabla f(y)\| \leq \beta((p+1)/(2p))^p \|x - y\|^p$ .

Then (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (iv) $\Rightarrow$ (v) $\Rightarrow$ (vi).

*Proof.* This follows from Theorem 18.13 with  $\phi = \beta|\cdot|^{p+1}$ . Indeed,  $\phi$  is an even convex function vanishing only at 0. Moreover,  $\theta: t \mapsto \beta|t|^{p+1}/(p+1)$  and we deduce from Example 13.2(i) and Proposition 13.20(i) that  $\theta^*: \nu \mapsto \beta^{-1/p} p(p+1)^{-1} |\nu|^{1+1/p}$ .  $\square$

Next, we provide several characterizations of convex functions with Lipschitz continuous gradients.

**Theorem 18.15** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\beta \in \mathbb{R}_{++}$ , and set  $h = f^* - (1/\beta)q$ , where  $q = (1/2)\|\cdot\|^2$ . Then the following are equivalent:*

- (i)  *$f$  is Fréchet differentiable on  $\mathcal{H}$  and  $\nabla f$  is  $\beta$ -Lipschitz continuous.*
- (ii)  *$f$  is Fréchet differentiable on  $\mathcal{H}$  and*

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad \langle x - y \mid \nabla f(x) - \nabla f(y) \rangle \leq \beta \|x - y\|^2.$$

- (iii) **(descent lemma)**  *$f$  is Fréchet differentiable on  $\mathcal{H}$  and*

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad f(y) \leq f(x) + \langle y - x \mid \nabla f(x) \rangle + (\beta/2)\|x - y\|^2.$$

- (iv)  *$f$  is Fréchet differentiable on  $\mathcal{H}$  and*

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad f^*(\nabla f(y)) \geq f^*(\nabla f(x)) + \langle x \mid \nabla f(y) - \nabla f(x) \rangle + (1/(2\beta))\|\nabla f(x) - \nabla f(y)\|^2.$$

- (v)  *$f$  is Fréchet differentiable on  $\mathcal{H}$  and  $\nabla f$  is  $(1/\beta)$ -cocoercive.*

- (vi)  *$\beta q - f$  is convex.*

- (vii)  *$f^* - (1/\beta)q$  is convex (i.e.,  $f^*$  is  $1/\beta$ -strongly convex).*

- (viii)  *$h \in \Gamma_0(\mathcal{H})$  and  $f = {}^{1/\beta}(h^*) = \beta q - {}^\beta h \circ \beta \text{Id}$ .*

- (ix)  *$h \in \Gamma_0(\mathcal{H})$  and  $\nabla f = \text{Prox}_{\beta h} \circ \beta \text{Id} = \beta(\text{Id} - \text{Prox}_{h^*}/\beta)$ .*

*Proof.* (i) $\Leftrightarrow$ (ii) $\Leftrightarrow$ (iii) $\Leftrightarrow$ (iv) $\Leftrightarrow$ (v): Set  $p = 1$  in Corollary 18.14.

(v) $\Rightarrow$ (vi): Since  $(1/\beta)\nabla f$  is firmly nonexpansive, so is  $\text{Id} - (1/\beta)\nabla f$  by Proposition 4.2. Hence,  $\nabla(\beta q - f) = \beta \text{Id} - \nabla f$  is monotone and it follows from Proposition 17.10 that  $\beta q - f$  is convex.

(vi) $\Leftrightarrow$ (vii): Proposition 14.2.

(vii) $\Rightarrow$ (viii): Since  $f \in \Gamma_0(\mathcal{H})$  and  $h$  is convex, Corollary 13.33 yields  $h \in \Gamma_0(\mathcal{H})$  and  $h^* \in \Gamma_0(\mathcal{H})$ . Hence, using successively Corollary 13.33, Proposition 14.1, and Theorem 14.3(i), we obtain  $f = f^{**} = (h + (1/\beta)q)^* = {}^{1/\beta}(h^*) = \beta q - {}^\beta h \circ \beta \text{Id}$ .

(viii) $\Rightarrow$ (ix) $\Rightarrow$ (i): Proposition 12.29. □

As seen in Example 4.9, nonexpansive operators may not be firmly nonexpansive. Remarkably, the equivalence (i) $\Leftrightarrow$ (v) in Theorem 18.15 asserts that this is true for the gradient of a convex function  $f: \mathcal{H} \rightarrow \mathbb{R}$ . We record this result next, which is known as the *Baillon–Haddad theorem*.

**Corollary 18.16 (Baillon–Haddad)** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be a Fréchet differentiable convex function and let  $\beta \in \mathbb{R}_{++}$ . Then  $\nabla f$  is  $\beta$ -Lipschitz continuous if and only if  $\nabla f$  is  $(1/\beta)$ -cocoercive. In particular,  $\nabla f$  is nonexpansive if and only if  $\nabla f$  is firmly nonexpansive.*

**Corollary 18.17** *Let  $L \in \mathcal{B}(\mathcal{H})$  be self-adjoint and positive, and let  $x \in \mathcal{H}$ . Then  $\|L\| \langle Lx \mid x \rangle \geq \|Lx\|^2$ .*



*Proof.* Set  $f: \mathcal{H} \rightarrow \mathbb{R}: y \mapsto \langle Ly \mid y \rangle / 2$ . It follows from Example 2.46 that  $\nabla f = L$  is Lipschitz continuous with constant  $\|L\|$ . Hence, the result follows from the implication (i) $\Rightarrow$ (v) in Theorem 18.15.  $\square$

**Corollary 18.18** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\beta \in \mathbb{R}_{++}$ , and set  $q = (1/2)\|\cdot\|^2$ . Then  $f$  is Fréchet differentiable on  $\mathcal{H}$  with a  $\beta$ -Lipschitz continuous gradient if and only if  $f$  is the Moreau envelope of parameter  $1/\beta$  of a function in  $\Gamma_0(\mathcal{H})$ ; more precisely,*

$$f = \left( f^* - \frac{1}{\beta} q \right)^* \square \beta q. \quad (18.38)$$

*Proof.* This follows from the equivalence (i) $\Leftrightarrow$ (viii) in Theorem 18.15.  $\square$

**Corollary 18.19** *Let  $(f_i)_{i \in I}$  be a finite family of functions in  $\Gamma_0(\mathcal{H})$ , let  $(\alpha_i)_{i \in I}$  be a finite family of real numbers in  $]0, 1]$  such that  $\sum_{i \in I} \alpha_i = 1$ , and set  $q = (1/2)\|\cdot\|^2$ . Then  $\sum_{i \in I} \alpha_i \text{Prox}_{f_i}$  is the proximity operator of a function in  $\Gamma_0(\mathcal{H})$ ; more precisely,*

$$\sum_{i=1}^m \alpha_i \text{Prox}_{f_i} = \text{Prox}_h, \quad \text{where} \quad h = \left( \sum_{i \in I} \alpha_i (f_i^* \square q) \right)^* - q. \quad (18.39)$$

*Proof.* Set  $f = \sum_{i \in I} \alpha_i (f_i^* \square q)$ . We derive from Proposition 12.15, Proposition 8.15, and Proposition 12.29 that  $f: \mathcal{H} \rightarrow \mathbb{R}$  is convex and Fréchet differentiable on  $\mathcal{H}$ . Moreover, it follows from (14.7) and Proposition 12.27 that  $\nabla f = \sum_{i \in \mathcal{H}} \alpha_i \text{Prox}_{f_i}$  is nonexpansive as a convex combination of nonexpansive operators. Hence, using the implication (i) $\Rightarrow$ (ix) in Theorem 18.15 with  $\beta = 1$ , we deduce that  $\nabla f = \text{Prox}_h$ , where  $h = f^* - q \in \Gamma_0(\mathcal{H})$ , which gives (18.39).  $\square$

**Remark 18.20** Suppose that  $I = \{1, 2\}$  and that  $\alpha_1 = \alpha_2 = 1/2$  in Corollary 18.19. Then, in view of Corollary 14.8(iv), the function  $h$  in (18.39) is the proximal average  $\text{pav}(f_1, f_2)$ .

## 18.6 Differentiability of the Distance to a Set

**Proposition 18.21** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in C$ . Then the following are equivalent:*

- (i)  $d_C$  is Gâteaux differentiable at  $x$  and  $\nabla d_C(x) = 0$ .
- (ii)  $x \notin \text{spts } C$ , i.e.,  $(\forall u \in \mathcal{H} \setminus \{0\}) \sigma_C(u) > \langle x \mid u \rangle$ .
- (iii)  $T_C x = \mathcal{H}$ , i.e.,  $\overline{\text{cone}}(C - x) = \mathcal{H}$ .

*Proof.* (i) $\Rightarrow$ (ii): Suppose that  $u \in \mathcal{H} \setminus \{0\}$ . Since  $\sigma_C$  and  $\langle \cdot \mid u \rangle$  are positively homogeneous, we assume that  $\|u\| = 1$ . Let  $y \in \mathcal{H}$ . Then, since  $d_C(x) = 0$ , Cauchy-Schwarz yields

$$\begin{aligned}
0 &= \langle y \mid \nabla d_C(x) \rangle \\
&= \lim_{\alpha \downarrow 0} \alpha^{-1} (d_C(x + \alpha y) - d_C(x)) \\
&= \lim_{\alpha \downarrow 0} \alpha^{-1} \inf_{z \in C} \|x + \alpha y - z\| \\
&\geq \overline{\lim}_{\alpha \downarrow 0} \alpha^{-1} \inf_{z \in C} \langle x + \alpha y - z \mid u \rangle \\
&= \langle y \mid u \rangle + \overline{\lim}_{\alpha \downarrow 0} (-\alpha^{-1}) \sup_{z \in C} \langle z - x \mid u \rangle \\
&= \langle y \mid u \rangle - \underline{\lim}_{\alpha \downarrow 0} \alpha^{-1} (\sigma_C(u) - \langle x \mid u \rangle) \\
&= \begin{cases} \langle y \mid u \rangle, & \text{if } \sigma_C(u) = \langle x \mid u \rangle; \\ -\infty, & \text{if } \sigma_C(u) > \langle x \mid u \rangle. \end{cases} \tag{18.40}
\end{aligned}$$

Thus, if  $\sigma_C(u) = \langle x \mid u \rangle$ , then  $\langle y \mid u \rangle \leq 0$ . Since  $y$  is an arbitrary point in  $\mathcal{H}$ , we obtain  $u = 0$ , in contradiction to our hypothesis. Hence  $\sigma_C(u) > \langle x \mid u \rangle$ .

(ii) $\Rightarrow$ (iii): Assume that  $\overline{\text{cone}}(C - x) = \bigcup_{\lambda \in \mathbb{R}_{++}} \lambda(C - x) \neq \mathcal{H}$  and take  $y \in \mathcal{H} \setminus \overline{\text{cone}}(C - x)$ . By Theorem 3.38, there exists  $u \in \mathcal{H} \setminus \{0\}$  such that  $\langle y \mid u \rangle > \sup_{\gamma \in \mathbb{R}_{++}} \gamma \sup \langle C - x \mid u \rangle = \sup_{\gamma \in \mathbb{R}_{++}} (\gamma(\sigma_C(u) - \langle x \mid u \rangle))$ . We deduce that  $\sigma_C(u) = \langle x \mid u \rangle$ , which is impossible.

(iii) $\Rightarrow$ (i): It follows from Example 16.49 that  $\text{dom } \partial d_C = \mathcal{H}$ . Hence, in view of Proposition 17.26(ii), it suffices to check that  $\partial d_C(x) = \{0\}$ . Take  $u \in \partial d_C(x)$ . Then  $(\forall y \in C) \langle y - x \mid u \rangle \leq d_C(y) - d_C(x) = 0$ . Hence  $\sup \langle C - x \mid u \rangle \leq 0$ , which implies that  $\sup_{\lambda \in \mathbb{R}_{++}} \sup \langle \lambda(C - x) \mid u \rangle \leq 0$  and thus that  $\sup \langle \mathcal{H} \mid u \rangle = \sup \langle \overline{\text{cone}}(C - x) \mid u \rangle \leq 0$ . Therefore,  $u = 0$ .  $\square$

The next result complements Corollary 12.30 and Example 16.49.

**Proposition 18.22** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then exactly one of the following holds:*

- (i) *Suppose that  $x \in \text{int } C$ . Then  $d_C$  is Fréchet differentiable at  $x$  with  $\nabla d_C(x) = 0$ .*
- (ii) *Suppose that  $x \in \text{bdry } C$ . Then the following hold:*
  - (a) *Suppose that  $x \notin \text{spts } C$ . Then  $d_C$  is not Fréchet differentiable at  $x$ , but  $d_C$  is Gâteaux differentiable at  $x$  with  $\nabla d_C(x) = 0$ .*
  - (b) *Suppose that  $x \in \text{spts } C$ . Then  $d_C$  is not Gâteaux differentiable at  $x$  and  $\partial d_C(x) = N_C(x) \cap B(0; 1)$ .*
- (iii) *Suppose that  $x \notin C$ . Then  $d_C$  is Fréchet differentiable at  $x$  with  $\nabla d_C(x) = (x - P_C x)/d_C(x)$ .*

*Proof.* (i): There exists  $\varepsilon \in \mathbb{R}_{++}$  such that  $d_C|_{B(x; \varepsilon)} = 0$ , which implies that  $d_C$  is Fréchet differentiable at  $x$  with  $\nabla d_C(x) = 0$ .

(ii)(a): Proposition 18.21 implies that  $\nabla d_C(x) = 0$ . Since Theorem 7.4 yields  $\text{spts } C = \text{bdry } C$ , there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  of support points of  $C$  such that  $x_n \rightarrow x$ . Let  $(u_n)_{n \in \mathbb{N}}$  be such that  $(\forall n \in \mathbb{N}) \sigma_C(u_n) = \langle x_n \mid u_n \rangle$

and  $\|u_n\| = 1$ . By Example 16.49,  $(\forall n \in \mathbb{N}) u_n \in N_C x_n \cap B(0; 1) = \partial d_C(x_n)$ . Altogether,  $(x_n, u_n)_{n \in \mathbb{N}}$  is a sequence in  $\text{gra } \partial d_C$  such that  $x_n \rightarrow x$  and  $u_n \not\rightarrow 0 = \nabla d_C(x)$ . In view of Proposition 17.32, we conclude that  $d_C$  is not Fréchet differentiable at  $x$ .

(ii)(b): Let  $u \in N_C x \setminus \{0\}$ . Then  $\{0, u/\|u\|\} \subset N_C x \cap B(0; 1) = \partial d_C(x)$  by Example 16.49. Thus,  $d_C$  is not Gâteaux differentiable at  $x$  by Proposition 17.26(i).

(iii): Set  $f = \|\cdot\|$  and  $g = \iota_C$ . Then  $d_C = f \square g$  and  $d_C(x) = f(x - P_C x) + g(P_C x)$ . Since  $\|\cdot\|$  is Fréchet differentiable at  $x - P_C x$  with  $\nabla \|\cdot\|(x - P_C x) = (x - P_C x)/\|x - P_C x\|$  by Example 2.52, it follows from Proposition 18.7 that  $d_C$  is Fréchet differentiable at  $x$  and that  $\nabla d_C(x) = (x - P_C x)/d_C(x)$ . Alternatively, combine Example 16.49, Proposition 17.26(ii), Proposition 4.8, and Fact 2.50.  $\square$

## Exercises

**Exercise 18.1** Let  $(u_a)_{a \in A}$  be a family in  $\mathcal{H}$ , let  $(\rho_a)_{a \in A}$  be a family in  $] -\infty, +\infty]$ , and set  $f = \sup_{a \in A} (\langle \cdot | u_a \rangle - \rho_a)$ . Show that  $\text{ran } \partial f \subset \overline{\text{conv}} \{u_a\}_{a \in A}$ .

**Exercise 18.2** Let  $f: \mathbb{R} \rightarrow \mathbb{R}: x \mapsto |x|$ . Compute  $\partial f(0)$  (see also Example 16.25) in two different ways, using Theorem 18.5 and Proposition 18.22(ii)(b).

**Exercise 18.3** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\gamma \in \mathbb{R}_{++}$ . Use Corollary 18.8 to establish the Fréchet differentiability of the Moreau envelope  $\gamma f$  (see Proposition 12.29).

**Exercise 18.4** By providing a counterexample, show that the equivalence in Corollary 18.11 fails if the assumption  $\text{dom } \partial f = \text{int dom } f$  is removed.

**Exercise 18.5** Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $r \in \mathcal{K}$ , and let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \|Lx - r\|^2 + \|x\|^2 + d_C^2(x)$ . Show that  $f$  is the Moreau envelope of a function in  $\Gamma_0(\mathcal{H})$ .

**Exercise 18.6 (Legendre function)** Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{dom } \partial f = \text{int dom } f$  and  $\text{dom } \partial f^* = \text{int dom } f^*$ . Suppose that  $f$  is Gâteaux differentiable on  $\text{int dom } f$  and that  $f^*$  is Gâteaux differentiable on  $\text{int dom } f^*$ . These assumptions together mean that  $f$  is a *Legendre function*, as is  $f^*$ . Show that both  $f$  and  $f^*$  are strictly convex on the interior of their domains, respectively. Furthermore, show that  $\nabla f: \text{int dom } f \rightarrow \text{int dom } f^*$  is a bijection, with inverse  $\nabla f^*$ .

**Exercise 18.7** Provide five examples of Legendre functions (see Exercise 18.6) and five examples of functions in  $\Gamma_0(\mathcal{H})$  that are not Legendre functions.

**Exercise 18.8** Let  $A \in \mathcal{B}(\mathcal{H})$  be positive, self-adjoint, and surjective. Using Proposition 17.28 and its notation, deduce from Exercise 18.6 that  $q_A$  is strictly convex, that  $q_A^*$  is strictly convex, and that  $A$  is bijective.

**Exercise 18.9** . Use Fact 2.51 and Corollary 12.30 to give an alternative proof of Proposition 18.22(iii).

**Exercise 18.10** Use Example 7.7 and Proposition 18.22 to construct a function  $f \in \Gamma_0(\mathcal{H})$  and a point  $x \in \mathcal{H}$  such that  $\text{dom } f = \mathcal{H}$ ,  $f$  is Gâteaux differentiable at  $x$ , but  $f$  is not Fréchet differentiable at  $x$ .

**Exercise 18.11** Suppose that  $\mathcal{H}$  is infinite-dimensional, that  $C$  is as in Example 6.18, and that  $Q$  is as in Exercise 8.17. Set

$$f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto m_Q(\|x\|^2, d_C(x)). \quad (18.41)$$

Prove the following:

- (i)  $f$  is continuous, convex, and  $\min f(\mathcal{H}) = 0 = f(0)$ .
- (ii)  $f$  is Gâteaux differentiable at 0 and  $\nabla f(0) = 0$ .
- (iii)  $f$  is Fréchet differentiable on  $\mathcal{H} \setminus \{0\}$ .
- (iv)  $f$  is not Fréchet differentiable at 0.

# Chapter 19

## Duality in Convex Optimization

A convex optimization problem can be paired with a dual problem involving the conjugates of the functions appearing in its (primal) formulation. In this chapter, we study the interplay between primal and dual problems in the context of Fenchel–Rockafellar duality and, more generally, for bivariate functions. The latter approach leads naturally to saddle points and Lagrangians. Special attention is given to minimization under equality constraints and under inequality constraints. We start with a discussion of instances in which all primal solutions can be recovered from an arbitrary dual solution.

Throughout this chapter,  $\mathcal{K}$  is a real Hilbert space.

### 19.1 Primal Solutions via Dual Solutions

In this section we investigate the connections between primal and dual solutions in the context of the minimization problem discussed in Definition 15.19. We recall that given  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$ ,  $g: \mathcal{K} \rightarrow ]-\infty, +\infty]$ , and  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , the primal problem associated with the composite function  $f + g \circ L$  is

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(Lx), \quad (19.1)$$

its dual problem is

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad f^*(L^*v) + g^*(-v), \quad (19.2)$$

the primal optimal value is  $\mu = \inf (f + g \circ L)(\mathcal{H})$ , and the dual optimal value is  $\mu^* = \inf (f^* \circ L^* + g^*)(\mathcal{K})$ . We saw in Proposition 15.21 that  $\mu \geq -\mu^*$ . A solution to (19.1) is called a *primal solution*, and a solution to (19.2) is called a *dual solution*. As noted in Remark 15.20, in principle, dual solutions depend on the ordered triple  $(f, g, L)$ , and we follow the common convention adopted in Definition 15.19.

**Theorem 19.1** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ . Furthermore, set  $\mu = \inf(f + g \circ L)(\mathcal{H})$ , set  $\mu^* = \inf(f^* \circ L^* + g^{*\vee})(\mathcal{K})$ , let  $x \in \mathcal{H}$ , and let  $v \in \mathcal{K}$ . Then the following are equivalent:*

- (i)  $x$  is a primal solution,  $v$  is a dual solution, and  $\mu = -\mu^*$ .
- (ii)  $L^*v \in \partial f(x)$  and  $-v \in \partial g(Lx)$ .
- (iii)  $x \in \partial f^*(L^*v) \cap L^{-1}(\partial g^*(-v))$ .

*Proof.* (i) $\Leftrightarrow$ (ii): Since  $f + g \circ L$  is proper, we derive from Theorem 16.23 the equivalences

$$\begin{aligned} \text{(i)} &\Leftrightarrow f(x) + g(Lx) = \mu = -\mu^* = -(f^*(L^*v) + g^*(-v)) \\ &\Leftrightarrow (f(x) + f^*(L^*v) - \langle x | L^*v \rangle) + (g(Lx) + g^*(-v) - \langle Lx | -v \rangle) = 0 \\ &\Leftrightarrow L^*v \in \partial f(x) \text{ and } -v \in \partial g(Lx). \end{aligned} \quad (19.3)$$

(ii) $\Leftrightarrow$ (iii): Corollary 16.24.  $\square$

**Corollary 19.2** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\inf(f + g \circ L)(\mathcal{H}) = -\min(f^* \circ L^* + g^{*\vee})(\mathcal{K})$  (e.g., one of the conditions listed in Theorem 15.23, Proposition 15.24, or Fact 15.25 holds), and let  $v$  be an arbitrary solution to the dual problem (19.2). Then the (possibly empty) set of primal solutions is*

$$\text{Argmin}(f + g \circ L) = \partial f^*(L^*v) \cap L^{-1}(\partial g^*(-v)). \quad (19.4)$$

Next, we present an instance in which the primal solution is uniquely determined by a dual solution.

**Proposition 19.3** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that the following hold:*

- (i)  $\inf(f + g \circ L)(\mathcal{H}) = -\min(f^* \circ L^* + g^{*\vee})(\mathcal{K})$  (e.g., one of the conditions listed in Theorem 15.23, Proposition 15.24, or Fact 15.25 holds).
- (ii) *There exists a solution  $v$  to the dual problem (19.2) such that  $f^*$  is Gâteaux differentiable at  $L^*v$ .*

*Then either the primal problem (19.1) has no solution or it has a unique solution, namely*

$$x = \nabla f^*(L^*v). \quad (19.5)$$

*Proof.* We derive from Corollary 19.2, (ii), and Proposition 17.26(i) that  $\text{Argmin}(f + g \circ L) = \partial f^*(L^*v) \cap L^{-1}(\partial g^*(-v)) \subset \{\nabla f^*(L^*v)\}$ , which yields (19.5).  $\square$

Here is an important application of Proposition 19.3. It implies that, if the objective function of the primal problem is strongly convex, then the dual problem can be formulated in terms of a Moreau envelope (see Definition 12.20).

**Proposition 19.4** *Let  $\varphi \in \Gamma_0(\mathcal{H})$ , let  $\psi \in \Gamma_0(\mathcal{K})$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $r \in \text{sri}(\text{dom } \psi - L(\text{dom } \varphi))$ . Consider the problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \varphi(x) + \psi(Lx - r) + \frac{1}{2}\|x - z\|^2, \quad (19.6)$$

*together with the problem*

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad {}^1(\varphi^*)(L^*v + z) + \psi^*(-v) - \langle v | r \rangle, \quad (19.7)$$

*which can also be written as*

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad \frac{1}{2}\|L^*v + z\|^2 - {}^1\varphi(L^*v + z) + \psi^*(-v) - \langle v | r \rangle. \quad (19.8)$$

*Let  $v$  be a solution to (19.7). Then (19.6) has a unique solution, namely*

$$x = \text{Prox}_\varphi(L^*v + z). \quad (19.9)$$

*Proof.* The fact that (19.6) admits a unique solution follows from Definition 12.23, and the fact that (19.7) and (19.8) are identical from (14.6). Now set  $f = \varphi + (1/2)\|\cdot - z\|^2$  and  $g = \psi(\cdot - r)$ . Then (19.1) reduces to (19.6). Moreover, using Proposition 14.1 and Proposition 13.20(iii), we obtain  $f^*: u \mapsto {}^1(\varphi^*)(u + z) - (1/2)\|z\|^2$  and  $g^*: v \mapsto \psi^*(v) + \langle r | v \rangle$ , so that (19.2) reduces to (19.7). Furthermore, it follows from (14.7) that  $\nabla f^*: u \mapsto \text{Prox}_\varphi(u + z)$ . Thus, the result follows from Proposition 19.3 and Theorem 15.23.  $\square$

**Remark 19.5** The (primal) problem (19.6) involves the sum of a composite convex function and of another convex function. Such problems are not easy to solve. By contrast, the (dual) problem (19.7) involves a Moreau envelope, i.e., an everywhere defined differentiable convex function with a Lipschitz continuous gradient (see Proposition 12.29). As will be seen in Corollary 27.9, such problems are much easier to solve. This observation will be exploited in Section 27.5.

**Corollary 19.6** *Let  $\varphi \in \Gamma_0(\mathcal{H})$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\|L\| \leq 1$  and  $-r \in \text{sri}(L(\text{dom } \varphi))$ , and set  $T: \mathcal{K} \rightarrow \mathcal{K}: v \mapsto r + v - L(\text{Prox}_\varphi(L^*v + z))$ . Then  $T$  is firmly nonexpansive,  $\text{Fix } T \neq \emptyset$ , and for every  $v \in \text{Fix } T$ , the unique solution of the problem*

$$\underset{\substack{x \in \mathcal{H} \\ Lx = r}}{\text{minimize}} \quad \varphi(x) + \frac{1}{2}\|x - z\|^2 \quad (19.10)$$

*is  $x = \text{Prox}_\varphi(L^*v + z)$ .*

*Proof.* Set  $S = \{v \in \mathcal{K} \mid L(\text{Prox}_\varphi(L^*v + z)) = r\}$ . The result follows from Proposition 19.4 with  $\psi = \iota_{\{0\}}$ . Indeed, (19.6) turns into (19.10), and the function to minimize in (19.7) is  $v \mapsto {}^1(\varphi^*)(L^*v + z) - \langle v | r \rangle$ , the gradient

of which is  $v \mapsto L(\text{Prox}_\varphi(L^*v + z)) - r$ . Hence,  $S$  is the nonempty set of solutions of (19.7). Since  $v \mapsto L(\text{Prox}_\varphi(L^*v + z))$  is firmly nonexpansive by Exercise 4.7, we deduce that  $T$  is firmly nonexpansive and that  $\text{Fix } T = S$ . The conclusion follows from Proposition 19.4.  $\square$

We close this section with some applications of Proposition 19.4.

**Example 19.7** Let  $K$  be a nonempty closed convex cone of  $\mathcal{H}$ , let  $\psi \in \Gamma_0(\mathcal{K})$  be positively homogeneous, set  $D = \partial\psi(0)$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $r \in \text{sri}(\text{dom } \psi - L(K))$ . Consider the problem

$$\underset{x \in K}{\text{minimize}} \quad \psi(Lx - r) + \frac{1}{2}\|x - z\|^2, \quad (19.11)$$

together with the problem

$$\underset{-v \in D}{\text{minimize}} \quad \frac{1}{2}d_{K^\ominus}^2(L^*v + z) - \langle v \mid r \rangle. \quad (19.12)$$

Let  $v$  be a solution to (19.12). Then the unique solution to (19.11) is  $x = P_K(L^*v + z)$ .

*Proof.* This is an application of Proposition 19.4 with  $\varphi = \iota_K$ . Indeed, Example 13.3(ii) yields  $\varphi^* = \iota_{K^\ominus}$ , and Proposition 16.18 yields  $\psi^* = \sigma_D^* = \iota_D$ .  $\square$

**Example 19.8** Let  $K$  be a nonempty closed convex cone of  $\mathcal{H}$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Consider the problem

$$\underset{x \in K}{\text{minimize}} \quad \|Lx - r\| + \frac{1}{2}\|x - z\|^2, \quad (19.13)$$

together with the problem

$$\underset{v \in \mathcal{K}, \|v\| \leq 1}{\text{minimize}} \quad \frac{1}{2}d_{K^\ominus}^2(L^*v + z) - \langle v \mid r \rangle. \quad (19.14)$$

Let  $v$  be a solution to (19.14). Then the unique solution to (19.13) is  $x = P_K(L^*v + z)$ .

*Proof.* This is a special case of Example 19.7 with  $\psi = \|\cdot\|$ . Indeed,  $\text{dom } \psi - L(K) = \mathcal{K} - L(K) = \mathcal{K}$  and Example 16.25 yields  $D = \partial\psi(0) = B(0; 1)$ .  $\square$

**Example 19.9** Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$ , let  $z \in \mathcal{H}$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $0 \in \text{sri}(D - L(C))$ . Consider the best approximation problem

$$\underset{\substack{x \in C \\ Lx \in D}}{\text{minimize}} \quad \|x - z\|, \quad (19.15)$$

together with the problem

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad \frac{1}{2}\|L^*v + z\|^2 - \frac{1}{2}d_C^2(L^*v + z) + \sigma_D(-v). \quad (19.16)$$



Let  $v$  be a solution to (19.16). Then the unique solution to (19.15) is  $x = P_C(L^*v + z)$ .

*Proof.* Apply Proposition 19.4 with  $\varphi = \iota_C$ ,  $\psi = \iota_D$ , and  $r = 0$ .  $\square$

## 19.2 Parametric Duality

We explore an abstract duality framework defined on the product  $\mathcal{H} \times \mathcal{K}$ .

**Definition 19.10** Let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$ . The associated *primal problem* is

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad F(x, 0), \quad (19.17)$$

the associated *dual problem* is

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad F^*(0, v), \quad (19.18)$$

and the associated *value function* is

$$\vartheta: \mathcal{K} \rightarrow [-\infty, +\infty] : y \mapsto \inf F(\mathcal{H}, y). \quad (19.19)$$

Furthermore,  $x \in \mathcal{H}$  is a *primal solution* if it solves (19.17), and  $v \in \mathcal{K}$  is a *dual solution* if it solves (19.18).

**Proposition 19.11** Let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$  and let  $\vartheta$  be the associated value function. Then the following hold:

- (i)  $\vartheta^* = F^*(0, \cdot)$ .
- (ii)  $-\inf F^*(0, \mathcal{K}) = \vartheta^{**}(0) \leq \vartheta(0) = \inf F(\mathcal{H}, 0)$ .

*Proof.* We assume that  $F$  is proper.

(i): Apply Proposition 13.28.

(ii): The first equality follows from (i) and Proposition 13.9(i), and the second equality follows from (19.19). The inequality follows from Proposition 13.14(i) applied to  $\vartheta$ .  $\square$

The next result gives a condition under which the inequality in Proposition 19.11(ii) becomes an equality.

**Proposition 19.12** Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$  and suppose that the associated value function  $\vartheta$  is lower semicontinuous at 0 with  $\vartheta(0) \in \mathbb{R}$ . Then

$$\inf F(\mathcal{H}, 0) = -\inf F^*(0, \mathcal{K}) \in \mathbb{R}. \quad (19.20)$$

*Proof.* Since  $F$  is convex, it follows from Proposition 8.26 that  $\vartheta$  is convex. Hence, Proposition 13.38 yields  $\vartheta^{**}(0) = \vartheta(0)$  and the result follows.  $\square$

We now describe the set of dual solutions using the subdifferential of the (biconjugate of the) value function at 0.

**Proposition 19.13** *Let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$ , suppose that the associated value function  $\vartheta$  is convex and that  $\vartheta^*$  is proper, and denote the set of dual solutions by  $U$ . Then the following hold:*

- (i)  $U = \partial\vartheta^{**}(0)$ .
- (ii)  $\partial\vartheta(0) \neq \emptyset$  if and only if  $\vartheta$  is lower semicontinuous at 0 with  $\vartheta(0) \in \mathbb{R}$  and  $U \neq \emptyset$ , in which case  $U = \partial\vartheta(0)$  and  $\inf F(\mathcal{H}, 0) = -\min F^*(0, \mathcal{K}) \in \mathbb{R}$ .

*Proof.* (i): In view of Proposition 19.11(i), Theorem 16.2, and Corollary 16.24, we have  $U = \text{Argmin } \vartheta^* = (\partial\vartheta^*)^{-1}(0) = \partial\vartheta^{**}(0)$ .

(ii): Suppose that  $\partial\vartheta(0) \neq \emptyset$ . Using Proposition 16.3(i), Proposition 16.4, (i), and Proposition 19.12, we obtain  $0 \in \text{dom } \vartheta$ ,  $\vartheta^{**}(0) = \vartheta(0)$ ,  $U = \partial\vartheta^{**}(0) = \partial\vartheta(0) \neq \emptyset$ , and hence  $\inf F(\mathcal{H}, 0) = -\min F^*(0, \mathcal{K}) \in \mathbb{R}$ . Now assume that  $0 \in \text{dom } \vartheta$ , that  $\vartheta$  is lower semicontinuous at 0, and that  $U \neq \emptyset$ . By Proposition 13.38 and (i),  $\vartheta(0) = \vartheta^{**}(0)$  and  $U = \partial\vartheta^{**}(0) \neq \emptyset$ . Now take  $u \in \partial\vartheta^{**}(0)$ . Since  $\vartheta^{**} \leq \vartheta$ , it follows that  $(\forall x \in \mathcal{H}) \langle x | u \rangle \leq \vartheta^{**}(x) - \vartheta^{**}(0) \leq \vartheta(x) - \vartheta(0)$ . Therefore,  $\emptyset \neq U = \partial\vartheta^{**}(0) \subset \partial\vartheta(0)$ .  $\square$

**Proposition 19.14** *Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$  and let  $(x, v) \in \mathcal{H} \times \mathcal{K}$ . Then the following are equivalent:*

- (i)  $x$  is a primal solution,  $v$  is a dual solution, and (19.20) holds.
- (ii)  $F(x, 0) + F^*(0, v) = 0$ .
- (iii)  $(0, v) \in \partial F(x, 0)$ .
- (iv)  $(x, 0) \in \partial F^*(0, v)$ .

*Proof.* (i) $\Rightarrow$ (ii):  $F(x, 0) = \inf F(\mathcal{H}, 0) = -\inf F^*(0, \mathcal{K}) = -F^*(0, v) \in \mathbb{R}$ . Hence  $F(x, 0) + F^*(0, v) = 0$ .

(ii) $\Rightarrow$ (i): By Proposition 19.11(ii),

$$-F^*(0, v) \leq -\inf F^*(0, \mathcal{K}) \leq \inf F(\mathcal{H}, 0) \leq F(x, 0). \quad (19.21)$$

(ii) $\Leftrightarrow$ (iii) $\Leftrightarrow$ (iv): It follows from Theorem 16.23 that (ii)  $\Leftrightarrow F(x, 0) + F^*(0, v) = \langle (x, 0) | (0, v) \rangle \Leftrightarrow$  (iii)  $\Leftrightarrow$  (iv).  $\square$

**Definition 19.15** Let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$ . The *Lagrangian* of  $F$  is the function

$$\mathcal{L}: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]: (x, v) \mapsto \inf_{y \in \mathcal{K}} (F(x, y) + \langle y | v \rangle). \quad (19.22)$$

Moreover,  $(x, v) \in \mathcal{H} \times \mathcal{K}$  is a *saddle point* of  $\mathcal{L}$  if

$$\sup \mathcal{L}(x, \mathcal{K}) = \mathcal{L}(x, v) = \inf \mathcal{L}(\mathcal{H}, v). \quad (19.23)$$

**Proposition 19.16** *Let  $F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]$  and let  $\mathcal{L}: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]$  be its Lagrangian. Then the following hold:*

- (i) *For every  $x \in \mathcal{H}$ ,  $\mathcal{L}(x, \cdot)$  is upper semicontinuous and concave.*
- (ii) *Suppose that  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ . Then  $(\forall x \in \mathcal{H}) \sup \mathcal{L}(x, \mathcal{K}) = F(x, 0)$ .*
- (iii) *Suppose that  $F$  is convex. Then, for every  $v \in \mathcal{K}$ ,  $\mathcal{L}(\cdot, v)$  is convex.*
- (iv)  $(\forall v \in \mathcal{K}) \inf \mathcal{L}(\mathcal{H}, v) = -F^*(0, -v)$ .

*Proof.* (i): Let  $x \in \mathcal{H}$  and set  $\varphi = F(x, \cdot)$ . Then, for every  $v \in \mathcal{K}$ ,

$$\mathcal{L}(x, v) = \inf_{y \in \mathcal{K}} (F(x, y) - \langle y | -v \rangle) = -\sup_{y \in \mathcal{K}} (\langle y | -v \rangle - \varphi(y)) = -\varphi^*(-v). \quad (19.24)$$

Since  $\varphi^*$  is lower semicontinuous and convex by Proposition 13.11, we deduce that  $\mathcal{L}(x, \cdot) = -\varphi^{*\vee}$  is upper semicontinuous and concave.

(ii): Let  $x \in \mathcal{H}$  and set  $\varphi = F(x, \cdot)$ . Then either  $\varphi \equiv +\infty$  or  $\varphi \in \Gamma_0(\mathcal{K})$ . It follows from Corollary 13.33 that  $\varphi^{**} = \varphi$ . In view of (19.24) and Proposition 13.9(i), we obtain  $\sup \mathcal{L}(x, \mathcal{K}) = -\inf \varphi^*(\mathcal{K}) = \varphi^{**}(0) = \varphi(0) = F(x, 0)$ .

(iii): Let  $v \in \mathcal{K}$ . Since  $(x, y) \mapsto F(x, y) + \langle y | v \rangle$  is convex, the conclusion follows from Proposition 8.26.

(iv): For every  $x \in \mathcal{H}$ ,  $\mathcal{L}(x, v) = -\sup_{y \in \mathcal{K}} (\langle x | 0 \rangle + \langle y | -v \rangle - F(x, y))$ . Therefore,  $\inf \mathcal{L}(\mathcal{H}, v) = -\sup_{x \in \mathcal{H}} \sup_{y \in \mathcal{K}} (\langle (x, y) | (0, -v) \rangle - F(x, y)) = -F^*(0, -v)$ .  $\square$

**Corollary 19.17** *Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ , let  $\mathcal{L}$  be its Lagrangian, and let  $(x, v) \in \mathcal{H} \times \mathcal{K}$ . Then the following are equivalent:*

- (i)  *$x$  is a primal solution,  $v$  is a dual solution, and  $\inf F(\mathcal{H}, 0) = -\inf F^*(0, \mathcal{K}) \in \mathbb{R}$ .*
- (ii)  $F(x, 0) + F^*(0, v) = 0$ .
- (iii)  $(0, v) \in \partial F(x, 0)$ .
- (iv)  $(x, 0) \in \partial F^*(0, v)$ .
- (v)  $(x, -v)$  is a saddle point of  $\mathcal{L}$ .

*Proof.* (i) $\Leftrightarrow$ (ii) $\Leftrightarrow$ (iii) $\Leftrightarrow$ (iv): Proposition 19.14.

(ii) $\Rightarrow$ (v): By Proposition 19.16(iv)&(ii),

$$F(x, 0) = -F^*(0, v) = \inf \mathcal{L}(\mathcal{H}, -v) \leq \mathcal{L}(x, -v) \leq \sup \mathcal{L}(x, \mathcal{K}) = F(x, 0). \quad (19.25)$$

(v) $\Rightarrow$ (ii): Using Proposition 19.16(iv)&(ii), we get

$$-F^*(0, v) = \inf \mathcal{L}(\mathcal{H}, -v) = \mathcal{L}(x, -v) = \sup \mathcal{L}(x, \mathcal{K}) = F(x, 0). \quad (19.26)$$

Hence, since  $F$  is proper,  $-F^*(0, v) = F(x, 0) \in \mathbb{R}$ .  $\square$

The following proposition illustrates how a bivariate function can be associated with a minimization problem, its second variable playing the role of a perturbation. In this particular case, we recover the Fenchel–Rockafellar duality framework discussed in Section 15.3 and in Section 19.1.

**Proposition 19.18** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{dom } g \cap L(\text{dom } f) \neq \emptyset$ . Set*

$$F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]: (x, y) \mapsto f(x) + g(Lx - y). \quad (19.27)$$

*Then the following hold:*

- (i)  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ .
- (ii) *The primal problem is*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(Lx). \quad (19.28)$$

- (iii) *The dual problem is*

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad f^*(L^*v) + g^*(-v). \quad (19.29)$$

- (iv) *The Lagrangian is*

$$\begin{aligned} \mathcal{L}: \mathcal{H} \times \mathcal{K} &\rightarrow [-\infty, +\infty] \\ (x, v) &\mapsto \begin{cases} -\infty, & \text{if } x \in \text{dom } f \text{ and } v \notin \text{dom } g^*; \\ f(x) + \langle Lx \mid v \rangle - g^*(v), & \text{if } x \in \text{dom } f \text{ and } v \in \text{dom } g^*; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \end{aligned} \quad (19.30)$$

- (v) *Suppose that the optimal values  $\mu$  of (19.28) and  $\mu^*$  of (19.29) satisfy  $\mu = -\mu^* \in \mathbb{R}$ , and let  $(\bar{x}, \bar{v}) \in \mathcal{H} \times \mathcal{K}$ . Then  $(\bar{x}, \bar{v})$  is a saddle point of  $\mathcal{L}$  if and only if*

$$-L^*\bar{v} \in \partial f(\bar{x}) \quad \text{and} \quad \bar{v} \in \partial g(L\bar{x}). \quad (19.31)$$

*Proof.* (i): This follows easily from the assumptions on  $f$ ,  $g$ , and  $L$ .

(ii): This follows from (19.17) and (19.27).

(iii): Let  $v \in \mathcal{K}$ . Then

$$\begin{aligned} F^*(0, v) &= \sup_{(x, y) \in \mathcal{H} \times \mathcal{K}} \left( \langle y \mid v \rangle - f(x) - g(Lx - y) \right) \\ &= \sup_{x \in \mathcal{H}} \left( \langle x \mid L^*v \rangle - f(x) + \sup_{z \in \mathcal{K}} \left( \langle z \mid -v \rangle - g(z) \right) \right) \\ &= (f^* \circ L^*)(v) + g^*(-v). \end{aligned} \quad (19.32)$$

Hence, the result follows from (19.18).

(iv): For every  $(x, v) \in \mathcal{H} \times \mathcal{K}$ , we derive from (19.22) and (19.27) that

$$\mathcal{L}(x, v) = f(x) + \inf_{y \in \mathcal{K}} (g(Lx - y) + \langle y \mid v \rangle)$$

$$\begin{aligned}
&= \begin{cases} f(x) + \langle Lx \mid v \rangle + \inf_{z \in \mathcal{K}} (g(z) - \langle z \mid v \rangle), & \text{if } x \in \text{dom } f; \\ +\infty, & \text{if } x \notin \text{dom } f \end{cases} \\
&= \begin{cases} f(x) + \langle Lx \mid v \rangle - \sup_{z \in \mathcal{K}} (\langle z \mid v \rangle - g(z)), & \text{if } x \in \text{dom } f; \\ +\infty, & \text{if } x \notin \text{dom } f \end{cases} \\
&= \begin{cases} f(x) + \langle Lx \mid v \rangle - g^*(v), & \text{if } x \in \text{dom } f; \\ +\infty, & \text{if } x \notin \text{dom } f, \end{cases} \quad (19.33)
\end{aligned}$$

which yields (19.30).

(v): Since  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ , we derive from Corollary 19.17 that  $(\bar{x}, \bar{v})$  is a saddle point of  $\mathcal{L}$  if and only if  $\bar{x}$  is a solution to (19.28) and  $-\bar{v}$  is a solution to (19.29). Hence, the conclusion follows from Theorem 19.1.  $\square$

## 19.3 Minimization under Equality Constraints

In this section, we apply the setting of Section 19.2 to convex optimization problems with affine equality constraints.

**Proposition 19.19** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $r \in L(\text{dom } f)$ . Set*

$$F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]: (x, y) \mapsto \begin{cases} f(x), & \text{if } Lx = y + r; \\ +\infty, & \text{if } Lx \neq y + r. \end{cases} \quad (19.34)$$

*Then the following hold:*

- (i)  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ .
- (ii) *The primal problem is*

$$\underset{\substack{x \in \mathcal{H} \\ Lx=r}}{\text{minimize}} \quad f(x). \quad (19.35)$$

- (iii) *The dual problem is*

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad f^*(L^*v) - \langle v \mid r \rangle. \quad (19.36)$$

- (iv) *The Lagrangian is*

$$\begin{aligned}
&\mathcal{L}: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty] \\
&(x, v) \mapsto \begin{cases} f(x) + \langle Lx - r \mid v \rangle, & \text{if } x \in \text{dom } f; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \quad (19.37)
\end{aligned}$$

- (v) *Suppose that the optimal values  $\mu$  of (19.35) and  $\mu^*$  of (19.36) satisfy  $\mu = -\mu^* \in \mathbb{R}$ , and let  $(\bar{x}, \bar{v}) \in \mathcal{H} \times \mathcal{K}$ . Then  $(\bar{x}, \bar{v})$  is a saddle point of*

$\mathcal{L}$  if and only if  $-L^*\bar{v} \in \partial f(\bar{x})$  and  $L\bar{x} = r$ . In this case,

$$f(\bar{x}) = \min_{x \in \mathcal{H}} f(L^{-1}(\{r\})) = \min \mathcal{L}(\mathcal{H}, \bar{v}). \quad (19.38)$$

*Proof.* All the results, except (19.38), are obtained by setting  $g = \iota_{\{r\}}$  in Proposition 19.18. To prove (19.38), note that Corollary 19.17 and Proposition 19.16 yield  $f(\bar{x}) = F(\bar{x}, 0) = -F^*(0, -\bar{v}) = \inf \mathcal{L}(\mathcal{H}, v) = \mathcal{L}(\bar{x}, \bar{v})$ .  $\square$

**Remark 19.20** In the setting of Proposition 19.19, let  $(\bar{x}, \bar{v})$  be a saddle point of the Lagrangian (19.37). Then  $\bar{v}$  is called a *Lagrange multiplier* associated with the solution  $\bar{x}$  to (19.35). In view of (19.38) and (19.37),  $\bar{x}$  solves the unconstrained minimization problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \langle x \mid L^*\bar{v} \rangle. \quad (19.39)$$

An important application of Proposition 19.19 is in the context of problems involving finitely many scalar affine equalities.

**Corollary 19.21** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(u_i)_{i \in I} \in \mathcal{H}^m$ , and let  $(\rho_i)_{i \in I} \in \{(\langle x \mid u_i \rangle)_{i \in I} \mid x \in \text{dom } f\}$ . Set

$$F: \mathcal{H} \times \mathbb{R}^m \rightarrow ]-\infty, +\infty] \quad (19.40)$$

$$(x, (\eta_i)_{i \in I}) \mapsto \begin{cases} f(x), & \text{if } (\forall i \in I) \quad \langle x \mid u_i \rangle = \eta_i + \rho_i; \\ +\infty, & \text{otherwise.} \end{cases} \quad (19.41)$$

Then the following hold:

- (i)  $F \in \Gamma_0(\mathcal{H} \times \mathbb{R}^m)$ .
- (ii) The primal problem is

$$\underset{\substack{x \in \mathcal{H} \\ \langle x \mid u_1 \rangle = \rho_1, \dots, \langle x \mid u_m \rangle = \rho_m}}{\text{minimize}} \quad f(x). \quad (19.42)$$

- (iii) The dual problem is

$$\underset{(\nu_i)_{i \in I} \in \mathbb{R}^m}{\text{minimize}} \quad f^* \left( \sum_{i \in I} \nu_i u_i \right) - \sum_{i \in I} \nu_i \rho_i. \quad (19.43)$$

- (iv) The Lagrangian is

$$\mathcal{L}: \mathcal{H} \times \mathbb{R}^m \rightarrow [-\infty, +\infty] \\ (x, (\nu_i)_{i \in I}) \mapsto \begin{cases} f(x) + \sum_{i \in I} \nu_i (\langle x \mid u_i \rangle - \rho_i), & \text{if } x \in \text{dom } f; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \quad (19.44)$$

(v) Suppose that the optimal values  $\mu$  of (19.42) and  $\mu^*$  of (19.43) satisfy  $\mu = -\mu^* \in \mathbb{R}$ , and let  $(\bar{x}, (\bar{\nu}_i)_{i \in I}) \in \mathcal{H} \times \mathbb{R}^m$ . Then  $(\bar{x}, (\bar{\nu}_i)_{i \in I})$  is a saddle point of  $\mathcal{L}$  if and only if

$$-\sum_{i \in I} \bar{\nu}_i u_i \in \partial f(\bar{x}) \quad \text{and} \quad (\forall i \in I) \quad \langle \bar{x} \mid u_i \rangle = \rho_i, \quad (19.45)$$

in which case  $\bar{x}$  is a solution to both (19.42) and the unconstrained minimization problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \sum_{i \in I} \bar{\nu}_i \langle x \mid u_i \rangle. \quad (19.46)$$

*Proof.* Apply Proposition 19.19 with  $\mathcal{K} = \mathbb{R}^m$ ,  $r = (\rho_i)_{i \in I}$ , and  $L: x \mapsto (\langle x \mid u_i \rangle)_{i \in I}$ .  $\square$

## 19.4 Minimization under Inequality Constraints

To use a fairly general notion of inequality, we require the following notion of convexity with respect to a cone.

**Definition 19.22** Let  $K$  be a nonempty closed convex cone in  $\mathcal{K}$  and let  $R: \mathcal{H} \rightarrow \mathcal{K}$ . Then  $R$  is *convex with respect to  $K$*  if

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H})(\forall \alpha \in ]0, 1[) \\ R(\alpha x + (1 - \alpha)y) - \alpha Rx - (1 - \alpha)Ry \in K. \quad (19.47)$$

**Proposition 19.23** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $K$  be a nonempty closed convex cone in  $\mathcal{K}$ , and let  $R: \mathcal{H} \rightarrow \mathcal{K}$  be continuous, convex with respect to  $K$ , and such that  $K \cap R(\text{dom } f) \neq \emptyset$ . Set

$$F: \mathcal{H} \times \mathcal{K} \rightarrow ]-\infty, +\infty]: (x, y) \mapsto \begin{cases} f(x), & \text{if } Rx \in y + K; \\ +\infty, & \text{if } Rx \notin y + K. \end{cases} \quad (19.48)$$

Then the following hold:

- (i)  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$ .
- (ii) The primal problem is

$$\underset{\substack{x \in \mathcal{H} \\ Rx \in K}}{\text{minimize}} \quad f(x). \quad (19.49)$$

- (iii) The dual problem is

$$\underset{v \in K^\oplus}{\text{minimize}} \quad \varphi(v), \quad \text{where} \quad \varphi: v \mapsto \sup_{x \in \mathcal{H}} (\langle Rx \mid v \rangle - f(x)). \quad (19.50)$$

(iv) The Lagrangian is

$$\mathcal{L}: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]$$

$$(x, v) \mapsto \begin{cases} -\infty, & \text{if } x \in \text{dom } f \text{ and } v \notin K^\ominus; \\ f(x) + \langle Rx \mid v \rangle, & \text{if } x \in \text{dom } f \text{ and } v \in K^\ominus; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \quad (19.51)$$

(v) Let  $(\bar{x}, \bar{v}) \in \mathcal{H} \times \mathcal{K}$ . Then  $(\bar{x}, \bar{v})$  is a saddle point of  $\mathcal{L}$  if and only if

$$\begin{cases} \bar{x} \in \text{dom } f, \\ R\bar{x} \in K, \\ \bar{v} \in K^\ominus, \\ f(\bar{x}) = \inf_{x \in \mathcal{H}} (f(x) + \langle Rx \mid \bar{v} \rangle), \end{cases} \quad (19.52)$$

in which case  $\langle R\bar{x} \mid \bar{v} \rangle = 0$  and  $\bar{x}$  is a primal solution.

*Proof.* (i): The assumptions imply that  $F$  is proper and that  $F_1: (x, y) \mapsto f(x)$  is lower semicontinuous and convex. Hence, by Corollary 9.4, it remains to check that  $F_2: (x, y) \mapsto \iota_K(Rx - y)$  is likewise, which amounts to showing that  $C = \{(x, y) \in \mathcal{H} \times \mathcal{K} \mid Rx - y \in K\}$  is closed and convex. The closedness of  $C$  follows easily from the continuity of  $R$  and the closedness of  $K$ . Now take  $(x_1, y_1)$  and  $(x_2, y_2)$  in  $C$  and  $\alpha \in ]0, 1[$ . Then  $Rx_1 - y_1 \in K$ ,  $Rx_2 - y_2 \in K$ , and, by convexity of  $K$ ,  $\alpha Rx_1 + (1 - \alpha)Rx_2 - (\alpha y_1 + (1 - \alpha)y_2) = \alpha(Rx_1 - y_1) + (1 - \alpha)(Rx_2 - y_2) \in K$ . On the other hand, it follows from (19.47) that  $R(\alpha x_1 + (1 - \alpha)x_2) - \alpha Rx_1 - (1 - \alpha)Rx_2 \in K$ . Adding these two inclusions yields  $R(\alpha x_1 + (1 - \alpha)x_2) - (\alpha y_1 + (1 - \alpha)y_2) \in K + K = K$ . We conclude that  $\alpha(x_1, y_1) + (1 - \alpha)(x_2, y_2) \in C$  and hence that  $C$  is convex.

(ii): This follows from (19.17) and (19.48).

(iii): Let  $v \in \mathcal{K}$ . Then

$$\begin{aligned} F^*(0, v) &= \sup_{(x, y) \in \mathcal{H} \times \mathcal{K}} (\langle y \mid v \rangle - f(x) - \iota_K(Rx - y)) \\ &= \sup_{x \in \mathcal{H}} \left( \langle Rx \mid v \rangle - f(x) + \sup_{\substack{y \in \mathcal{K} \\ Rx - y \in K}} \langle Rx - y \mid -v \rangle \right) \\ &= \sup_{x \in \mathcal{H}} \left( \langle Rx \mid v \rangle - f(x) + \sup_{z \in K} \langle z \mid -v \rangle \right) \\ &= \sup_{x \in \mathcal{H}} (\langle Rx \mid v \rangle - f(x)) + \iota_{K^\ominus}(-v) \\ &= \sup_{x \in \mathcal{H}} (\langle Rx \mid v \rangle - f(x)) + \iota_{K^\oplus}(v). \end{aligned} \quad (19.53)$$

Hence, the result follows from (19.18).

(iv): Let  $(x, v) \in \mathcal{H} \times \mathcal{K}$ . Then (19.22) and (19.48) yield



$$\mathcal{L}(x, v) = \inf_{y \in \mathcal{K}} (f(x) + \iota_K(Rx - y) + \langle y | v \rangle). \quad (19.54)$$

If  $x \notin \text{dom } f$ , then, for every  $y \in \mathcal{K}$ ,  $f(x) + \iota_K(Rx - y) + \langle y | v \rangle = +\infty$  and, therefore,  $\mathcal{L}(x, v) = +\infty$ . Now suppose that  $x \in \text{dom } f$ . Then  $f(x) \in \mathbb{R}$  and we derive from (19.54) that

$$\begin{aligned} \mathcal{L}(x, v) &= f(x) + \inf_{\substack{y \in \mathcal{K} \\ Rx - y \in K}} \langle y | v \rangle \\ &= f(x) + \langle Rx | v \rangle - \sup_{z \in K} \langle z | v \rangle \\ &= f(x) + \langle Rx | v \rangle - \iota_{K^\ominus}(v), \end{aligned} \quad (19.55)$$

which yields (19.51).

(v): We derive from (19.48) and (19.53) that

$$\begin{aligned} F(\bar{x}, 0) + F^*(0, -\bar{v}) = 0 &\Leftrightarrow \begin{cases} \bar{x} \in \text{dom } f, \\ R\bar{x} \in K, \\ -\bar{v} \in K^\oplus, \\ f(\bar{x}) + \sup_{x \in \mathcal{H}} (\langle Rx | -\bar{v} \rangle - f(x)) = 0 \end{cases} \\ &\Leftrightarrow (19.52). \end{aligned} \quad (19.56)$$

Using the equivalence (v) $\Leftrightarrow$ (ii) in Corollary 19.17, we obtain the first result. Next, we derive from (19.52) that  $\langle R\bar{x} | \bar{v} \rangle \leq 0$  and  $f(\bar{x}) \leq f(\bar{x}) + \langle R\bar{x} | \bar{v} \rangle$ . Hence, since  $f(\bar{x}) \in \mathbb{R}$ ,  $\langle R\bar{x} | \bar{v} \rangle = 0$ . Finally,  $\bar{x}$  is a primal solution by Corollary 19.17.  $\square$

**Remark 19.24** In the setting of Proposition 19.23, suppose that  $(\bar{x}, \bar{v})$  is a saddle point of the Lagrangian (19.51). Then  $\bar{v}$  is called a *Lagrange multiplier* associated with the solution  $\bar{x}$  to (19.49). It follows from Proposition 19.23(v) that  $\bar{x}$  solves the unconstrained minimization problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \langle Rx | \bar{v} \rangle. \quad (19.57)$$

Conditions for the existence of Lagrange multipliers will be discussed in Section 26.3, in Section 26.4, and in Section 26.5. For specific examples, see Chapter 28.

**Example 19.25** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $z \in \mathcal{H}$ , and let  $K$  be a closed convex cone in  $\mathcal{H}$  such that  $z \in (\text{dom } f) - K$ . Set

$$F: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty] : (x, y) \mapsto \begin{cases} f(x), & \text{if } x \in y + z + K; \\ +\infty, & \text{if } x \notin y + z + K. \end{cases} \quad (19.58)$$

Then the following hold:

(i)  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ .

(ii) The primal problem is

$$\underset{x \in z+K}{\text{minimize}} \quad f(x). \quad (19.59)$$

(iii) The dual problem is

$$\underset{u \in K^\oplus}{\text{minimize}} \quad f^*(u) - \langle z \mid u \rangle. \quad (19.60)$$

(iv) The Lagrangian is

$$\begin{aligned} \mathcal{L}: \mathcal{H} \times \mathcal{H} &\rightarrow [-\infty, +\infty] \\ (x, u) &\mapsto \begin{cases} -\infty, & \text{if } x \in \text{dom } f \text{ and } u \notin K^\ominus; \\ f(x) + \langle x - z \mid u \rangle, & \text{if } x \in \text{dom } f \text{ and } u \in K^\ominus; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \end{aligned} \quad (19.61)$$

(v) Let  $(\bar{x}, \bar{u}) \in \mathcal{H} \times \mathcal{H}$ . Then  $(\bar{x}, \bar{u})$  is a saddle point of  $\mathcal{L}$  if and only if

$$\begin{cases} \bar{x} \in (z + K) \cap \text{dom } f, \\ \bar{u} \in K^\ominus, \\ \bar{x} \in \text{Argmin}(f + \langle \cdot \mid \bar{u} \rangle), \end{cases} \quad (19.62)$$

in which case  $\langle \bar{x} \mid \bar{u} \rangle = \langle z \mid \bar{u} \rangle$  and  $\bar{x}$  is a primal solution.

*Proof.* Apply Proposition 19.23 to  $\mathcal{K} = \mathcal{H}$  and  $R: x \mapsto x - z$ . □

**Example 19.26** Suppose that  $\mathcal{H} = \mathbb{R}^2$ , let  $\phi \in \Gamma_0(\mathbb{R})$  be such that  $0 \in \text{dom } \phi$ , set  $\gamma = \inf \phi(\mathbb{R})$ , and set

$$\begin{aligned} F: \mathcal{H} \times \mathbb{R} &\rightarrow ]-\infty, +\infty] \\ ((\xi_1, \xi_2), y) &\mapsto \begin{cases} \phi(\xi_2), & \text{if } \|(\xi_1, \xi_2)\| - \xi_1 \leq y; \\ +\infty, & \text{otherwise.} \end{cases} \end{aligned} \quad (19.63)$$

Then the following hold:

(i)  $F \in \Gamma_0(\mathcal{H} \times \mathbb{R})$ .

(ii) The primal problem is

$$\underset{\substack{(\xi_1, \xi_2) \in \mathcal{H} \\ \|(\xi_1, \xi_2)\| \leq \xi_1}}{\text{minimize}} \quad \phi(\xi_2), \quad (19.64)$$

the optimal value of (19.64) is  $\mu = \phi(0)$ , and the set of primal solutions is  $\mathbb{R}_+ \times \{0\}$ .

(iii) The dual problem is

$$\underset{v \in \mathbb{R}_-}{\text{minimize}} \quad \sup_{(\xi_1, \xi_2) \in \mathcal{H}} \left( v(\|(\xi_1, \xi_2)\| - \xi_1) - \phi(\xi_2) \right), \quad (19.65)$$

the optimal value of (19.65) is  $\mu^* = -\gamma$ , and the set of dual solutions is  $\mathbb{R}_-$ .

(iv) The Lagrangian is

$$\mathcal{L}: \mathcal{H} \times \mathbb{R} \rightarrow [-\infty, +\infty]$$

$$((\xi_1, \xi_2), v) \mapsto \begin{cases} -\infty, & \text{if } \xi_2 \in \text{dom } \phi \text{ and } v < 0; \\ \phi(\xi_2) + v(\|(\xi_1, \xi_2)\| - \xi_1), & \text{if } \xi_2 \in \text{dom } \phi \text{ and } v \geq 0; \\ +\infty, & \text{if } \xi_2 \notin \text{dom } \phi. \end{cases} \quad (19.66)$$

(v) The Lagrangian  $\mathcal{L}$  admits a saddle point if and only if  $\phi(0) = \gamma$ , in which case the set of saddle points is  $(\mathbb{R}_+ \times \{0\}) \times \mathbb{R}_+$ .

(vi) The value function (see (19.19)) is

$$\vartheta: \mathbb{R} \rightarrow [-\infty, +\infty] : y \mapsto \begin{cases} \gamma, & \text{if } y > 0; \\ \phi(0), & \text{if } y = 0; \\ +\infty, & \text{if } y < 0. \end{cases} \quad (19.67)$$

*Proof.* Set  $f: \mathcal{H} \rightarrow ]-\infty, +\infty] : (\xi_1, \xi_2) \mapsto \phi(\xi_2)$ . Then  $f \in \Gamma_0(\mathcal{H})$ . Set  $\mathcal{K} = \mathbb{R}$ ,  $K = \mathbb{R}_-$ , and  $R: \mathcal{H} \rightarrow \mathbb{R} : (\xi_1, \xi_2) \mapsto \|(\xi_1, \xi_2)\| - \xi_1$ . Then  $(0, 0) \in \mathbb{R} \times \text{dom } \phi = \text{dom } f$  and hence  $0 \in K \cap R(\text{dom } f)$ . Moreover, the bivariate function  $F$  in (19.48) is that deriving from (19.63).

(i): Proposition 19.23(i).

(ii): Proposition 19.23(ii) yields (19.64). Now let  $(\xi_1, \xi_2) \in \mathcal{H}$ . Since  $0 \leq \|(\xi_1, \xi_2)\| - \xi_1$ , it follows that  $\|(\xi_1, \xi_2)\| \leq \xi_1 \Leftrightarrow \|(\xi_1, \xi_2)\| = \xi_1 \Leftrightarrow (\xi_1, \xi_2) \in \mathbb{R}_+ \times \{0\}$ , in which case  $f(\xi_1, \xi_2) = \phi(0)$ .

(iii): Proposition 19.23(iii) yields (19.65). Let us determine

$$\varphi: \mathbb{R}_- \rightarrow ]-\infty, +\infty] : v \mapsto \sup_{(\xi_1, \xi_2) \in \mathcal{H}} \left( v(\|(\xi_1, \xi_2)\| - \xi_1) - \phi(\xi_2) \right). \quad (19.68)$$

Let  $v \in \mathbb{R}_-$  and let  $(\xi_1, \xi_2) \in \mathcal{H}$ . Then  $v(\|(\xi_1, \xi_2)\| - \xi_1) - \phi(\xi_2) \leq -\phi(\xi_2) \leq -\gamma$  and thus  $\varphi(v) \leq -\gamma$ . On the other hand, since  $\|(\xi_1, \xi_2)\| - \xi_1 \rightarrow 0$  as  $\xi_1 \rightarrow +\infty$ , we deduce that  $\varphi(v) = -\gamma$ .

(iv): Proposition 19.23(iv).

(v): This follows from Proposition 19.23(v) and computation of  $\varphi$  in (iii).

(vi): The details are left as Exercise 19.9.  $\square$

**Remark 19.27** Consider the setting of Example 19.26. If  $\gamma < \phi(0)$ , then  $\mathcal{L}$  has no saddle point and it follows from Corollary 19.17 that the duality gap is

nonzero, even though primal and dual solutions exist. The choice  $\phi = \exp^\vee$ , for which  $\gamma = 0 < 1 = \phi(0)$ , then leads to *Duffin's duality gap*.

The application presented next is referred to as the *convex programming problem*.

**Corollary 19.28** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $m$  and  $p$  be integers such that  $m > p > 0$ , set  $I = \{1, \dots, p\}$ , set  $J = \{p+1, \dots, m\}$ , let  $(g_i)_{i \in I}$  be real-valued functions in  $\Gamma_0(\mathcal{H})$ , suppose that  $(u_j)_{j \in J}$  are vectors in  $\mathcal{H}$ , and let  $(\rho_j)_{j \in J} \in \mathbb{R}^{m-p}$  be such that*

$$\left\{ x \in \text{dom } f \mid \max_{i \in I} g_i(x) \leq 0 \text{ and } \max_{j \in J} |\langle x \mid u_j \rangle - \rho_j| = 0 \right\} \neq \emptyset. \quad (19.69)$$

Set

$$F: \mathcal{H} \times \mathbb{R}^m \rightarrow ]-\infty, +\infty] \quad (19.70)$$

$$(x, (\eta_i)_{i \in I \cup J}) \mapsto \begin{cases} f(x), & \text{if } \begin{cases} (\forall i \in I) \ g_i(x) \leq \eta_i, \\ (\forall j \in J) \ \langle x \mid u_j \rangle = \eta_j + \rho_j; \end{cases} \\ +\infty, & \text{otherwise.} \end{cases} \quad (19.71)$$

Then the following hold:

- (i)  $F \in \Gamma_0(\mathcal{H} \times \mathbb{R}^m)$ .
- (ii) The primal problem is

$$\begin{aligned} & \underset{\substack{x \in \mathcal{H} \\ g_1(x) \leq 0, \dots, g_p(x) \leq 0 \\ \langle x \mid u_{p+1} \rangle = \rho_{p+1}, \dots, \langle x \mid u_m \rangle = \rho_m}}{\text{minimize}} & f(x). \end{aligned} \quad (19.72)$$

- (iii) The dual problem is

$$\begin{aligned} & \underset{\substack{(\nu_i)_{i \in I} \in \mathbb{R}_+^p \\ (\nu_j)_{j \in J} \in \mathbb{R}^{m-p}}}{\text{minimize}} & \sup_{x \in \mathcal{H}} \left( \sum_{i \in I} \nu_i g_i(x) + \sum_{j \in J} \nu_j (\langle x \mid u_j \rangle - \rho_j) - f(x) \right). \end{aligned} \quad (19.73)$$

- (iv) The Lagrangian is

$$\mathcal{L}: \mathcal{H} \times \mathbb{R}^m \rightarrow [-\infty, +\infty]$$

$$(x, (\nu_i)_{i \in I \cup J}) \mapsto \begin{cases} -\infty, & \text{if } x \in \text{dom } f \text{ and } (\nu_i)_{i \in I} \notin \mathbb{R}_+^p; \\ f(x) + \sum_{i \in I} \nu_i g_i(x) + \sum_{j \in J} \nu_j (\langle x \mid u_j \rangle - \rho_j), & \\ & \text{if } x \in \text{dom } f \text{ and } (\nu_i)_{i \in I} \in \mathbb{R}_+^p; \\ +\infty, & \text{if } x \notin \text{dom } f. \end{cases} \quad (19.74)$$

(v) Suppose that  $(\bar{x}, (\bar{\nu}_i)_{i \in I \cup J})$  is a saddle point of  $\mathcal{L}$ . Then  $\bar{x}$  is a solution to both (19.72) and the unconstrained minimization problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \sum_{i \in I} \bar{\nu}_i g_i(x) + \sum_{j \in J} \bar{\nu}_j \langle x \mid u_j \rangle. \quad (19.75)$$

Moreover,

$$(\forall i \in I) \quad \begin{cases} \bar{\nu}_i = 0, & \text{if } g_i(\bar{x}) < 0; \\ \bar{\nu}_i < 0, & \text{if } g_i(\bar{x}) = 0. \end{cases} \quad (19.76)$$

*Proof.* Apply Proposition 19.23 with  $\mathcal{K} = \mathbb{R}^m$ ,  $K = \mathbb{R}_+^p \times \{0\}^{m-p}$  (hence  $K^\ominus = \mathbb{R}_+^p \times \mathbb{R}^{m-p}$  and  $K^\oplus = \mathbb{R}_-^p \times \mathbb{R}^{m-p}$ ), and  $R: x \mapsto (g_1(x), \dots, g_p(x), \langle x \mid u_{p+1} \rangle - \rho_{p+1}, \dots, \langle x \mid u_m \rangle - \rho_m)$ . Note that the continuity of  $R$  follows from Corollary 8.30(ii).  $\square$

**Remark 19.29** As in Remark 19.24, the parameters  $(\bar{\nu}_i)_{i \in I}$  in Corollary 19.28(v) are the Lagrange multipliers associated with the solution  $\bar{x}$  to (19.72). Condition (19.76) on the multipliers  $(\bar{\nu}_i)_{i \in I}$  corresponding to the inequality constraints is a *complementary slackness* condition.

## Exercises

**Exercise 19.1** Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$  and suppose that

$$\inf F(\mathcal{H}, 0) = -\inf F^*(0, \mathcal{K}) \in \mathbb{R}. \quad (19.77)$$

Show that the associated value function defined in (19.19) is lower semicontinuous at 0.

**Exercise 19.2** Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{K})$  with associated value function  $\vartheta$ . Show that  $\inf F(\mathcal{H}, 0) = -\inf F^*(0, \mathcal{K}) \in \mathbb{R}$  if and only if  $\vartheta(0) \in \mathbb{R}$  and  $\vartheta$  is lower semicontinuous at 0.

**Exercise 19.3** Let  $\mathcal{L}: \mathcal{H} \times \mathcal{K} \rightarrow [-\infty, +\infty]$ , let  $(\bar{x}, \bar{v}) \in \mathcal{H} \times \mathcal{K}$ , and denote the set of saddle points of  $\mathcal{L}$  by  $S$ . Prove the following:

- (i)  $\sup_{v \in \mathcal{K}} \inf_{x \in \mathcal{H}} \mathcal{L}(x, v) \leq \inf_{x \in \mathcal{H}} \sup_{v \in \mathcal{K}} \mathcal{L}(x, v)$ .  
(ii)  $(\bar{x}, \bar{v}) \in S \Leftrightarrow \inf_{x \in \mathcal{H}} \mathcal{L}(x, \bar{v}) = \sup_{v \in \mathcal{K}} \mathcal{L}(\bar{x}, v)$ .  
(iii) Suppose that  $(\bar{x}, \bar{v}) \in S$ . Then

$$\sup_{v \in \mathcal{K}} \inf_{x \in \mathcal{H}} \mathcal{L}(x, v) = \mathcal{L}(\bar{x}, \bar{v}) = \inf_{x \in \mathcal{H}} \sup_{v \in \mathcal{K}} \mathcal{L}(x, v). \quad (19.78)$$

**Exercise 19.4** Let  $f: \mathcal{H} \rightarrow \mathbb{R}$ , and let  $K \in \{\{0\}, \mathbb{R}_+, \mathbb{R}_-, \mathbb{R}\}$ . In each case, what does it mean to say that  $f$  is convex with respect to  $K$  (see Definition 19.22)?

**Exercise 19.5** Derive Proposition 19.23(iii) from Proposition 19.16(iv) and Proposition 19.23(iv).

**Exercise 19.6** Recover Corollary 19.21 from Corollary 19.28.

**Exercise 19.7** In the setting of Proposition 19.23, prove that  $-v$  is a dual solution if and only if

$$\inf_{x \in \text{dom } f} \mathcal{L}(x, v) = \sup_{w \in K^\ominus} \inf_{x \in \text{dom } f} \mathcal{L}(x, w). \quad (19.79)$$

**Exercise 19.8** Consider Corollary 19.28 when  $\mathcal{H} = \mathbb{R}$ ,  $f: x \mapsto x$ ,  $p = 1$ ,  $m = 2$ ,  $g_1: x \mapsto x^2$ ,  $u_2 = 0$ , and  $\rho_2 = 0$ . Determine  $F$ , the primal problem, the dual problem, the Lagrangian, all saddle points, and the value function.

**Exercise 19.9** Check the details in Example 19.26(vi).

**Exercise 19.10** In the setting of Example 19.26(vi), determine when the value function  $\vartheta$  is lower semicontinuous at 0 and when  $\partial\vartheta(0) \neq \emptyset$ . Is it possible that  $\vartheta(0) = -\infty$ ?

# Chapter 20

## Monotone Operators

The theory of monotone set-valued operators plays a central role in many areas of nonlinear analysis. A prominent example of a monotone operator is the subdifferential operator investigated in Chapter 16. Single-valued monotone operators will be seen to be closely related to the nonexpansive operators studied in Chapter 4. Our investigation of monotone operators will rely heavily on the Fitzpatrick function.

The conventions introduced in Section 1.2 will be used. In particular, an operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that, for every  $x \in \mathcal{H}$ ,  $Ax$  is either empty or a singleton will be identified with the corresponding (at most) single-valued operator. Conversely, if  $D$  is a nonempty subset of  $\mathcal{H}$  and  $T: D \rightarrow \mathcal{H}$ , then  $T$  will be identified with the set-valued operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ , where, for every  $x \in \mathcal{H}$ ,  $Ax = \{Tx\}$  if  $x \in D$ , and  $Ax = \emptyset$  otherwise.

### 20.1 Monotone Operators

**Definition 20.1** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ . Then  $A$  is *monotone* if

$$(\forall(x, u) \in \text{gra } A)(\forall(y, v) \in \text{gra } A) \quad \langle x - y \mid u - v \rangle \geq 0. \quad (20.1)$$

A subset of  $\mathcal{H} \times \mathcal{H}$  is *monotone* if it is the graph of a monotone operator.

**Proposition 20.2** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ . Then the following are equivalent:

- (i)  $A$  is monotone.
- (ii)  $A$  is accretive, i.e.,

$$(\forall(x, u) \in \text{gra } A)(\forall(y, v) \in \text{gra } A)(\forall\alpha \in [0, 1]) \\ \|x - y + \alpha(u - v)\| \geq \|x - y\|. \quad (20.2)$$

- (iii) The following holds.

$$\begin{aligned}
& (\forall(x, u) \in \text{gra } A) (\forall(y, v) \in \text{gra } A) \\
& \|y - u\|^2 + \|x - v\|^2 \geq \|x - u\|^2 + \|y - v\|^2. \quad (20.3)
\end{aligned}$$

*Proof.* (20.2) $\Leftrightarrow$ (20.1): This follows from (20.1) and Lemma 2.12(i).

(20.3) $\Leftrightarrow$ (20.1): Use Lemma 2.11(i).  $\square$

**Example 20.3** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $\partial f$  is monotone.

*Proof.* Take  $(x, u)$  and  $(y, v)$  in  $\text{gra } \partial f$ . Then, by (16.1),  $\langle x - y \mid u \rangle + f(y) \geq f(x)$  and  $\langle y - x \mid v \rangle + f(x) \geq f(y)$ . Adding these inequalities, we conclude that  $\langle x - y \mid u - v \rangle \geq 0$ .  $\square$

**Example 20.4** Suppose that  $\mathcal{H} = \mathbb{R}$ , let  $D$  be a nonempty subset of  $\mathcal{H}$ , and let  $A: D \rightarrow \mathcal{H}$  be increasing. Then  $A$  is monotone.

**Example 20.5** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be cocoercive (in particular, firmly nonexpansive). Then  $T$  is monotone.

*Proof.* See (4.5).  $\square$

**Example 20.6** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be  $\alpha$ -averaged, with  $\alpha \in ]0, 1/2]$ . Then  $T$  is monotone.

*Proof.* Combine Remark 4.27 with Example 20.5, or use the equivalence (i) $\Leftrightarrow$ (iv) in Proposition 4.25.  $\square$

**Example 20.7** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$  be nonexpansive, let  $\alpha \in [-1, 1]$ , and set  $A = \text{Id} + \alpha T$ . Then, for every  $x \in D$  and every  $y \in D$ ,

$$\begin{aligned}
\langle x - y \mid Ax - Ay \rangle &= \|x - y\|^2 + \alpha \langle x - y \mid Tx - Ty \rangle \\
&\geq \|x - y\| (\|x - y\| - |\alpha| \|Tx - Ty\|) \\
&\geq 0.
\end{aligned} \quad (20.4)$$

Consequently,  $A$  is monotone.

**Example 20.8** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , and set  $A = \text{Id} - T$ . Then the following are equivalent:

(i)  $T$  is *pseudononexpansive* (or *pseudocontractive*), i.e.,

$$(\forall x \in D)(\forall y \in D) \quad \|Tx - Ty\|^2 \leq \|x - y\|^2 + \|(\text{Id} - T)x - (\text{Id} - T)y\|^2. \quad (20.5)$$

(ii)  $A$  is monotone.

*Proof.* Take  $x$  and  $y$  in  $D$ . Then  $\|x - y\|^2 + \|(\text{Id} - T)x - (\text{Id} - T)y\|^2 \geq \|Tx - Ty\|^2 \Leftrightarrow \|x - y\|^2 + \|x - y\|^2 - 2\langle x - y \mid Tx - Ty \rangle + \|Tx - Ty\|^2 \geq \|Tx - Ty\|^2 \Leftrightarrow \|x - y\|^2 - \langle x - y \mid Tx - Ty \rangle \geq 0 \Leftrightarrow \langle x - y \mid Ax - Ay \rangle \geq 0$ .  $\square$



**Example 20.9** Let  $\mathbf{H}$  be a real Hilbert space, let  $T \in \mathbb{R}_{++}$ , and suppose that  $\mathcal{H} = L^2([0, T]; \mathbf{H})$  (see Example 2.7). Furthermore, let  $A$  be the time-derivative operator (see Example 2.9)

$$A: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \begin{cases} \{x'\}, & \text{if } x \in D; \\ \emptyset, & \text{otherwise,} \end{cases} \quad (20.6)$$

where (initial condition)

$$D = \{x \in W^{1,2}([0, T]; \mathbf{H}) \mid x(0) = x_0\} \quad \text{for some } x_0 \in \mathbf{H} \quad (20.7)$$

or (periodicity condition)

$$D = \{x \in W^{1,2}([0, T]; \mathbf{H}) \mid x(0) = x(T)\}. \quad (20.8)$$

Then  $A$  is monotone.

*Proof.* Take  $x$  and  $y$  in  $D = \text{dom } A$ . Then

$$\begin{aligned} \langle x - y \mid Ax - Ay \rangle &= \int_0^T \langle x(t) - y(t) \mid x'(t) - y'(t) \rangle_{\mathbf{H}} dt \\ &= \frac{1}{2} \int_0^T \frac{d\|x(t) - y(t)\|_{\mathbf{H}}^2}{dt} dt \\ &= \frac{1}{2} (\|x(T) - y(T)\|_{\mathbf{H}}^2 - \|x(0) - y(0)\|_{\mathbf{H}}^2) \\ &\geq 0, \end{aligned} \quad (20.9)$$

which shows that  $A$  is monotone.  $\square$

Further examples can be constructed via the following monotonicity-preserving operations.

**Proposition 20.10** *Let  $\mathcal{K}$  be a real Hilbert space, let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and  $B: \mathcal{K} \rightarrow 2^{\mathcal{K}}$  be monotone operators, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $\gamma \in \mathbb{R}_+$ . Then the operators  $A^{-1}$ ,  $\gamma A$ , and  $A + L^*BL$  are monotone.*

**Proposition 20.11** *Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space, let  $(\mathbf{H}, \langle \cdot \mid \cdot \rangle_{\mathbf{H}})$  be a real Hilbert space, and let  $\mathbf{A}: \mathbf{H} \rightarrow 2^{\mathbf{H}}$  be a monotone operator. Suppose that  $\mathcal{H} = L^2((\Omega, \mathcal{F}, \mu); \mathbf{H})$  and define  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  via*

$$\text{gra } A = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid (x(\omega), u(\omega)) \in \text{gra } \mathbf{A} \quad \mu\text{-a.e. on } \Omega\}. \quad (20.10)$$

*Then  $A$  is monotone.*

*Proof.* Take  $(x, u)$  and  $(y, v)$  in  $\text{gra } A$ . By monotonicity of  $\mathbf{A}$ ,

$$\langle x(\omega) - y(\omega) \mid u(\omega) - v(\omega) \rangle_{\mathbf{H}} \geq 0 \quad \mu\text{-a.e. on } \Omega. \quad (20.11)$$

In view of Example 2.5, integrating these inequalities over  $\Omega$  with respect to  $\mu$ , we obtain

$$\langle x - y \mid u - v \rangle = \int_{\Omega} \langle x(\omega) - y(\omega) \mid u(\omega) - v(\omega) \rangle_{\mathcal{H}} \mu(d\omega) \geq 0, \quad (20.12)$$

which shows that  $A$  is monotone.  $\square$

Monotone operators also arise naturally in the study of best approximation and farthest-point problems.

**Example 20.12** Let  $C$  be a nonempty subset of  $\mathcal{H}$  and let  $\Pi_C$  be the set-valued projector onto  $C$  defined in (3.12). Then  $\Pi_C$  is monotone.

*Proof.* Take  $(x, u)$  and  $(y, v)$  in  $\text{gra } \Pi_C$ . Then  $\|x - u\| = d_C(x) \leq \|x - v\|$  and  $\|y - v\| = d_C(y) \leq \|y - u\|$ . Hence  $\|x - u\|^2 + \|y - v\|^2 \leq \|x - v\|^2 + \|y - u\|^2$ . Now expand the squares and simplify to obtain the monotonicity of  $\Pi_C$ .  $\square$

**Example 20.13** Let  $C$  be a nonempty bounded subset of  $\mathcal{H}$  and denote by

$$\Phi_C: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \{u \in C \mid \|x - u\| = \sup \|x - C\|\} \quad (20.13)$$

its *farthest-point operator*. Then  $-\Phi_C$  is monotone.

*Proof.* Suppose that  $(x, u)$  and  $(y, v)$  are in  $\text{gra } \Phi_C$ . Then  $\|x - u\| \geq \|x - v\|$  and  $\|y - v\| \geq \|y - u\|$ . Hence  $\|x - u\|^2 + \|y - v\|^2 \geq \|x - v\|^2 + \|y - u\|^2$ . Now expand the squares and simplify to see that  $-\Phi_C$  is monotone.  $\square$

Next, we provide two characterizations of monotonicity.

**Proposition 20.14** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and set  $F = \langle \cdot \mid \cdot \rangle$ . Then the following are equivalent:

- (i)  $A$  is monotone.
- (ii) For all finite families  $(\alpha_i)_{i \in I}$  in  $]0, 1[$  such that  $\sum_{i \in I} \alpha_i = 1$  and  $(x_i, u_i)_{i \in I}$  in  $\text{gra } A$ , we have

$$F\left(\sum_{i \in I} \alpha_i (x_i, u_i)\right) \leq \sum_{i \in I} \alpha_i F(x_i, u_i). \quad (20.14)$$

- (iii)  $F$  is convex on  $\text{gra } A$ .

*Proof.* This follows from Lemma 2.13(i).  $\square$

We devote the remainder of this section to linear monotone operators.

**Example 20.15** Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be linear. Then  $A$  is monotone if and only if  $(\forall x \in \mathcal{H}) \langle Ax \mid x \rangle \geq 0$ .

**Example 20.16** Let  $A \in \mathcal{B}(\mathcal{H})$ . Then the following hold:

- (i)  $A$  is monotone  $\Leftrightarrow A + A^*$  is monotone  $\Leftrightarrow A^*$  is monotone.
- (ii)  $A^*A$ ,  $AA^*$ ,  $A - A^*$ , and  $A^* - A$  are monotone.

**Proposition 20.17** *Let  $A \in \mathcal{B}(\mathcal{H})$  be monotone. Then  $\ker A = \ker A^*$  and  $\overline{\text{ran}} A = \overline{\text{ran}} A^*$ .*

*Proof.* Take  $x \in \ker A$  and  $v \in \text{ran } A$ , say  $v = Ay$ . Then  $(\forall \alpha \in \mathbb{R})$   $0 \leq \langle \alpha x + y | A(\alpha x + y) \rangle = \alpha \langle x | v \rangle + \langle y | Ay \rangle$ . Hence  $\langle x | v \rangle = 0$  and thus  $\ker A \subset (\text{ran } A)^\perp = \ker A^*$  by Fact 2.18(iv). Since  $A^* \in \mathcal{B}(\mathcal{H})$  is also monotone, we obtain  $\ker A^* \subset \ker A^{**} = \ker A$ . Altogether,  $\ker A = \ker A^*$  and therefore  $\overline{\text{ran}} A = \overline{\text{ran}} A^*$  by Fact 2.18(iii).  $\square$

**Fact 20.18** [214, Chapter VII] *Let  $A$  and  $B$  be self-adjoint monotone operators in  $\mathcal{B}(\mathcal{H})$  such that  $AB = BA$ . Then  $AB$  is monotone.*

**Example 20.19** Set

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad \text{and} \quad C = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}. \quad (20.15)$$

Then  $A$ ,  $B$ ,  $C$ , and  $-C$  are continuous, linear, and monotone operators from  $\mathbb{R}^2$  to  $\mathbb{R}^2$ . However, neither  $AB$  nor  $C^2$  is monotone. This shows that the assumption on self-adjointness and commutativity in Fact 20.18 are important.

## 20.2 Maximally Monotone Operators

**Definition 20.20** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. Then  $A$  is *maximally monotone* (or *maximal monotone*) if there exists no monotone operator  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\text{gra } B$  properly contains  $\text{gra } A$ , i.e., for every  $(x, u) \in \mathcal{H} \times \mathcal{H}$ ,

$$(x, u) \in \text{gra } A \quad \Leftrightarrow \quad (\forall (y, v) \in \text{gra } A) \quad \langle x - y | u - v \rangle \geq 0. \quad (20.16)$$

**Theorem 20.21** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. Then there exists a maximally monotone extension of  $A$ , i.e., a maximally monotone operator  $\tilde{A}: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\text{gra } A \subset \text{gra } \tilde{A}$ .*

*Proof.* Assume first that  $\text{gra } A \neq \emptyset$  and set

$$\mathcal{M} = \{B: \mathcal{H} \rightarrow 2^{\mathcal{H}} \mid B \text{ is monotone and } \text{gra } A \subset \text{gra } B\}. \quad (20.17)$$

Then  $\mathcal{M}$  is nonempty and partially ordered via  $(\forall B_1 \in \mathcal{M})(\forall B_2 \in \mathcal{M})$   $B_1 \preceq B_2 \Leftrightarrow \text{gra } B_1 \subset \text{gra } B_2$ . Let  $\mathcal{C}$  be a chain in  $\mathcal{M}$ . Then the operator the graph of which is  $\bigcup_{C \in \mathcal{C}} \text{gra } C$  is an upper bound of  $\mathcal{C}$ . Therefore, Zorn's lemma (Fact 1.1) guarantees the existence of a maximal element  $\tilde{A} \in \mathcal{M}$ . The operator  $\tilde{A}$  possesses the required properties. Now assume that  $\text{gra } A = \emptyset$ .

Then any maximally monotone extension  $\tilde{A}$  of the operator the graph of which is  $\{(0, 0)\}$  is as required.  $\square$

The proofs of the next two propositions are left as exercises.

**Proposition 20.22** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $z \in \mathcal{H}$ , let  $u \in \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Then  $A^{-1}$  and  $x \mapsto u + \gamma A(x + z)$  are maximally monotone.*

**Proposition 20.23** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and  $B: \mathcal{K} \rightarrow 2^{\mathcal{K}}$  be maximally monotone. Then  $A \times B: \mathcal{H} \times \mathcal{K} \rightarrow 2^{\mathcal{H} \times \mathcal{K}}: (x, y) \mapsto Ax \times By$  is maximally monotone.*

**Proposition 20.24** *Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be monotone and hemicontinuous, i.e., for every  $(x, y, z) \in \mathcal{H}^3$ ,  $\lim_{\alpha \downarrow 0} \langle z \mid A(x + \alpha y) \rangle = \langle z \mid Ax \rangle$ . Then  $A$  is maximally monotone.*

*Proof.* Let  $(x, u) \in \mathcal{H} \times \mathcal{H}$  be such that  $(\forall y \in \mathcal{H}) \langle x - y \mid u - Ay \rangle \geq 0$ . We must show that  $u = Ax$ . For every  $\alpha \in ]0, 1]$ , set  $y_\alpha = x + \alpha(u - Ax)$  and observe that  $\langle u - Ax \mid u - Ay_\alpha \rangle = -\langle x - y_\alpha \mid u - Ay_\alpha \rangle / \alpha \leq 0$ . Since  $A$  is hemicontinuous, we conclude that  $\|u - Ax\|^2 \leq 0$ , i.e., that  $u = Ax$ .  $\square$

**Corollary 20.25** *Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be monotone and continuous. Then  $A$  is maximally monotone.*

**Example 20.26** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be nonexpansive and let  $\alpha \in [-1, 1]$ . Then  $\text{Id} + \alpha T$  is maximally monotone.

*Proof.* Combine Example 20.7 with Corollary 20.25.  $\square$

**Example 20.27** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be  $\alpha$ -averaged, with  $\alpha \in ]0, 1/2]$  (in particular, firmly nonexpansive). Then  $T$  is maximally monotone.

*Proof.* This follows from Corollary 20.25 since  $T$  is continuous and, by Example 20.6, monotone.  $\square$

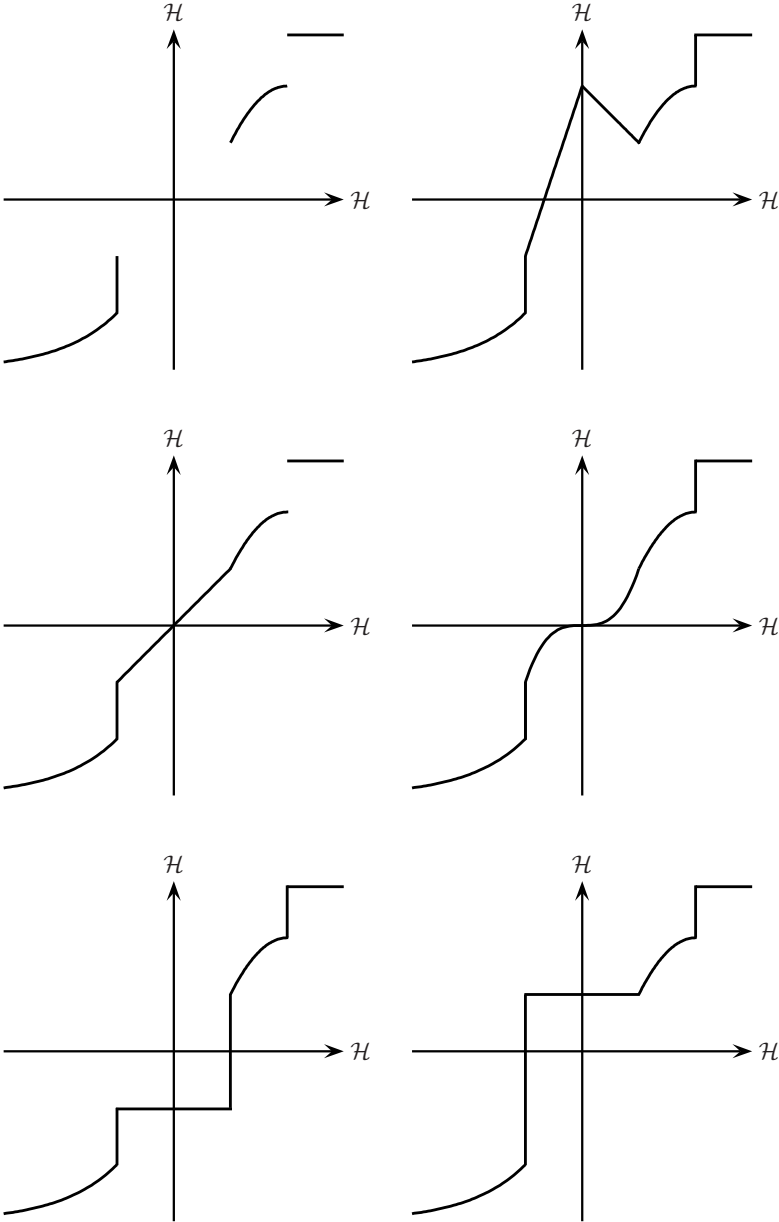
**Example 20.28** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive, with  $\beta \in \mathbb{R}_{++}$ . Then  $T$  is maximally monotone.

*Proof.* Since  $\beta T$  is firmly nonexpansive, it is maximally monotone by Example 20.27.  $\square$

**Example 20.29** Let  $A \in \mathcal{B}(\mathcal{H})$  be monotone. Then  $A$  is maximally monotone.

*Proof.* This follows from Example 20.15 and Proposition 20.24.  $\square$

**Example 20.30** Let  $A \in \mathcal{B}(\mathcal{H})$  be such that  $A^* = -A$ . Then  $A$  is maximally monotone.



**Fig. 20.1** Extensions of a monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  ( $\mathcal{H} = \mathbb{R}$ ). Top left: Graph of  $A$ . Top right: Graph of a nonmonotone extension of  $A$ . Center left: Graph of a monotone extension of  $A$ . Center right, bottom left, and bottom right: Graphs of maximally monotone extensions of  $A$ .

*Proof.* We have  $(\forall x \in \mathcal{H}) \langle x \mid Ax \rangle = 0$ . Hence  $A$  is monotone, and maximally so by Example 20.29.  $\square$

We now present some basic properties of maximally monotone operators.

**Proposition 20.31** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \mathcal{H}$ . Then  $Ax$  is closed and convex.*

*Proof.* We assume that  $x \in \text{dom } A$ . Then (20.16) yields

$$Ax = \bigcap_{(y,v) \in \text{gra } A} \{u \in \mathcal{H} \mid \langle x - y \mid u - v \rangle \geq 0\}, \quad (20.18)$$

which is an intersection of closed convex sets.  $\square$

The next two propositions address various closedness properties of the graph of a maximally monotone operator.

**Proposition 20.32** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $(x_b, u_b)_{b \in B}$  be a bounded net in  $\text{gra } A$ , and let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then the following hold:*

- (i) *Suppose that  $x_b \rightarrow x$  and  $u_b \rightarrow u$ . Then  $(x, u) \in \text{gra } A$ .*
- (ii) *Suppose that  $x_b \rightharpoonup x$  and  $u_b \rightarrow u$ . Then  $(x, u) \in \text{gra } A$ .*

*Proof.* (i): Take  $(y, v) \in \text{gra } A$ . Then  $(\forall b \in B) \langle x_b - y \mid u_b - v \rangle \geq 0$  by (20.16). In turn, Lemma 2.36 implies that  $\langle x - y \mid u - v \rangle \geq 0$ . Hence  $(x, u) \in \text{gra } A$  by (20.16).

(ii): Apply (i) to  $A^{-1}$ .  $\square$

**Proposition 20.33** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then the following hold:*

- (i)  *$\text{gra } A$  is sequentially closed in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ , i.e., for every sequence  $(x_n, u_n)_{n \in \mathbb{N}}$  in  $\text{gra } A$  and every  $(x, u) \in \mathcal{H} \times \mathcal{H}$ , if  $x_n \rightarrow x$  and  $u_n \rightharpoonup u$ , then  $(x, u) \in \text{gra } A$ .*
- (ii)  *$\text{gra } A$  is sequentially closed in  $\mathcal{H}^{\text{weak}} \times \mathcal{H}^{\text{strong}}$ , i.e., for every sequence  $(x_n, u_n)_{n \in \mathbb{N}}$  in  $\text{gra } A$  and every  $(x, u) \in \mathcal{H} \times \mathcal{H}$ , if  $x_n \rightharpoonup x$  and  $u_n \rightarrow u$ , then  $(x, u) \in \text{gra } A$ .*
- (iii)  *$\text{gra } A$  is closed in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{strong}}$ .*

*Proof.* (i): Combine Lemma 2.38 and Proposition 20.32(i).

(ii): Apply (i) to  $A^{-1}$ .

(iii): A consequence of (i) (see Section 1.12).  $\square$

As we shall see in Example 21.5 and Remark 21.6, Proposition 20.32(i) is sharp in the sense that the boundedness assumption on the net cannot be removed. Regarding Proposition 20.33(i), the next example shows that it is not possible to replace  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$  by  $\mathcal{H}^{\text{weak}} \times \mathcal{H}^{\text{weak}}$ .

**Example 20.34** The graph of a maximally monotone operator need not be sequentially closed in  $\mathcal{H}^{\text{weak}} \times \mathcal{H}^{\text{weak}}$ . Indeed, suppose that  $\mathcal{H} = \ell^2(\mathbb{N})$  and set  $C = B(0; 1)$ . Then  $\text{Id} - P_C$  is firmly nonexpansive by Corollary 4.10, and hence maximally monotone by Example 20.27. Consider the sequence  $(x_n)_{n \in \mathbb{N}} = (e_1 + e_{2n})_{n \in \mathbb{N}}$ , where  $(e_n)_{n \in \mathbb{N}}$  is the sequence of standard unit vectors in  $\ell^2(\mathbb{N})$ . Then the sequence  $(x_n, (1 - 1/\sqrt{2})x_n)_{n \in \mathbb{N}}$  lies in  $\text{gra}(\text{Id} - P_C)$  and it converges weakly to  $(e_1, (1 - 1/\sqrt{2})e_1)$ . However, the weak limit  $(e_1, (1 - 1/\sqrt{2})e_1)$  does not belong to  $\text{gra}(\text{Id} - P_C)$ .

**Proposition 20.35** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and at most single-valued. Suppose that  $\text{dom } A$  is a linear subspace, that  $A|_{\text{dom } A}$  is linear, and that  $(\forall x \in \text{dom } A)(\forall y \in \text{dom } A) \langle x \mid Ay \rangle = \langle Ax \mid y \rangle$ . Set*

$$h: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \begin{cases} \frac{1}{2} \langle x \mid Ax \rangle, & \text{if } x \in \text{dom } A; \\ +\infty, & \text{otherwise,} \end{cases} \quad (20.19)$$

and

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sup_{y \in \text{dom } A} (\langle x \mid Ay \rangle - h(y)). \quad (20.20)$$

Then the following hold:

- (i)  $f + \iota_{\text{dom } A} = h$ .
- (ii)  $f \in \Gamma_0(\mathcal{H})$ .
- (iii)  $\partial f = A$ .
- (iv)  $f = h^{**}$ .

*Proof.* Take  $x \in \text{dom } A = \text{dom } h$ .

(i): For every  $y \in \text{dom } A$ ,  $0 \leq \langle x - y \mid Ax - Ay \rangle = \langle x \mid Ax \rangle + \langle y \mid Ay \rangle - 2 \langle x \mid Ay \rangle$ , which implies that  $\langle x \mid Ay \rangle - h(y) \leq h(x)$ . Hence  $f(x) \leq h(x)$ . On the other hand,  $f(x) \geq \langle x \mid Ax \rangle - h(x) = h(x)$ . Altogether,  $f + \iota_{\text{dom } A} = h$ .

(ii): As a supremum of continuous affine functions,  $f \in \Gamma(\mathcal{H})$  by Proposition 9.3. In addition, since  $f(0) = 0$ ,  $f$  is proper.

(iii): For every  $y \in \mathcal{H}$ , we have  $f(x) + \langle y - x \mid Ax \rangle = \langle y \mid Ax \rangle - h(x) \leq f(y)$ . Consequently,  $Ax \in \partial f(x)$ . It follows that  $\text{gra } A \subset \text{gra } \partial f$ , which implies that  $A = \partial f$  since  $A$  is maximally monotone, while  $\partial f$  is monotone by Example 20.3.

(iv): Using (ii), Corollary 16.31, (iii), and (i), we see that  $f = (f + \iota_{\text{dom } \partial f})^{**} = (f + \iota_{\text{dom } A})^{**} = h^{**}$ .  $\square$

**Example 20.36** Suppose that  $\mathcal{H}$  is infinite-dimensional and separable with an orthonormal basis  $(e_n)_{n \in \mathbb{N}}$ , let  $(\beta_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1]$  such that  $\beta_n \downarrow 0$ , and set  $B: \mathcal{H} \rightarrow \mathcal{H}: x \mapsto \sum_{n \in \mathbb{N}} \beta_n \langle x \mid e_n \rangle e_n$ . Then  $B \in \mathcal{B}(\mathcal{H})$ ,  $B$  is maximally monotone and self-adjoint, and  $\text{ran } B$  is a proper linear subspace of  $\mathcal{H}$  that is dense in  $\mathcal{H}$ . Now set  $A = B^{-1}$  and

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sup_{y \in \text{dom } A} \left( \langle x \mid Ay \rangle - \frac{1}{2} \langle y \mid Ay \rangle \right). \quad (20.21)$$

Then  $A$  is maximally monotone,  $\text{dom } A$  is a dense and proper linear subspace of  $\mathcal{H}$ ,  $f$  is nowhere Gâteaux differentiable, and  $\partial f = A$  is at most single-valued.

*Proof.* Clearly,  $B$  is linear and bounded,  $\|B\| \leq 1$ ,  $B$  is self-adjoint, and  $B$  is strictly positive and thus injective. By Example 20.29,  $B$  is maximally monotone and so is  $A$  by Proposition 20.22. Note that  $A$  is linear and single-valued on its domain, and that  $\{\beta_n e_n\}_{n \in \mathbb{N}} \subset \text{ran } B$ . Thus,  $\text{dom } A = \text{ran } B$  is a dense linear subspace of  $\mathcal{H}$ . Since  $\sup_{n \in \mathbb{N}} \|A e_n\| = \sup_{n \in \mathbb{N}} \beta_n^{-1} = +\infty$ , it follows that  $A$  is not continuous and hence that  $\text{dom } A \neq \mathcal{H}$ . Proposition 20.35(ii)&(iii) implies that  $f \in \Gamma_0(\mathcal{H})$  and that  $\partial f = A$  is at most single-valued. Since  $\text{int dom } \partial f = \text{int dom } A = \emptyset$ ,  $f$  is nowhere Gâteaux differentiable by Proposition 17.41.  $\square$

### 20.3 Bivariate Functions and Maximal Monotonicity

We start with a technical fact.

**Lemma 20.37** *Let  $(z, w) \in \mathcal{H} \times \mathcal{H}$ , and set*

$$\begin{aligned} G: \mathcal{H} \times \mathcal{H} &\rightarrow \mathbb{R} \\ (x, u) &\mapsto -\langle x \mid u \rangle + \langle z - x \mid w - u \rangle + \frac{1}{2}\|z - x\|^2 + \frac{1}{2}\|w - u\|^2 \\ &= -\langle x \mid u \rangle + \frac{1}{2}\|(x - z) + (u - w)\|^2 \\ &= \langle z \mid w \rangle - \langle (x, u) \mid (w, z) \rangle + \frac{1}{2}\|(x, u) - (z, w)\|^2 \end{aligned} \quad (20.22)$$

and  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: (u, x) \mapsto (-x, -u)$ . Then  $G^* = G \circ L$ . Furthermore, let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then the following hold:

- (i)  $G(x, u) + \langle x \mid u \rangle \geq 0$ .
- (ii)  $G(x, u) + \langle x \mid u \rangle = 0 \Leftrightarrow x - z = w - u$ .
- (iii)  $[G(x, u) + \langle x \mid u \rangle = 0 \text{ and } \langle z - x \mid w - u \rangle \geq 0] \Leftrightarrow (x, u) = (z, w)$ .

*Proof.* The formula  $G^*(u, x) = G(-x, -u)$  is a consequence of Proposition 13.16 (applied in  $\mathcal{H} \times \mathcal{H}$ ) and Proposition 13.20(iii). The remaining statements follow from (20.22).  $\square$

**Theorem 20.38** *Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a convex function such that  $F^*$  is proper and  $F^* \geq \langle \cdot \mid \cdot \rangle$ . Define  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  by*

$$\text{gra } A = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid F^*(u, x) = \langle x \mid u \rangle\}. \quad (20.23)$$

*Then the following hold:*

- (i)  $A$  is monotone.



(ii) Suppose that  $F \geq \langle \cdot \mid \cdot \rangle$ . Then  $A$  is maximally monotone.

*Proof.* (i): Suppose that  $(x, u)$  and  $(y, v)$  belong to  $\text{gra } A$ . Then by convexity of  $F^*$ ,

$$\begin{aligned} \langle x \mid u \rangle + \langle y \mid v \rangle &= F^*(u, x) + F^*(v, y) \\ &\geq 2F^*\left(\frac{1}{2}(u+v), \frac{1}{2}(x+y)\right) \\ &\geq \frac{1}{2} \langle x+y \mid u+v \rangle. \end{aligned} \quad (20.24)$$

Hence  $\langle x-y \mid u-v \rangle \geq 0$  and  $A$  is therefore monotone.

(ii): Since  $F^*$  is proper, Proposition 13.9(iii) and Proposition 13.10(ii) imply that  $F$  is convex and that it possesses a continuous affine minorant. Using Proposition 13.40(iii) and the continuity of  $\langle \cdot \mid \cdot \rangle$ , we obtain  $F \geq \langle \cdot \mid \cdot \rangle \Rightarrow (\forall (x, u) \in \mathcal{H}) \lim_{(y, v) \rightarrow (x, u)} F(y, v) \geq \lim_{(y, v) \rightarrow (x, u)} \langle y \mid v \rangle$ . Hence  $F^{**} \geq \langle \cdot \mid \cdot \rangle$ . Now suppose that  $(z, w) \in \mathcal{H} \times \mathcal{H}$  satisfies

$$(\forall (x, u) \in \text{gra } A) \quad \langle z-x \mid w-u \rangle \geq 0. \quad (20.25)$$

We must show that  $(z, w) \in \text{gra } A$ . Define  $G$  and  $L$  as in Lemma 20.37, where it was observed that  $G^* = G \circ L$  and that  $G + \langle \cdot \mid \cdot \rangle \geq 0$ . Since  $F^{**} - \langle \cdot \mid \cdot \rangle \geq 0$ , we see that  $F^{**} + G \geq 0$ . By Proposition 13.14(iii) and Corollary 15.15 (applied to  $F^{**}$ ,  $G$ , and  $L$ ), there exists  $(v, y) \in \mathcal{H} \times \mathcal{H}$  such that  $0 \geq F^*(v, y) + (G \circ L)(-v, -y) = F^*(v, y) + G(y, v)$ . The assumption that  $F^* \geq \langle \cdot \mid \cdot \rangle$  and Lemma 20.37(i) then result in

$$0 \geq F^*(v, y) + G(y, v) \geq \langle y \mid v \rangle + G(y, v) \geq 0. \quad (20.26)$$

Hence

$$\langle y \mid v \rangle + G(y, v) = 0 \quad (20.27)$$

and  $F^*(v, y) = \langle y \mid v \rangle$ , i.e.,

$$(y, v) \in \text{gra } A. \quad (20.28)$$

In view of (20.25) and (20.28), we obtain

$$\langle z-y \mid w-v \rangle \geq 0. \quad (20.29)$$

Lemma 20.37(iii) shows that (20.27) and (20.29) are equivalent to  $(z, w) = (y, v)$ . Therefore, using (20.28), we deduce that  $(z, w) \in \text{gra } A$ .  $\square$

**Corollary 20.39** *Let  $F \in \Gamma_0(\mathcal{H} \times \mathcal{H})$  be autoconjugate and define  $A$  via*

$$\text{gra } A = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid F(x, u) = \langle x \mid u \rangle\}. \quad (20.30)$$

*Then  $A$  is maximally monotone.*

*Proof.* This follows from Proposition 13.31, Proposition 16.52, and Theorem 20.38(ii).  $\square$

A fundamental consequence of Corollary 20.39 is the following result on the maximality of the subdifferential.

**Theorem 20.40** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\partial f$  is maximally monotone.*

*Proof.* On the one hand,  $f \oplus f^*$  is autoconjugate. On the other hand,  $\{(x, u) \in \mathcal{H} \times \mathcal{H} \mid (f \oplus f^*)(x, u) = \langle x \mid u \rangle\} = \text{gra } \partial f$  by Proposition 16.9. Altogether,  $\partial f$  is maximally monotone by Corollary 20.39.  $\square$

**Example 20.41** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $N_C$  is maximally monotone.

*Proof.* Apply Theorem 20.40 to  $f = \iota_C$  and use Example 16.12.  $\square$

## 20.4 The Fitzpatrick Function

**Definition 20.42** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. The *Fitzpatrick function* of  $A$  is

$$F_A: \mathcal{H} \times \mathcal{H} \rightarrow [-\infty, +\infty]$$

$$(x, u) \mapsto \sup_{(y, v) \in \text{gra } A} (\langle y \mid u \rangle + \langle x \mid v \rangle - \langle y \mid v \rangle) \quad (20.31)$$

$$= \langle x \mid u \rangle - \inf_{(y, v) \in \text{gra } A} \langle x - y \mid u - v \rangle. \quad (20.32)$$

**Example 20.43**  $F_{\text{Id}}: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (x, u) \mapsto (1/4)\|x + u\|^2$ .

**Example 20.44** Let  $A \in \mathcal{B}(\mathcal{H})$  be such that  $A^* = -A$ . Then  $F_A = \iota_{\text{gra } A}$ .

**Example 20.45** Let  $A \in \mathcal{B}(\mathcal{H})$  be monotone and set  $q_A: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2)\langle x \mid Ax \rangle$ . Then  $(\forall (x, u) \in \mathcal{H} \times \mathcal{H})$   $F_A(x, u) = 2q_A^*\left(\frac{1}{2}u + \frac{1}{2}A^*x\right)$ .

*Proof.* Take  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then

$$\begin{aligned} F_A(x, u) &= \sup_{y \in \mathcal{H}} (\langle y \mid u \rangle + \langle x \mid Ay \rangle - \langle y \mid Ay \rangle) \\ &= 2 \sup_{y \in \mathcal{H}} \left( \left\langle y \mid \frac{1}{2}u + \frac{1}{2}A^*x \right\rangle - \frac{1}{2} \langle y \mid Ay \rangle \right) \\ &= 2q_A^*\left(\frac{1}{2}u + \frac{1}{2}A^*x\right). \end{aligned} \quad (20.33)$$

$\square$

**Example 20.46** Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $F_{\partial f} \leq f \oplus f^*$  and  $\text{dom } f \times \text{dom } f^* \subset \text{dom } F_{\partial f}$ .

*Proof.* Take  $(x, u) \in \text{dom } f \times \text{dom } f^*$  and  $(y, v) \in \text{gra } \partial f$ . Then  $\langle y | u \rangle + \langle x | v \rangle - \langle y | v \rangle = (\langle y | u \rangle - f(y)) + (\langle x | v \rangle - f^*(v)) \leq f^*(u) + f^{**}(x)$  by Proposition 16.9 and Proposition 13.13. Hence,  $F_{\partial f} \leq f \oplus f^*$ .  $\square$

**Proposition 20.47** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator such that  $\text{gra } A \neq \emptyset$  and let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Then the following hold:*

- (i) *Suppose that  $(x, u) \in \text{gra } A$ . Then  $F_A(x, u) = \langle x | u \rangle$ .*
- (ii)  $F_A = (\iota_{\text{gra } A^{-1}} + \langle \cdot | \cdot \rangle)^* \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ .
- (iii)  $F_A(x, u) \leq \langle x | u \rangle$  if and only if  $\{(x, u)\} \cup \text{gra } A$  is monotone.
- (iv)  $F_A(x, u) \leq F_A^*(u, x)$ .
- (v) *Suppose that  $(x, u) \in \text{gra } A$ . Then  $F_A^*(u, x) = \langle x | u \rangle$ .*
- (vi)  $F_A(x, u) = F_{A^{-1}}(u, x)$ .
- (vii) *Let  $\gamma \in \mathbb{R}_{++}$ . Then  $F_{\gamma A}(x, u) = \gamma F_A(x, u/\gamma)$ .*
- (viii) *Suppose that  $(x, u) \in \text{gra } A$ . Then  $(x, u) = \text{Prox}_{F_A}(x + u, x + u)$ .*

*Proof.* (i): We have  $\inf_{(y, v) \in \text{gra } A} \langle x - y | u - v \rangle = 0$ , and so (20.32) implies that  $F_A(x, u) = \langle x | u \rangle$ .

(ii): The identity  $F_A = (\iota_{\text{gra } A^{-1}} + \langle \cdot | \cdot \rangle)^*$  is clear from (20.31). Hence, (i) and Proposition 13.11 yield  $F_A \in \Gamma_0(\mathcal{H} \times \mathcal{H})$ .

(iii): Clear from (20.32).

(iv): We derive from (i) that

$$\begin{aligned}
 F_A(x, u) &= \sup_{(y, v) \in \text{gra } A} (\langle y | u \rangle + \langle x | v \rangle - \langle y | v \rangle) \\
 &= \sup_{(y, v) \in \text{gra } A} (\langle y | u \rangle + \langle x | v \rangle - F_A(y, v)) \\
 &\leq \sup_{(y, v) \in \mathcal{H} \times \mathcal{H}} (\langle y | u \rangle + \langle x | v \rangle - F_A(y, v)) \\
 &= \sup_{(y, v) \in \mathcal{H} \times \mathcal{H}} (\langle (y, v) | (u, x) \rangle - F_A(y, v)) \\
 &= F_A^*(u, x).
 \end{aligned} \tag{20.34}$$

(v): By (ii) and Proposition 13.14(i),  $F_A^* = (\iota_{\text{gra } A^{-1}} + \langle \cdot | \cdot \rangle)^{**} \leq \iota_{\text{gra } A^{-1}} + \langle \cdot | \cdot \rangle$ . This, (i), and (iv) imply that  $F_A^*(u, x) \leq \langle x | u \rangle = F_A(x, u) \leq F_A^*(u, x)$ , as required.

(vi)&(vii): Direct consequences of (20.31).

(viii): By (i) and (v),  $F_A(x, u) + F_A^*(u, x) = 2\langle x | u \rangle = \langle (x, u) | (u, x) \rangle$ . Hence  $(u, x) \in \partial F_A(x, u)$  and thus  $(x + u, x + u) \in (\text{Id} + \partial F_A)(x, u)$ , which yields the result.  $\square$

**Proposition 20.48** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then  $F_A \geq \langle \cdot | \cdot \rangle$  and*

$$\text{gra } A = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid F_A(x, u) = \langle x | u \rangle\}. \tag{20.35}$$

*Proof.* Take  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . If  $(x, u) \in \text{gra } A$ , then  $F_A(x, u) = \langle x | u \rangle$  by Proposition 20.47(i). On the other hand, if  $(x, u) \notin \text{gra } A$ , then  $\{(x, u)\} \cup \text{gra } A$  is not monotone, and Proposition 20.47(iii) yields  $F_A(x, u) > \langle x | u \rangle$ .  $\square$

**Corollary 20.49** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $x$  and  $u$  be in  $\mathcal{H}$ , and let  $(x_n, u_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } A$  such that  $(x_n, u_n) \rightharpoonup (x, u)$ . Then the following hold:*

- (i)  $\langle x | u \rangle \leq \underline{\lim} \langle x_n | u_n \rangle$ .
- (ii) Suppose that  $\underline{\lim} \langle x_n | u_n \rangle = \langle x | u \rangle$ . Then  $(x, u) \in \text{gra } A$ .
- (iii) Suppose that  $\underline{\lim} \langle x_n | u_n \rangle \leq \langle x | u \rangle$ . Then  $\langle x_n | u_n \rangle \rightarrow \langle x | u \rangle$  and  $(x, u) \in \text{gra } A$ .

*Proof.* (i): By Proposition 20.47(ii) and Theorem 9.1,  $F_A$  is weakly lower semicontinuous. Hence, Proposition 20.48 and the assumptions imply that

$$\langle x | u \rangle \leq F_A(x, u) \leq \underline{\lim} F_A(x_n, u_n) = \underline{\lim} \langle x_n | u_n \rangle. \quad (20.36)$$

(ii): In this case, (20.36) implies that  $\langle x | u \rangle = F_A(x, u)$ . By Proposition 20.48,  $(x, u) \in \text{gra } A$ .

(iii): Using (i), we see that  $\langle x | u \rangle \leq \underline{\lim} \langle x_n | u_n \rangle \leq \overline{\lim} \langle x_n | u_n \rangle \leq \langle x | u \rangle$ . Hence  $\langle x_n | u_n \rangle \rightarrow \langle x | u \rangle$  and, by (ii),  $(x, u) \in \text{gra } A$ .  $\square$

**Proposition 20.50** *Let  $C$  and  $D$  be closed affine subspaces of  $\mathcal{H}$  such that  $D - D = (C - C)^\perp$ , let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $(x_n, u_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } A$ , and let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Suppose that*

$$x_n \rightharpoonup x, \quad u_n \rightharpoonup u, \quad x_n - P_C x_n \rightarrow 0, \quad \text{and} \quad u_n - P_D u_n \rightarrow 0. \quad (20.37)$$

*Then  $(x, u) \in (C \times D) \cap \text{gra } A$  and  $\langle x_n | u_n \rangle \rightarrow \langle x | u \rangle$ .*

*Proof.* Set  $V = C - C$ . Since  $P_C x_n \rightharpoonup x$  and  $C$  is weakly sequentially closed by Theorem 3.32, we have  $x \in C$  and likewise  $u \in D$ . Hence,  $C = x + V$  and  $D = u + V^\perp$ . Thus, using Corollary 3.20(i),

$$P_C: w \mapsto P_V w + P_{V^\perp} x \quad \text{and} \quad P_D: w \mapsto P_{V^\perp} w + P_V u. \quad (20.38)$$

Therefore, since  $P_V$  and  $P_{V^\perp}$  are weakly continuous by Proposition 4.11(i), it follows from Lemma 2.41(iii) that

$$\begin{aligned} \langle x_n | u_n \rangle &= \langle P_V x_n + P_{V^\perp} x_n | P_V u_n + P_{V^\perp} u_n \rangle \\ &= \langle P_V x_n | P_V u_n \rangle + \langle P_{V^\perp} x_n | P_{V^\perp} u_n \rangle \\ &= \langle P_V x_n | u_n - P_{V^\perp} u_n \rangle + \langle x_n - P_V x_n | P_{V^\perp} u_n \rangle \\ &= \langle P_V x_n | u_n - (P_D u_n - P_V u) \rangle + \langle x_n - (P_C x_n - P_{V^\perp} x) | P_{V^\perp} u_n \rangle \\ &= \langle P_V x_n | u_n - P_D u_n \rangle + \langle P_V x_n | P_V u \rangle \\ &\quad + \langle x_n - P_C x_n | P_{V^\perp} u_n \rangle + \langle P_{V^\perp} x | P_{V^\perp} u_n \rangle \\ &\rightarrow \langle P_V x | P_V u \rangle + \langle P_{V^\perp} x | P_{V^\perp} u \rangle \\ &= \langle x | u \rangle. \end{aligned} \quad (20.39)$$

Thus, the result follows from Corollary 20.49(iii).  $\square$

**Proposition 20.51** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator such that  $\text{gra } A \neq \emptyset$ . Then the following hold:*

- (i)  $F_A^* = (\iota_{\text{gra } A^{-1}} + \langle \cdot | \cdot \rangle)^{**}$ .
- (ii)  $\text{conv gra } A^{-1} \subset \text{dom } F_A^* \subset \overline{\text{conv}} \text{ gra } A^{-1} \subset \overline{\text{conv}} \text{ ran } A \times \overline{\text{conv}} \text{ dom } A$ .
- (iii)  $F_A^* \geq \langle \cdot | \cdot \rangle$ .
- (iv) *Suppose that  $A$  is maximally monotone. Then*

$$\text{gra } A = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid F_A^*(u, x) = \langle x | u \rangle\}. \quad (20.40)$$

*Proof.* (i): Proposition 20.47(ii).

(ii): Combine (i), Proposition 9.8(iv), and Proposition 13.39.

(iii)&(iv): Let  $B$  be a maximally monotone extension of  $A$  and take  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Since  $F_A \leq F_B$ , we derive from Proposition 13.14(ii) that  $F_A^* \geq F_B^*$ . Hence, Proposition 20.47(iv) and Proposition 20.48 imply that

$$F_A^*(u, x) \geq F_B^*(u, x) \geq F_B(x, u) \geq F_A(x, u) \geq \langle x | u \rangle = \langle u | x \rangle, \quad (20.41)$$

which yields (iii). If  $(x, u) \notin \text{gra } B$ , then (20.41) and (20.35) yield  $F_B^*(u, x) \geq F_B(x, u) > \langle x | u \rangle$ . Now assume that  $(x, u) \in \text{gra } B$  and take  $(y, v) \in \mathcal{H} \times \mathcal{H}$ . Then  $F_B(y, v) \geq \langle (y, v) | (u, x) \rangle - \langle x | u \rangle$  and hence  $\langle x | u \rangle \geq \langle (y, v) | (u, x) \rangle - F_B(y, v)$ . Taking the supremum over  $(y, v) \in \mathcal{H} \times \mathcal{H}$ , we obtain  $\langle x | u \rangle \geq F_B^*(u, x)$ . In view of (20.41),  $\langle x | u \rangle = F_B^*(u, x)$ . Thus, if  $A$  is maximally monotone, then  $B = A$  and (20.40) is verified.  $\square$

**Remark 20.52** Proposition 20.48 and Proposition 20.51 provide converses to Theorem 20.38.

Using the proximal average, it is possible to construct maximally monotone extensions.

**Theorem 20.53** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator such that  $\text{gra } A \neq \emptyset$ . Set  $G = \text{pav}(F_A, F_A^{*\top})$  and define  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  via*

$$\text{gra } B = \{(x, u) \in \mathcal{H} \times \mathcal{H} \mid G(x, u) = \langle x | u \rangle\}. \quad (20.42)$$

*Then  $B$  is a maximally monotone extension of  $A$ .*

*Proof.* By Proposition 20.47(ii),  $F_A$  belongs to  $\Gamma_0(\mathcal{H} \times \mathcal{H})$  and hence so does  $F_A^{*\top}$ . Using Corollary 14.8(i)&(ii), Proposition 13.30, Proposition 14.7(i), and Proposition 14.10, we obtain that  $G \in \Gamma_0(\mathcal{H} \times \mathcal{H})$  and that  $G^* = (\text{pav}(F_A, F_A^{*\top}))^* = \text{pav}(F_A^*, F_A^{*\top*}) = \text{pav}(F_A^*, F_A^{\top**}) = \text{pav}(F_A^*, F_A^{\top}) = \text{pav}(F_A^{\top}, F_A^*) = \text{pav}(F_A^{\top}, F_A^{*\top\top}) = (\text{pav}(F_A, F_A^{*\top}))^{\top} = G^{\top}$ . Hence  $G$  is autoconjugate and Corollary 20.39 implies that  $B$  is maximally monotone. Set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: (x, u) \mapsto (u, x)$ , and let  $(x, u) \in \text{gra } A$ . Using Corollary 14.8(iv), Proposition 20.47(viii), and Proposition 16.53, we see that

$$\text{Prox}_G(x + u, x + u) = \frac{1}{2} \text{Prox}_{F_A}(x + u, x + u) + \frac{1}{2} \text{Prox}_{F_A^{*\top}}(x + u, x + u)$$

$$\begin{aligned}
&= \frac{1}{2}(x, u) + \frac{1}{2}(\text{Id} - L \text{Prox}_{F_A} L)(x + u, x + u) \\
&= \frac{1}{2}(x, u) + \frac{1}{2}((x + u, x + u) - (u, x)) \\
&= (x, u).
\end{aligned} \tag{20.43}$$

In view of Proposition 16.52, it follows that  $G(x, u) = \langle x \mid u \rangle$  and hence that  $(x, u) \in \text{gra } B$ . Therefore,  $\text{gra } A \subset \text{gra } B$  and the proof is complete.  $\square$

## Exercises

**Exercise 20.1** Prove Proposition 20.10.

**Exercise 20.2** Verify Example 20.15 and Example 20.16.

**Exercise 20.3** Provide a proof rooted in linear algebra for Fact 20.18 when  $\mathcal{H} = \mathbb{R}^N$ .

**Exercise 20.4** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and suppose that  $\text{gra } A$  is a convex set. Prove that  $\text{gra } A$  is an affine subspace.

**Exercise 20.5** Prove Proposition 20.22.

**Exercise 20.6** Deduce Proposition 20.33(i)&(ii) from Proposition 20.50.

**Exercise 20.7** Suppose that  $\mathcal{H}$  is infinite-dimensional. Use Exercise 18.11 to construct a maximally monotone operator  $A: \mathcal{H} \rightarrow \mathcal{H}$  that is strong-to-weak continuous and hence hemicontinuous on  $\mathcal{H}$ , strong-to-strong continuous on  $\mathcal{H} \setminus \{0\}$ , but not strong-to-strong continuous at 0.

**Exercise 20.8** Verify Example 20.43.

**Exercise 20.9** Verify Example 20.44.

**Exercise 20.10** Consider Example 20.46. Find  $f \in \Gamma_0(\mathcal{H})$  such that  $\text{dom } f \times \text{dom } f^* = \text{dom } F_{\partial f} = \overline{\text{dom } F_{\partial f}}$ .

**Exercise 20.11** Consider Example 20.46 when  $\mathcal{H} = \mathbb{R}$  and  $f$  is the negative Burg entropy function (defined in Example 9.30(viii)). Demonstrate that  $\text{dom } F_{\partial f}$  is closed and properly contains  $\text{dom } f \times \text{dom } f^*$ .

**Exercise 20.12** Suppose that  $\mathcal{H} = L^2([0, 1])$  (see Example 2.7) and define the *Volterra integration operator*  $A: \mathcal{H} \rightarrow \mathcal{H}$  by  $(\forall x \in \mathcal{H})(\forall t \in [0, 1])$   $(Ax)(t) = \int_0^t x(s) ds$ . Show that  $A$  is continuous, linear, and monotone, and that  $\text{ran } A \neq \text{ran } A^*$ . Conclude that the closures in Proposition 20.17 are important.

**Exercise 20.13** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Show that  $F_{N_C} = \iota_C \oplus \iota_C^*$ .

**Exercise 20.14** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Show that for every  $(x, u) \in \mathcal{H} \times \mathcal{H}$ ,

$$F_{P_C}(x, u) = \begin{cases} \left\langle P_C\left(\frac{1}{2}x + \frac{1}{2}u\right) \mid x + u \right\rangle - \left\| P_C\left(\frac{1}{2}x + \frac{1}{2}u\right) \right\|^2, & \text{if } u \in C; \\ +\infty, & \text{if } u \notin C. \end{cases} \quad (20.44)$$

**Exercise 20.15** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, and let  $\varepsilon \in \mathbb{R}_+$ . The  $\varepsilon$ -enlargement of  $A$  is

$$A^\varepsilon: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \bigcap_{(z, w) \in \text{gra } A} \{u \in \mathcal{H} \mid \langle x - z \mid u - w \rangle \geq -\varepsilon\}. \quad (20.45)$$

Now let  $(x, u)$  and  $(y, v)$  be in  $\text{gra } A^\varepsilon$ . Use Proposition 20.48 to show that  $\langle x - y \mid u - v \rangle \geq -4\varepsilon$ .





# Chapter 21

## Finer Properties of Monotone Operators

In this chapter, we deepen our study of (maximally) monotone operators. The main results are Minty's theorem, which conveniently characterizes maximal monotonicity, and the Debrunner–Flor theorem, which concerns the existence of a maximally monotone extension with a prescribed domain localization. Another highlight is the fact that the closures of the range and of the domain of a maximally monotone operator are convex, which yields the classical Bunt–Motzkin result concerning the convexity of Chebyshev sets in Euclidean spaces. Results on local boundedness, surjectivity, and single-valuedness are also presented.

### 21.1 Minty's Theorem

A very useful characterization of maximal monotonicity is provided by the following theorem. Recall that  $F_A$  designates the Fitzpatrick function of a monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  (see Definition 20.42).

**Theorem 21.1 (Minty)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. Then  $A$  is maximally monotone if and only if  $\text{ran}(\text{Id} + A) = \mathcal{H}$ .*

*Proof.* Suppose first that  $\text{ran}(\text{Id} + A) = \mathcal{H}$  and fix  $(x, u) \in \mathcal{H} \times \mathcal{H}$  such that

$$(\forall (y, v) \in \text{gra } A) \quad \langle x - y \mid u - v \rangle \geq 0. \quad (21.1)$$

Since  $\text{ran}(\text{Id} + A) = \mathcal{H}$ , there exists  $(y, v) \in \mathcal{H}$  such that

$$v \in Ay \quad \text{and} \quad x + u = y + v \in (\text{Id} + A)y. \quad (21.2)$$

It follows from (21.1) and (21.2) that  $0 \leq \langle y - x \mid v - u \rangle = \langle y - x \mid x - y \rangle = -\|y - x\|^2 \leq 0$ . Hence  $y = x$  and thus  $v = u$ . Therefore,  $(x, u) = (y, v) \in \text{gra } A$ , and  $A$  is maximally monotone. Conversely, assume that  $A$  is maximally monotone. Then Proposition 20.48 implies that

$$\begin{aligned}
(\forall (x, u) \in \mathcal{H} \times \mathcal{H}) \quad 2F_A(x, u) + \|(x, u)\|^2 &= 2F_A(x, u) + \|x\|^2 + \|u\|^2 \\
&\geq 2\langle x \mid u \rangle + \|x\|^2 + \|u\|^2 \\
&= \|x + u\|^2 \\
&\geq 0.
\end{aligned} \tag{21.3}$$

Hence, Corollary 15.17 guarantees the existence of a vector  $(v, y) \in \mathcal{H} \times \mathcal{H}$  such that

$$(\forall (x, u) \in \mathcal{H} \times \mathcal{H}) \quad F_A(x, u) + \frac{1}{2}\|(x, u)\|^2 \geq \frac{1}{2}\|(x, u) + (v, y)\|^2, \tag{21.4}$$

which yields

$$\begin{aligned}
(\forall (x, u) \in \mathcal{H} \times \mathcal{H}) \quad F_A(x, u) &\geq \frac{1}{2}\|v\|^2 + \langle x \mid v \rangle + \frac{1}{2}\|y\|^2 + \langle y \mid u \rangle \\
&\geq -\langle y \mid v \rangle + \langle x \mid v \rangle + \langle y \mid u \rangle.
\end{aligned} \tag{21.5}$$

This and Proposition 20.47(i) imply that

$$(\forall (x, u) \in \text{gra } A) \quad \langle x \mid u \rangle \geq \frac{1}{2}\|v\|^2 + \langle x \mid v \rangle + \frac{1}{2}\|y\|^2 + \langle y \mid u \rangle \tag{21.6}$$

$$\geq -\langle y \mid v \rangle + \langle x \mid v \rangle + \langle y \mid u \rangle, \tag{21.7}$$

and hence that  $\langle x - y \mid u - v \rangle \geq 0$ . Since  $A$  is maximally monotone, we deduce that

$$v \in Ay. \tag{21.8}$$

Using (21.8) in (21.6), we obtain  $2\langle y \mid v \rangle \geq \|v\|^2 + 2\langle y \mid v \rangle + \|y\|^2 + 2\langle y \mid v \rangle$ . Hence  $0 \geq \|v\|^2 + 2\langle y \mid v \rangle + \|y\|^2 = \|y + v\|^2$  and thus

$$-v = y. \tag{21.9}$$

Combining (21.8) and (21.9) yields  $0 \in (\text{Id} + A)y \subset \text{ran}(\text{Id} + A)$ . Now fix  $w \in \mathcal{H}$  and define a maximally monotone operator  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto -w + Ax$ . Then the above reasoning shows that  $0 \in \text{ran}(\text{Id} + B)$  and hence that  $w \in \text{ran}(\text{Id} + A)$ .  $\square$

We now provide some applications of Theorem 21.1. First, we revisit Theorem 20.40 with an alternative proof.

**Theorem 21.2** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\partial f$  is maximally monotone.*

*Proof.* Combine Example 20.3, Proposition 16.35, and Theorem 21.1.  $\square$

**Proposition 21.3** *Let  $\mathbf{H}$  be a real Hilbert space, let  $x_0 \in \mathbf{H}$ , suppose that  $\mathcal{H} = L^2([0, T]; \mathbf{H})$ , and let  $A$  be the time-derivative operator (see Example 2.9)*

$$A: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \begin{cases} \{x'\}, & \text{if } x \in W^{1,2}([0, T]; \mathbf{H}) \text{ and } x(0) = x_0; \\ \emptyset, & \text{otherwise.} \end{cases} \tag{21.10}$$

*Then  $A$  is maximally monotone.*

*Proof.* Monotonicity is shown in Example 20.9. To show maximality, let us fix  $u \in L^2([0, T]; \mathbf{H})$ . In view of Theorem 21.1, we must show that there exists a solution  $x \in W^{1,2}([0, T]; \mathbf{H})$  to the evolution equation

$$\begin{cases} x(t) + x'(t) = u(t) & \text{a.e. on } ]0, T[ \\ x(0) = x_0. \end{cases} \quad (21.11)$$

Let us set  $v: [0, T] \rightarrow \mathbf{H}: t \mapsto e^t u(t)$ . Then  $v \in L^2([0, T]; \mathbf{H})$  and the function  $y \in W^{1,2}([0, T]; \mathbf{H})$  given by

$$y(0) = x_0 \quad \text{and} \quad (\forall t \in [0, T]) \quad y(t) = y(0) + \int_0^t v(s) ds \quad (21.12)$$

is therefore well defined. Now set  $x: [0, T] \rightarrow \mathbf{H}: t \mapsto e^{-t} y(t)$ . Then  $x \in L^2([0, T]; \mathbf{H})$ ,  $x(0) = x_0$ , and

$$x'(t) = -e^{-t} y(t) + e^{-t} y'(t) = -x(t) + u(t) \quad \text{a.e. on } ]0, T[. \quad (21.13)$$

Thus,  $x' \in L^2([0, T]; \mathbf{H})$ ,  $x \in W^{1,2}([0, T]; \mathbf{H})$ , and  $x$  solves (21.11).  $\square$

**Proposition 21.4** *Let  $C$  be a nonempty compact subset of  $\mathcal{H}$ . Suppose that the farthest-point operator  $\Phi_C$  of  $C$  defined in (20.13) is single-valued. Then  $-\Phi_C$  is maximally monotone and  $C$  is a singleton.*

*Proof.* It is clear that  $\text{dom } \Phi_C = \mathcal{H}$  since  $C$  is nonempty and compact. Set  $f_C: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto \|x - \Phi_C x\|$ , and take  $x$  and  $y$  in  $\mathcal{H}$ . Then for every  $z \in C$ , we have  $\|x - z\| \leq \|x - y\| + \|y - z\|$  and hence  $f_C(x) = \sup \|x - C\| \leq \|x - y\| + \sup \|y - C\| = \|x - y\| + f_C(y)$ . It follows that  $f_C$  is Lipschitz continuous with constant 1. Now let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  converging to  $x$ . Then

$$\|x_n - \Phi_C x_n\| = f_C(x_n) \rightarrow f_C(x) = \|x - \Phi_C x\|. \quad (21.14)$$

Assume that

$$\Phi_C x_n \not\rightarrow \Phi_C x. \quad (21.15)$$

After passing to a subsequence and relabeling if necessary, we assume that there exist  $\varepsilon \in \mathbb{R}_{++}$  and  $u \in C$  such that  $\|\Phi_C x_n - \Phi_C x\| \geq \varepsilon$  and  $\Phi_C x_n \rightarrow u$ . Taking the limit in (21.14) yields  $\|x - u\| = \|x - \Phi_C x\|$ , and hence  $u = \Phi_C x$ , which is impossible. Hence (21.15) is false and  $\Phi_C$  is therefore continuous. In view of Example 20.13 and Corollary 20.25,  $-\Phi_C$  is maximally monotone. By Theorem 21.1,  $\text{ran}(\text{Id} - \Phi_C) = \mathcal{H}$  and thus  $0 \in \text{ran}(\text{Id} - \Phi_C)$ . We deduce the existence of  $w \in \mathcal{H}$  such that  $0 = \|w - \Phi_C w\| = \sup \|w - C\|$ . Hence  $w \in C$  and therefore  $C = \{w\}$ .  $\square$

**Example 21.5** Set  $\mathbb{P} = \mathbb{N} \setminus \{0\}$ , let  $\mathcal{H} = \ell^2(\mathbb{P})$ , let  $(e_n)_{n \in \mathbb{P}}$  be the canonical orthonormal basis of  $\mathcal{H}$ , and set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \max \left\{ 1 + \langle x \mid e_1 \rangle, \sup_{2 \leq n \in \mathbb{N}} \langle x \mid \sqrt{n}e_n \rangle \right\}. \quad (21.16)$$

Then  $f \in \Gamma_0(\mathcal{H})$  and  $\partial f$  is maximally monotone. However,  $\text{gra } \partial f$  is not closed in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ . Furthermore,

$$\text{Argmin } f = \left\{ (\xi_n)_{n \in \mathbb{P}} \in \mathcal{H} \mid \xi_1 \leq -1, \sup_{2 \leq n \in \mathbb{N}} \xi_n \leq 0 \right\}. \quad (21.17)$$

*Proof.* It follows from Proposition 9.3 that  $f \in \Gamma(\mathcal{H})$ . Note that, for every integer  $n \geq 2$ ,  $0 = f(-e_1) < 1 = f(0) = f(e_n/\sqrt{n}) < 2 = f(e_1)$ ; hence  $f \in \Gamma_0(\mathcal{H})$  and  $\partial f$  is maximally monotone by Theorem 21.2. Since  $0 \notin \text{Argmin } f$ , it follows from Theorem 16.2 that

$$(0, 0) \notin \text{gra } \partial f. \quad (21.18)$$

Now take an integer  $n \geq 2$  and  $x \in \mathcal{H}$ . Then  $f(x) - f(e_n/\sqrt{n}) = f(x) - 1 \geq \langle x \mid \sqrt{n}e_n \rangle - \langle e_n/\sqrt{n} \mid \sqrt{n}e_n \rangle = \langle x - e_n/\sqrt{n} \mid \sqrt{n}e_n \rangle$ . Furthermore,  $f(x) - f(e_1) = f(x) - 2 \geq 1 + \langle x \mid e_1 \rangle - 2 = \langle x - e_1 \mid e_1 \rangle$ . Consequently,  $(\forall n \in \mathbb{P}) \sqrt{n}e_n \in \partial f(e_n/\sqrt{n})$  and thus

$$\{(e_n/\sqrt{n}, \sqrt{n}e_n)\}_{n \in \mathbb{P}} \subset \text{gra } \partial f. \quad (21.19)$$

In view of Example 3.31, there exists a net  $(\sqrt{n(a)}e_{n(a)})_{a \in A}$  that converges weakly to 0. Now set  $C = \{0\} \cup \{e_n/\sqrt{n}\}_{n \in \mathbb{P}}$ . Since  $e_n/\sqrt{n} \rightarrow 0$ , the set  $C$  is compact and the net  $(e_{n(a)}/\sqrt{n(a)})_{a \in A}$  lies in  $C$ . Thus, by Fact 1.11, there exists a subnet  $(e_{n(b)}/\sqrt{n(b)})_{b \in B}$  of  $(e_{n(a)}/\sqrt{n(a)})_{a \in A}$  that converges strongly to some point in  $C$ . Since  $(\sqrt{n(b)}e_{n(b)})_{b \in B}$  is a subnet of  $(\sqrt{n(a)}e_{n(a)})_{a \in A}$ , it is clear that

$$\sqrt{n(b)}e_{n(b)} \rightarrow 0. \quad (21.20)$$

We claim that

$$e_{n(b)}/\sqrt{n(b)} \rightarrow 0. \quad (21.21)$$

Assume that this is not true. Then  $e_m/\sqrt{m} = \lim_{b \in B} e_{n(b)}/\sqrt{n(b)}$  for some  $m \in \mathbb{P}$ . Since  $e_m/\sqrt{m}$  is an isolated point of  $C$ , the elements of the net  $(e_{n(b)}/\sqrt{n(b)})_{b \in B}$  are eventually equal to this point. In turn, this implies that the elements of the net  $(\sqrt{n(b)}e_{n(b)})_{b \in B}$  are eventually equal to  $\sqrt{m}e_m$ , which contradicts (21.20). We thus have verified (21.21). To sum up, (21.18)–(21.21) imply that  $(e_{n(b)}/\sqrt{n(b)}, \sqrt{n(b)}e_{n(b)})_{b \in B}$  lies in  $\text{gra } \partial f$ , that it converges to  $(0, 0)$  in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ , and that its limit lies outside  $\text{gra } \partial f$ . Now fix  $\varepsilon \in \mathbb{R}_{++}$  and assume that  $x = (\xi_n)_{n \in \mathbb{P}} \in \ell^2(\mathbb{P})$  satisfies  $f(x) \leq -\varepsilon$ . Then for every integer  $n \geq 2$ ,  $\sqrt{n}\xi_n \leq -\varepsilon \Rightarrow \varepsilon^2/n \leq \xi_n^2 \Rightarrow \|x\|^2 = +\infty$ , which is impossible. Hence  $\inf f(\mathcal{H}) \geq 0$ , and (21.17) is proven.  $\square$

**Remark 21.6** The proof of Example 21.5 implies that  $\text{gra } \partial f$  is not closed in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$  due to the existence of a net

$$(x_b, u_b)_{b \in B} = (e_{n(b)}/\sqrt{n(b)}, \sqrt{n(b)}e_{n(b)})_{b \in B} \quad (21.22)$$

in  $\text{gra } \partial f$  converging to  $(0, 0) \notin \text{gra } \partial f$  in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ , and that is unbounded. Clearly,

$$\lim \langle x_b \mid u_b \rangle = 1 \neq \langle 0 \mid 0 \rangle, \quad (21.23)$$

which shows that Lemma 2.36 is false without the assumption on boundedness.

## 21.2 The Debrunner–Flor Theorem

**Theorem 21.7 (Debrunner–Flor)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator such that  $\text{gra } A \neq \emptyset$ . Then*

$$(\forall w \in \mathcal{H})(\exists x \in \overline{\text{conv}} \text{ dom } A) \quad 0 \leq \inf_{(y,v) \in \text{gra } A} \langle y - x \mid v - (w - x) \rangle. \quad (21.24)$$

*Proof.* Set  $C = \overline{\text{conv}} \text{ dom } A$ . In view of Proposition 20.47(iii), we must show that  $(\forall w \in \mathcal{H})(\exists x \in C) F_A(x, w - x) \leq \langle x \mid w - x \rangle$ , i.e., that

$$(\forall w \in \mathcal{H}) \quad \min_{x \in \mathcal{H}} (F_A(x, w - x) + \|x\|^2 - \langle x \mid w \rangle + \iota_C(x)) \leq 0. \quad (21.25)$$

Let  $w \in \mathcal{H}$ . We consider two cases.

(a)  $w = 0$ : It suffices to show that

$$\min_{x \in \mathcal{H}} (F_A(x, -x) + (\|x\|^2 + \iota_C(x))) \leq 0. \quad (21.26)$$

Set  $q = (1/2)\|\cdot\|^2$ ,  $f: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty]: (y, x) \mapsto (1/2)F_A^*(2y, 2x)$ ,  $g = (q + \iota_C)^* = q - (1/2)d_C^2$  (by Example 13.5), and  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (y, x) \rightarrow x - y$ . We claim that

$$\inf_{(y,x) \in \mathcal{H} \times \mathcal{H}} (f(y, x) + g(L(y, x))) \geq 0. \quad (21.27)$$

To see this, fix  $(y, x) \in \mathcal{H} \times \mathcal{H}$ . By Proposition 20.51(ii),  $\text{dom } F_A^* \subset \overline{\text{conv}} \text{ ran } A \times \overline{\text{conv}} \text{ dom } A$ , and we thus have to consider only the case  $2x \in C$ . Then, since  $\langle \cdot \mid \cdot \rangle \leq F_A^*$  by Proposition 20.51(iii), we obtain

$$\begin{aligned} 0 &= 4 \langle y \mid x \rangle + \|x - y\|^2 - \|x + y\|^2 \\ &= \langle 2y \mid 2x \rangle + \|x - y\|^2 - \|(x - y) - 2x\|^2 \\ &\leq F_A^*(2y, 2x) + \|x - y\|^2 - d_C^2(x - y) \\ &= 2(f(y, x) + g(x - y)) \\ &= 2(f(y, x) + g(L(y, x))). \end{aligned} \quad (21.28)$$

Since  $\text{dom } g = \mathcal{H}$ , Theorem 15.23 implies that

$$\min_{x \in \mathcal{H}} (f^*(-L^*x) + g^*(x)) \leq 0. \quad (21.29)$$

Since  $f^* = (1/2)F_A$ ,  $g^* = q + \iota_C$ , and  $L^*: \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: x \mapsto (-x, x)$ , we see that (21.29) is equivalent to (21.26).

(b)  $w \neq 0$ : Let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be defined via  $\text{gra } B = -(0, w) + \text{gra } A$ . The above reasoning yields a point  $(x, -x) \in C \times \mathcal{H}$  such that  $\{(x, -x)\} \cup \text{gra } B$  is monotone. Therefore,  $\{(x, w - x)\} \cup \text{gra } A$  is monotone.  $\square$

**Theorem 21.8** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. Then there exists a maximally monotone extension  $\tilde{A}$  of  $A$  such that  $\text{dom } \tilde{A} \subset \overline{\text{conv}} \text{ dom } A$ .*

*Proof.* Set  $C = \overline{\text{conv}} \text{ dom } A$  and let  $\mathcal{M}$  be the set of all monotone extensions  $B$  of  $A$  such that  $\text{dom } B \subset C$ . Order  $\mathcal{M}$  partially via  $(\forall B_1 \in \mathcal{M})(\forall B_2 \in \mathcal{M}) B_1 \preceq B_2 \Leftrightarrow \text{gra } B_1 \subset \text{gra } B_2$ . Since every chain in  $\mathcal{M}$  has its union as an upper bound, Zorn's lemma (Fact 1.1) yields a maximal element  $\tilde{A}$ . Now let  $w \in \mathcal{H}$  and assume that  $w \in \mathcal{H} \setminus \text{ran}(\text{Id} + \tilde{A})$ . Theorem 21.7 provides  $x \in C$  such that  $0 \leq \inf_{(y,v) \in \text{gra } \tilde{A}} \langle y - x \mid v - (w - x) \rangle$ . Thus,  $(x, w - x) \notin \text{gra } \tilde{A}$  and hence  $\{(x, w - x)\} \cup \text{gra } \tilde{A}$  is the graph of an operator in  $\mathcal{M}$  that properly extends  $\tilde{A}$ . This contradicts the maximality of  $\tilde{A}$ . We deduce that  $\text{ran}(\text{Id} + \tilde{A}) = \mathcal{H}$  and, by Theorem 21.1, that  $\tilde{A}$  is maximally monotone.  $\square$

## 21.3 Domain and Range

**Definition 21.9** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $x \in \mathcal{H}$ . Then  $A$  is *locally bounded* at  $x$  if there exists  $\delta \in \mathbb{R}_{++}$  such that  $A(B(x; \delta))$  is bounded.

**Proposition 21.10** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone, set  $Q_1: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, u) \mapsto x$ , and suppose that  $z \in \text{int } Q_1(\text{dom } F_A)$ . Then  $A$  is locally bounded at  $z$ .*

*Proof.* Define

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sup_{(y,v) \in \text{gra } A} \frac{\langle x - y \mid v \rangle}{1 + \|y\|}, \quad (21.30)$$

and observe that  $f \in \Gamma(\mathcal{H})$  by Proposition 9.3. Fix  $(x, u) \in \mathcal{H} \times \mathcal{H}$  and let  $(y, v) \in \text{gra } A$ . Then  $\langle x \mid v \rangle + \langle y \mid u \rangle - \langle y \mid v \rangle \leq F_A(x, u)$  and hence  $\langle x - y \mid v \rangle \leq F_A(x, u) - \langle y \mid u \rangle \leq \max\{F_A(x, u), \|u\|\}(1 + \|y\|)$ . Dividing by  $1 + \|y\|$  and then taking the supremum over  $(y, v) \in \text{gra } A$  yields

$$f(x) \leq \max\{F_A(x, u), \|u\|\}. \quad (21.31)$$

Hence  $Q_1(\text{dom } F_A) \subset \text{dom } f$  and thus  $z \in \text{int dom } f$ . Corollary 8.30 implies the existence of  $\delta \in \mathbb{R}_{++}$  such that  $\sup f(B(z; 2\delta)) \leq 1 + f(z)$ . Hence, for every  $(y, v) \in \text{gra } A$  and every  $w \in B(0; 2\delta)$ ,  $\langle z + w - y \mid v \rangle \leq (1 + f(z))(1 + \|y\|)$  or, equivalently,

$$2\delta\|v\| + \langle z - y \mid v \rangle \leq (1 + f(z))(1 + \|y\|). \quad (21.32)$$

Now assume that  $(y, v) \in \text{gra } A$  and  $y \in B(z; \delta)$ . Using Cauchy–Schwarz and (21.32), we deduce that

$$\begin{aligned} \delta\|v\| &= 2\delta\|v\| - \delta\|v\| \\ &\leq 2\delta\|v\| + \langle z - y \mid v \rangle \\ &\leq (1 + f(z))(1 + \|y\|) \\ &\leq (1 + f(z))(1 + \|y - z\| + \|z\|) \\ &\leq (1 + f(z))(1 + \delta + \|z\|). \end{aligned} \quad (21.33)$$

It follows that  $\sup \|A(B(z; \delta))\| \leq (1 + f(z))(1 + \delta + \|z\|)/\delta < +\infty$ .  $\square$

**Proposition 21.11** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and set  $Q_1: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, u) \mapsto x$ . Then*

$$\text{int dom } A \subset \text{int } Q_1(\text{dom } F_A) \subset \text{dom } A \subset Q_1(\text{dom } F_A) \subset \overline{\text{dom } A}. \quad (21.34)$$

Consequently,  $\text{int dom } A = \text{int } Q_1(\text{dom } F_A)$  and  $\overline{\text{dom } A} = \overline{Q_1(\text{dom } F_A)}$ .

*Proof.* Fix  $x \in \mathcal{H}$ . If  $x \in \text{dom } A$ , then there exists  $u \in \mathcal{H}$  such that  $(x, u) \in \text{gra } A$ . Hence, by Proposition 20.47(i),  $F_A(x, u) = \langle x \mid u \rangle \in \mathbb{R}$ , which implies that  $x \in Q_1(\text{dom } F_A)$ . As a result, the third, and thus the first, inclusions in (21.34) are verified. Now assume that  $x \in Q_1(\text{dom } F_A)$ , say  $F_A(x, u) < +\infty$  for some  $u \in \mathcal{H}$ . Let  $\varepsilon \in ]0, 1[$ , set

$$\beta = \max \{F_A(x, u), \|u\|, \|x\| \|u\|\}, \quad (21.35)$$

and pick  $\lambda \in ]0, \varepsilon^2/(9\beta)[$ . Define  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  via

$$\text{gra } B = \{((y - x)/\lambda, v) \mid (y, v) \in \text{gra } A\}. \quad (21.36)$$

Then  $B$  is maximally monotone and therefore Theorem 21.1 yields  $0 \in \text{ran}(\text{Id} + B)$ . Hence there exists  $z \in \mathcal{H}$  such that  $-z \in Bz$ , and thus  $(y, v) \in \text{gra } A$  such that  $(z, -z) = ((y - x)/\lambda, v)$ . In turn,  $\langle x \mid v \rangle + \langle y \mid u \rangle - \langle y \mid v \rangle \leq F_A(x, u) \leq \beta$  and

$$-\langle y - x \mid v \rangle + \langle y - x \mid u \rangle \leq \beta - \langle x \mid u \rangle \leq \beta + \|x\| \|u\| \leq 2\beta. \quad (21.37)$$

Consequently,  $-\langle y - x \mid v \rangle = -\langle y - x \mid -z \rangle = \|y - x\|^2/\lambda$ ,  $\langle y - x \mid u \rangle \geq -\|y - x\| \|u\| \geq -\beta\|y - x\|$ , and we obtain

$$\|y - x\|^2 - \lambda\beta\|y - x\| - 2\lambda\beta \leq 0. \quad (21.38)$$

Hence  $\|y - x\|$  lies between the roots of the quadratic function  $\rho \mapsto \rho^2 - \lambda\beta\rho - 2\lambda\beta$  and so, in particular,  $\|y - x\| \leq (\lambda\beta + \sqrt{(\lambda\beta)^2 + 8(\lambda\beta)})/2 \leq \sqrt{\lambda\beta(\lambda\beta + 8)} < \sqrt{(\varepsilon^2/9)(1 + 8)} = \varepsilon$ . Thus, the fourth inclusion in (21.34) holds, i.e.,

$$Q_1(\text{dom } F_A) \subset \overline{\text{dom } A}. \quad (21.39)$$

It remains to verify the second inclusion in (21.34). Assume that  $x \in \text{int } Q_1(\text{dom } F_A)$ . By (21.39), there exists a sequence  $(x_n)_{n \in \mathbb{N}}$  in  $\text{dom } A$  such that  $x_n \rightarrow x$ . Let  $(u_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  such that  $(x_n, u_n)_{n \in \mathbb{N}}$  lies in  $\text{gra } A$ . It follows from Proposition 21.10 that  $(u_n)_{n \in \mathbb{N}}$  is bounded. Hence  $(u_n)_{n \in \mathbb{N}}$  possesses a weakly convergent subsequence, say  $u_{k_n} \rightharpoonup u \in \mathcal{H}$ . Proposition 20.33(i) implies that  $(x, u) \in \text{gra } A$  and thus  $x \in \text{dom } A$ .  $\square$

**Corollary 21.12** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, and set  $Q_1: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, u) \mapsto x$  and  $Q_2: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, u) \mapsto u$ . Then  $\overline{\text{dom } A} = \overline{Q_1(\text{dom } F_A)}$ ,  $\text{int } \text{dom } A = \text{int } Q_1(\text{dom } F_A)$ ,  $\overline{\text{ran } A} = \overline{Q_2(\text{dom } F_A)}$ , and  $\text{int } \text{ran } A = \text{int } Q_2(\text{dom } F_A)$ . Consequently, the sets  $\overline{\text{dom } A}$ ,  $\text{int } \text{dom } A$ ,  $\overline{\text{ran } A}$ , and  $\text{int } \text{ran } A$  are convex.*

*Proof.* The first claims follow from Proposition 21.11, applied to  $A$  and  $A^{-1}$ , and from Proposition 20.47(vi). On the one hand, since  $F_A$  is convex, its domain is convex. On the other hand, since  $Q_1$  and  $Q_2$  are linear, the sets  $Q_1(\text{dom } F_A)$  and  $Q_2(\text{dom } F_A)$  are convex by Proposition 3.5, and so are their closures and interiors by Proposition 3.36(i)&(ii).  $\square$

**Corollary 21.13 (Bunt–Motzkin)** *Suppose that  $\mathcal{H}$  is finite-dimensional and let  $C$  be a Chebyshev subset of  $\mathcal{H}$ . Then  $C$  is closed and convex.*

*Proof.* It is clear that  $\text{dom } P_C = \mathcal{H}$  and that  $P_C$  is single-valued. Remark 3.9(i) implies that  $C$  is closed. Thus  $\overline{C} = C = \text{ran } P_C$ . On the one hand,  $P_C$  is continuous by Proposition 3.10. On the other hand,  $P_C$  is monotone by Example 20.12. Altogether, we deduce from Corollary 20.25 that  $P_C$  is maximally monotone. Therefore, by Corollary 21.12,  $\overline{\text{ran } P_C} = \overline{C} = C$  is convex.  $\square$

## 21.4 Local Boundedness and Surjectivity

**Proposition 21.14** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \text{dom } A$ . Then  $\text{rec}(Ax) = N_{\overline{\text{dom } A}}x$ .*

*Proof.* Fix  $u \in Ax$  and  $w \in \mathcal{H}$ , and assume that  $w \in N_{\overline{\text{dom } A}}x$ . Then  $(\forall(y, v) \in \text{gra } A) 0 \leq \langle x - y \mid w \rangle \leq \langle x - y \mid w \rangle + \langle x - y \mid u - v \rangle = \langle x - y \mid (u + w) - v \rangle$ . The maximal monotonicity of  $A$  implies that  $u + w \in Ax$ . Hence  $w + Ax \subset Ax$ , i.e.,  $w \in \text{rec}(Ax)$ . Now assume that  $w \in \text{rec}(Ax)$ , which implies that



$(\forall \rho \in \mathbb{R}_{++}) u + \rho w \in Ax$ . Let  $y \in \text{dom } A$  and  $v \in Ay$ . Then  $(\forall \rho \in \mathbb{R}_{++})$   
 $0 \leq \langle x - y \mid (u + \rho w) - v \rangle = \langle x - y \mid u - v \rangle + \rho \langle x - y \mid w \rangle$ . Since  $\rho$  can be  
arbitrarily large, we have  $0 \leq \langle x - y \mid w \rangle$ . Therefore,  $w \in N_{\overline{\text{dom } A}} x$ .  $\square$

The next result, combined with Theorem 21.2, provides an extension to Proposition 16.14(iii).

**Theorem 21.15 (Rockafellar–Veselý)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \mathcal{H}$ . Then  $A$  is locally bounded at  $x$  if and only if  $x \notin \text{bdry dom } A$ .*

*Proof.* Let  $S$  be the set of all points at which  $A$  is locally bounded. Clearly,  $\mathcal{H} \setminus \text{dom } A \subset S$ . In addition, Proposition 21.10 and Proposition 21.11 imply that  $\text{int dom } A \subset S$ . We claim that

$$S \cap \overline{\text{dom } A} = S \cap \text{dom } A. \quad (21.40)$$

Let  $x \in S \cap \overline{\text{dom } A}$ . Then there exists a sequence  $(x_n, u_n)_{n \in \mathbb{N}}$  in  $\text{gra } A$  such that  $x_n \rightarrow x$  and such that  $(u_n)_{n \in \mathbb{N}}$  is bounded. After passing to a subsequence if necessary, we assume that  $u_n \rightharpoonup u$ . Now Proposition 20.33(i) yields  $(x, u) \in \text{gra } A$ . Hence  $x \in \text{dom } A$ , which verifies (21.40). It remains to show that  $S \cap \text{bdry dom } A = \emptyset$ . Assume to the contrary that  $x \in S \cap \text{bdry dom } A$  and take  $\delta \in \mathbb{R}_{++}$  such that  $A(B(x; 2\delta))$  is bounded. Set  $C = \overline{\text{dom } A}$ , which is convex by Corollary 21.12. Theorem 7.4 guarantees the existence of a support point  $z$  of  $C$  in  $(\text{bdry } C) \cap B(x; \delta)$ . Take  $w \in N_C z \setminus \{0\}$ . Since  $B(z; \delta) \subset B(x; 2\delta)$ ,  $z \in S \cap \text{bdry } C$ . Hence, by (21.40),  $z \in S \cap \text{dom } A \cap \text{bdry } C$ . Consequently,  $Az \neq \emptyset$  and, by Proposition 21.14,  $w \in \text{rec}(Az)$ . This implies that  $Az$  is unbounded, which, since  $z \in S$ , is impossible.  $\square$

**Corollary 21.16** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then  $A$  is locally bounded everywhere on  $\mathcal{H}$  if and only if  $\text{dom } A = \mathcal{H}$ .*

**Corollary 21.17** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and at most single-valued. Then  $A$  is strong-to-weak continuous everywhere on  $\text{int dom } A$ .*

*Proof.* Fix a point  $x \in \text{int dom } A$  and a sequence  $(x_n)_{n \in \mathbb{N}}$  that converges to  $x \in \text{int dom } A$ . By Theorem 21.15,  $(Ax_n)_{n \in \mathbb{N}}$  is bounded. Let  $y$  be a weak sequential cluster point of  $(Ax_n)_{n \in \mathbb{N}}$ , say  $Ax_{k_n} \rightharpoonup y$ . Proposition 20.33(i) implies that  $(x, y) \in \text{gra } A$ . Since  $A$  is at most single-valued, we deduce that  $y = Ax$ . It follows from Lemma 2.38 that  $Ax_n \rightharpoonup Ax$ .  $\square$

The verification of the next result is left as Exercise 21.7.

**Example 21.18** Let  $\mathcal{H} = \ell^2(\mathbb{N})$  and set  $A: \mathcal{H} \rightarrow \mathcal{H}: x = (\xi_n)_{n \in \mathbb{N}} \mapsto (\xi_n/2^n)_{n \in \mathbb{N}}$ . Then  $A$  is maximally monotone, locally bounded everywhere on  $\mathcal{H}$ , and  $\text{dom } A = \mathcal{H}$ . Now set  $B = A^{-1}$ . Then  $B$  is maximally monotone, nowhere locally bounded, nowhere continuous, and  $\text{dom } B$  is a dense proper linear subspace of  $\mathcal{H}$ .

**Corollary 21.19** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then  $A$  is surjective if and only if  $A^{-1}$  is locally bounded everywhere on  $\mathcal{H}$ .*

**Corollary 21.20** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator such that*

$$\lim_{\|x\| \rightarrow +\infty} \inf \|Ax\| = +\infty. \quad (21.41)$$

*Then  $A$  is surjective.*

*Proof.* In view of Corollary 21.19, let us show that  $A^{-1}$  is locally bounded on  $\mathcal{H}$ . Assume that  $A^{-1}$  is not locally bounded at  $u \in \mathcal{H}$ . Then there exists a sequence  $(x_n, u_n)_{n \in \mathbb{N}}$  in  $\text{gra } A$  such that  $u_n \rightarrow u$  and  $\|x_n\| \rightarrow +\infty$ . Hence,  $+\infty = \lim_{\|x\| \rightarrow +\infty} \inf \|Ax\| = \liminf \|Ax_n\| \leq \lim \|u_n\| = \|u\|$ , which is impossible.  $\square$

**Corollary 21.21** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone with bounded domain. Then  $A$  is surjective.*

## 21.5 Kenderov's Theorem and Fréchet Differentiability

**Theorem 21.22 (Kenderov)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator such that  $\text{int dom } A \neq \emptyset$ . Then there exists a subset  $C$  of  $\text{int dom } A$  that is a dense  $G_\delta$  subset of  $\overline{\text{dom } A}$  and such that, for every point  $x \in C$ ,  $Ax$  is a singleton and every selection of  $A$  is strong-to-strong continuous at  $x$ .*

*Proof.* Let  $(\varepsilon_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  converging to 0, fix  $n \in \mathbb{N}$ , and define the set  $C_n$  by requiring that  $y \in C_n$  if and only if there exists an open neighborhood  $V$  of  $y$  such that  $\text{diam } A(V) < \varepsilon_n$ . It is clear that  $C_n$  is an open subset of  $\text{int dom } A$ . To show that  $C_n$  is dense in  $\text{int dom } A$ , take  $y \in \text{int dom } A$ . Since  $A$  is locally bounded at  $y$  by Theorem 21.15, there exists an open bounded set  $D$  such that  $A(D)$  is bounded. Proposition 18.4 yields  $z \in \mathcal{H}$  and  $\alpha \in \mathbb{R}_{++}$  such that

$$\text{diam } S < \varepsilon_n, \quad \text{where } S = \{u \in A(D) \mid \langle z \mid u \rangle > \sigma_{A(D)}(z) - \alpha\}. \quad (21.42)$$

Take  $u_1 \in S$ . Then there exists  $x_1 \in D$  such that  $u_1 \in Ax_1$ . Now let  $\gamma \in \mathbb{R}_{++}$  be small enough to satisfy  $x_0 = x_1 + \gamma z \in D$ , and take  $u_0 \in Ax_0$ . Then  $u_0 \in A(D)$  and  $0 \leq \langle x_0 - x_1 \mid u_0 - u_1 \rangle = \gamma \langle z \mid u_0 - u_1 \rangle$ , which implies that  $\langle z \mid u_0 \rangle \geq \langle z \mid u_1 \rangle$  and so  $u_0 \in S$ . Thus,

$$x_0 \in D \quad \text{and} \quad Ax_0 \subset S. \quad (21.43)$$

In view of Proposition 20.33(i), there exists  $\rho \in \mathbb{R}_{++}$  such that

$$B(x_0; \rho) \subset D \quad \text{and} \quad A(B(x_0; \rho)) \subset \{u \in \mathcal{H} \mid \langle z \mid u \rangle > \sigma_{A(D)}(z) - \alpha\}. \quad (21.44)$$

Consequently,  $A(B(x_0; \rho)) \subset S$ ; hence  $\text{diam } A(B(x_0; \rho)) < \varepsilon_n$ , and thus  $x_0 \in C_n$ .

We have shown that for every  $n \in \mathbb{N}$ ,  $C_n$  is a dense open subset of  $\text{int dom } A$ . Set  $C = \bigcap_{n \in \mathbb{N}} C_n$ . By Corollary 1.44,  $C$  is a dense  $G_\delta$  subset of  $\overline{\text{dom } A}$ . Take  $x \in C$ . From the definition of  $C$ , it is clear that  $Ax$  is a singleton and that every selection of  $A$  is norm-to-norm continuous at  $x$ .  $\square$

**Corollary 21.23** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{int dom } f \neq \emptyset$ . Then there exists a subset  $C$  of  $\text{int dom } f$  such that  $C$  is a dense  $G_\delta$  subset of  $\overline{\text{dom } f}$ , and such that  $f$  is Fréchet differentiable on  $C$ .*

## Exercises

**Exercise 21.1** Let  $N$  be strictly positive integer, and let  $A \in \mathbb{R}^{N \times N}$  be such that  $(\forall x \in \mathbb{R}^N) \langle x | Ax \rangle \geq 0$ . Show that  $\text{Id} + A$  is surjective without appealing to Theorem 21.1.

**Exercise 21.2** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\varepsilon \in \mathbb{R}_{++}$ , denote the  $\varepsilon$ -enlargement of  $A$  by  $A^\varepsilon$  (see Exercise 20.15), and let  $(x, u) \in \text{gra } A^\varepsilon$ . Use Theorem 21.1 to show that  $d_{\text{gra } A}(x, u) \leq \sqrt{2\varepsilon}$ .

**Exercise 21.3** Let  $f \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable on  $\mathcal{H}$ . Show that  $\nabla f$  is strong-to-weak continuous without appealing to Corollary 17.33(i).

**Exercise 21.4** Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be monotone. Show the equivalence of the following: (i)  $A$  is maximally monotone; (ii)  $A$  is strong-to-weak continuous; (iii)  $A$  is hemicontinuous.

**Exercise 21.5** Let  $f \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable on  $\mathcal{H}$ . Show that if  $\lim_{\|x\| \rightarrow +\infty} \|\nabla f(x)\| = +\infty$ , then  $\text{dom } f^* = \mathcal{H}$ .

**Exercise 21.6** Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be monotone and hemicontinuous. Suppose that  $\mathcal{H}$  is finite-dimensional or that  $A$  is linear. Show that  $A$  is maximally monotone and continuous.

**Exercise 21.7** Verify Example 21.18.

**Exercise 21.8** Suppose that  $\mathcal{H} \neq \{0\}$ , let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator such that  $\text{gra } A$  is bounded, and let  $\tilde{A}$  be a maximally monotone extension of  $A$ . Show that  $\text{dom } \tilde{A}$  or  $\text{ran } \tilde{A}$  is unbounded.

**Exercise 21.9** Prove Corollary 21.23.

**Exercise 21.10** Show that Theorem 21.22 fails if  $\text{int dom } A = \emptyset$ .



# Chapter 22

## Stronger Notions of Monotonicity

This chapter collects basic results on various stronger notions of monotonicity (para, strict, uniform, strong, and cyclic) and their relationships to properties of convex functions. A fundamental result is Rockafellar's characterization of maximally cyclically monotone operators as subdifferential operators and a corresponding uniqueness result for the underlying convex function.

### 22.1 Para, Strict, Uniform, and Strong Monotonicity

**Definition 22.1** An operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is

(i) *paramonotone* if it is monotone and

$$(\forall (x, u) \in \text{gra } A)(\forall (y, v) \in \text{gra } A) \quad \langle x - y \mid u - v \rangle = 0 \Rightarrow (x, v) \in \text{gra } A; \quad (22.1)$$

(ii) *strictly monotone* if

$$(\forall (x, u) \in \text{gra } A)(\forall (y, v) \in \text{gra } A) \quad x \neq y \Rightarrow \langle x - y \mid u - v \rangle > 0; \quad (22.2)$$

(iii) *uniformly monotone* with modulus  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  if  $\phi$  is increasing, vanishes only at 0, and

$$(\forall (x, u) \in \text{gra } A)(\forall (y, v) \in \text{gra } A) \quad \langle x - y \mid u - v \rangle \geq \phi(\|x - y\|); \quad (22.3)$$

(iv) *strongly monotone* with constant  $\beta \in \mathbb{R}_{++}$  if  $A - \beta \text{Id}$  is monotone, i.e.,

$$(\forall (x, u) \in \text{gra } A)(\forall (y, v) \in \text{gra } A) \quad \langle x - y \mid u - v \rangle \geq \beta \|x - y\|^2. \quad (22.4)$$

It is clear that strong monotonicity implies uniform monotonicity, which itself implies strict monotonicity, which itself implies paramonotonicity, which itself implies monotonicity.

**Remark 22.2** The notions of strict, uniform, and strong monotonicity of  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  can naturally be localized to a subset  $C$  of  $\text{dom } A$ . For instance,  $A$  is uniformly monotone on  $C$  if there exists an increasing function  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  vanishing only at 0 such that

$$(\forall x \in C)(\forall y \in C)(\forall u \in Ax)(\forall v \in Ay) \quad \langle x - y \mid u - v \rangle \geq \phi(\|x - y\|). \quad (22.5)$$

**Example 22.3** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Then the following hold:

- (i)  $\partial f$  is paramonotone.
- (ii) Suppose that  $f$  is strictly convex. Then  $\partial f$  is strictly monotone.
- (iii) Suppose that  $f$  is uniformly convex with modulus  $\phi$ . Then  $\partial f$  is uniformly monotone with modulus  $2\phi$ .
- (iv) Suppose that  $f$  is strongly convex with constant  $\beta \in \mathbb{R}_{++}$ . Then  $\partial f$  is strongly monotone with constant  $\beta$ .

*Proof.* We assume that  $\text{dom } \partial f$  contains at least two elements since the conclusion is clear otherwise. Take  $(x, u)$  and  $(y, v)$  in  $\text{gra } \partial f$ , and  $\alpha \in ]0, 1[$ . Then (16.1) yields

$$\alpha \langle x - y \mid v \rangle = \langle (\alpha x + (1 - \alpha)y) - y \mid v \rangle \leq f(\alpha x + (1 - \alpha)y) - f(y). \quad (22.6)$$

(i): Assume that  $\langle x - y \mid u - v \rangle = 0$ . It follows from Proposition 16.9 that  $0 = \langle x - y \mid u - v \rangle = \langle x \mid u \rangle + \langle y \mid v \rangle - \langle x \mid v \rangle - \langle y \mid u \rangle = f(x) + f^*(u) + f(y) + f^*(v) - \langle x \mid v \rangle - \langle y \mid u \rangle = (f(x) + f^*(v) - \langle x \mid v \rangle) + (f(y) + f^*(u) - \langle y \mid u \rangle)$ . Hence, by Proposition 13.13,  $v \in \partial f(x)$  and  $u \in \partial f(y)$ .

(ii): Assume that  $x \neq y$ . Then (22.6) and (8.3) imply that  $\langle x - y \mid v \rangle < f(x) - f(y)$ . Likewise,  $\langle y - x \mid u \rangle < f(y) - f(x)$ . Adding these two inequalities yields (22.2).

(iii): It follows from (22.6) and (10.1) that  $\langle x - y \mid v \rangle + (1 - \alpha)\phi(\|x - y\|) \leq f(x) - f(y)$ . Letting  $\alpha \downarrow 0$ , we obtain  $\langle x - y \mid v \rangle + \phi(\|x - y\|) \leq f(x) - f(y)$ . Likewise,  $\langle y - x \mid u \rangle + \phi(\|x - y\|) \leq f(y) - f(x)$ . Adding these two inequalities yields  $\langle x - y \mid u - v \rangle \geq 2\phi(\|x - y\|)$ .

(iv): Since  $f$  is uniformly convex with modulus  $t \mapsto (1/2)\beta t^2$ , we deduce from (iii) that  $\partial f$  is uniformly monotone with modulus  $t \mapsto \beta t^2$ , i.e., that  $\partial f$  is strongly monotone with constant  $\beta$ .  $\square$

**Example 22.4** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex, let  $C$  be a nonempty subset of  $\text{dom } \partial f$ , and suppose that  $f$  is uniformly convex on  $C$ . Then  $\partial f$  is uniformly monotone on  $C$ .

*Proof.* Use Remark 22.2 and proceed as in the proof of Example 22.3(iii).  $\square$

The next result refines Example 20.7.

**Example 22.5** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , let  $\alpha \in [-1, 1]$ , and set  $A = \text{Id} + \alpha T$ . Then the following hold:

- (i) Suppose that  $T$  is *strictly nonexpansive*, i.e.,  $(\forall x \in D)(\forall y \in D) x \neq y \Rightarrow \|Tx - Ty\| < \|x - y\|$ . Then  $A$  is strictly monotone.
- (ii) Suppose that  $T$  is  $\beta$ -Lipschitz continuous, with  $\beta \in [0, 1[$ . Then  $A$  is strongly monotone with modulus  $1 - |\alpha|\beta$ .

*Proof.* (i): This follows from (20.4).

(ii): It follows from (20.4) that, for every  $x$  and  $y$  in  $D$ ,  $\langle x - y \mid Ax - Ay \rangle \geq (1 - |\alpha|\beta)\|x - y\|^2$ .  $\square$

**Example 22.6** Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $\beta \in \mathbb{R}_{++}$ , let  $T: D \rightarrow \mathcal{H}$ , and set  $A = T^{-1}$ . Then  $T$  is  $\beta$ -cocoercive if and only if  $A$  is strongly monotone with constant  $\beta$ .

*Proof.* Take  $(x, u)$  and  $(y, v)$  in  $\text{gra } A$ . Then  $u \in T^{-1}x$  and  $v \in T^{-1}y$ . Hence,  $x = Tu$ ,  $y = Tv$ , and  $\langle u - v \mid x - y \rangle \geq \beta\|x - y\|^2 \Leftrightarrow \langle u - v \mid Tu - Tv \rangle \geq \beta\|Tu - Tv\|^2$ . Thus, the conclusion follows from (4.5) and (22.4).  $\square$

**Example 22.7** Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be linear. Then

- (i)  $A$  is strictly monotone if and only if  $(\forall x \in \mathcal{H} \setminus \{0\}) \langle x \mid Ax \rangle > 0$ .
- (ii)  $A$  is strongly monotone with constant  $\beta \in \mathbb{R}_{++}$  if and only if  $(\forall x \in \mathcal{H}) \langle x \mid Ax \rangle \geq \beta\|x\|^2$ .

**Proposition 22.8** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and suppose that one of the following holds:

- (i)  $A$  is uniformly monotone with a supercoercive modulus.
- (ii)  $A$  is strongly monotone.

Then  $\lim_{\|x\| \rightarrow +\infty} \inf \|Ax\| = +\infty$  and  $A$  is surjective.

*Proof.* (i): Let  $\phi$  be the modulus of uniform convexity of  $A$  and fix  $(y, v) \in \text{gra } A$ . Then (22.3) and Cauchy–Schwarz yield

$$\begin{aligned} (\forall (x, u) \in \text{gra } A) \quad & \|x - y\| \|u\| \geq \langle x - y \mid u \rangle \\ & = \langle x - y \mid u - v \rangle + \langle x - y \mid v \rangle \\ & \geq \phi(\|x - y\|) - \|x - y\| \|v\|. \end{aligned} \quad (22.7)$$

Accordingly, since  $\lim_{t \rightarrow +\infty} \phi(t)/t = +\infty$ , we have  $\inf_{u \in Ax} \|u\| \rightarrow +\infty$  as  $\|x\| \rightarrow +\infty$ . In view of Corollary 21.20, the proof is complete.

(ii) $\Rightarrow$ (i): Clear.  $\square$

**Example 22.9** Let  $A: \mathcal{H} \rightarrow \mathcal{H}$  be monotone and hemicontinuous, and let  $r \in \mathcal{H}$ . Suppose that one of the following holds:

- (i)  $A$  is strictly monotone and  $\lim_{\|x\| \rightarrow +\infty} \|Ax\| = +\infty$ .
- (ii)  $A$  is uniformly monotone with a supercoercive modulus.

(iii)  $A$  is strongly monotone.

Then the equation  $Ax = r$  has exactly one solution.

*Proof.* By Proposition 20.24,  $A$  is maximally monotone.

(i): The existence of a solution follows from Corollary 21.20, and the uniqueness from (22.2).

(ii)&(iii): On the one hand,  $A$  is strictly monotone. On the other hand,  $\lim_{\|x\| \rightarrow +\infty} \|Ax\| = +\infty$  by Proposition 22.8. Altogether, the conclusion follows from (i).  $\square$

## 22.2 Cyclic Monotonicity

**Definition 22.10** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . Then  $A$  is *n-cyclically monotone* if, for every  $(x_1, \dots, x_{n+1}) \in \mathcal{H}^{n+1}$  and  $(u_1, \dots, u_n) \in \mathcal{H}^n$ ,

$$\left. \begin{array}{c} (x_1, u_1) \in \text{gra } A, \\ \vdots \\ (x_n, u_n) \in \text{gra } A, \\ x_{n+1} = x_1, \end{array} \right\} \Rightarrow \sum_{i=1}^n \langle x_{i+1} - x_i \mid u_i \rangle \leq 0. \quad (22.8)$$

If  $A$  is  $n$ -cyclically monotone for every integer  $n \geq 2$ , then  $A$  is *cyclically monotone*. If  $A$  is cyclically monotone and if there exists no cyclically monotone operator  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\text{gra } B$  properly contains  $\text{gra } A$ , then  $A$  is *maximally cyclically monotone*.

It is clear that the notions of monotonicity and 2-cyclic monotonicity coincide, and that  $n$ -cyclic monotonicity implies  $m$ -cyclic monotonicity for every  $m \in \{2, \dots, n\}$ . A maximally monotone operator that is cyclically monotone is also maximally cyclically monotone.

**Proposition 22.11** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper. Then  $\partial f$  is cyclically monotone.

*Proof.* Fix an integer  $n \geq 2$ . For every  $i \in \{1, \dots, n\}$ , take  $(x_i, u_i) \in \text{gra } \partial f$ . Set  $x_{n+1} = x_1$ . Then (16.1) yields

$$(\forall i \in \{1, \dots, n\}) \quad \langle x_{i+1} - x_i \mid u_i \rangle \leq f(x_{i+1}) - f(x_i). \quad (22.9)$$

Adding up these inequalities, we obtain  $\sum_{i=1}^n \langle x_{i+1} - x_i \mid u_i \rangle \leq 0$ .  $\square$

However, there are maximally monotone operators that are not 3-cyclically monotone.



**Example 22.12** Suppose that  $\mathcal{H} = \mathbb{R}^2$ , and set

$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}. \quad (22.10)$$

Then  $A$  is maximally monotone but not 3-cyclically monotone.

*Proof.* The maximal monotonicity of  $A$  follows from Example 20.30. Observe that  $A^2 = -\text{Id}$  and that  $(\forall x \in \mathcal{H}) \|Ax\| = \|x\|$  and  $\langle x | Ax \rangle = 0$ . Now let  $x_1 \in \mathcal{H} \setminus \{0\}$ , and set  $x_2 = Ax_1$ ,  $x_3 = Ax_2 = A^2x_1 = -x_1$ , and  $x_4 = x_1$ . Then

$$\begin{aligned} \sum_{i=1}^3 \langle x_{i+1} - x_i | Ax_i \rangle &= \langle x_2 | Ax_1 \rangle + \langle x_3 | Ax_2 \rangle + \langle x_1 | Ax_3 \rangle \\ &= \|Ax_1\|^2 + \langle -x_1 | -x_1 \rangle + \langle x_1 | -Ax_1 \rangle \\ &= 2\|x_1\|^2 \\ &> 0. \end{aligned} \quad (22.11)$$

Therefore,  $A$  is not 3-cyclically monotone.  $\square$

**Remark 22.13** The operator  $A$  of Example 22.12 is the rotator by  $\pi/2$ . More generally, let  $\theta \in [0, \pi/2]$ , let  $n \geq 2$  be an integer, and let  $A$  be the rotator in  $\mathbb{R}^2$  by  $\theta$ . Then  $A$  is  $n$ -cyclically monotone if and only if  $\theta \in [0, \pi/n]$ ; see [4] and [27].

## 22.3 Rockafellar's Cyclic Monotonicity Theorem

**Theorem 22.14 (Rockafellar)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ . Then  $A$  is maximally cyclically monotone if and only if there exists  $f \in \Gamma_0(\mathcal{H})$  such that  $A = \partial f$ .*

*Proof.* Suppose that  $A = \partial f$  for some  $f \in \Gamma_0(\mathcal{H})$ . Theorem 21.2 and Proposition 22.11 imply that  $A$  is maximally monotone and cyclically monotone. Hence  $A$  is maximally cyclically monotone. Conversely, suppose that  $A$  is maximally cyclically monotone. Then  $\text{gra } A \neq \emptyset$ . Take  $(x_0, u_0) \in \text{gra } A$  and set

$$f: \mathcal{H} \rightarrow [-\infty, +\infty]$$

$$x \mapsto \sup_{\substack{n \in \mathbb{N} \\ n \geq 1}} \sup_{\substack{(x_1, u_1) \in \text{gra } A \\ \vdots \\ (x_n, u_n) \in \text{gra } A}} \left\{ \langle x - x_n | u_n \rangle + \sum_{i=0}^{n-1} \langle x_{i+1} - x_i | u_i \rangle \right\}. \quad (22.12)$$

Since  $\text{gra } A \neq \emptyset$ , we deduce that  $-\infty \notin f(\mathcal{H})$ . On the other hand, it follows from Proposition 9.3 that  $f \in \Gamma(\mathcal{H})$ . The cyclic monotonicity of  $A$  implies

that  $f(x_0) = 0$ . Altogether,  $f \in \Gamma_0(\mathcal{H})$ . Now take  $(x, u) \in \text{gra } A$  and  $\eta \in ]-\infty, f(x)[$ . Then there exist finitely many points  $(x_1, u_1), \dots, (x_n, u_n)$  in  $\text{gra } A$  such that

$$\langle x - x_n \mid u_n \rangle + \sum_{i=0}^{n-1} \langle x_{i+1} - x_i \mid u_i \rangle > \eta. \quad (22.13)$$

Set  $(x_{n+1}, u_{n+1}) = (x, u)$ . Using (22.13), we deduce that, for every  $y \in \mathcal{H}$ ,

$$\begin{aligned} f(y) &\geq \langle y - x_{n+1} \mid u_{n+1} \rangle + \sum_{i=0}^n \langle x_{i+1} - x_i \mid u_i \rangle \\ &= \langle y - x \mid u \rangle + \langle x - x_n \mid u_n \rangle + \sum_{i=0}^{n-1} \langle x_{i+1} - x_i \mid u_i \rangle \\ &> \langle y - x \mid u \rangle + \eta. \end{aligned} \quad (22.14)$$

Letting  $\eta \uparrow f(x)$ , we deduce that  $(\forall y \in \mathcal{H}) \ f(y) \geq f(x) + \langle y - x \mid u \rangle$ , i.e.,  $u \in \partial f(x)$ . Therefore,

$$\text{gra } A \subset \text{gra } \partial f. \quad (22.15)$$

Since  $\partial f$  is cyclically monotone by Proposition 22.11, and since  $A$  is maximally cyclically monotone, we conclude that  $A = \partial f$ .  $\square$

**Proposition 22.15** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $\partial f = \partial g$ . Then there exists  $\gamma \in \mathbb{R}$  such that  $f = g + \gamma$ .*

*Proof.* Consider first the special case in which  $f$  and  $g$  are differentiable on  $\mathcal{H}$ . Fix  $x$  and  $y$  in  $\mathcal{H}$ , and set  $\varphi: \mathbb{R} \rightarrow \mathbb{R}: t \mapsto f(x + t(y - x))$  and  $\psi: \mathbb{R} \rightarrow \mathbb{R}: t \mapsto g(x + t(y - x))$ . We have

$$\begin{aligned} (\forall t \in \mathbb{R}) \quad \varphi'(t) &= \langle y - x \mid \nabla f(x + t(y - x)) \rangle \\ &= \langle y - x \mid \nabla g(x + t(y - x)) \rangle \\ &= \psi'(t). \end{aligned} \quad (22.16)$$

Hence, using Corollary 17.33,  $f(y) - f(x) = \varphi(1) - \varphi(0) = \int_0^1 \varphi'(t) dt = \int_0^1 \psi'(t) dt = \psi(1) - \psi(0) = g(y) - g(x)$ . We conclude that  $f - g$  is a constant. Now we turn our attention to the general case. Set  $q = (1/2)\|\cdot\|^2$  and recall from Example 13.6 that  $q^* = q$ . Using Corollary 16.38(iii), Corollary 16.24, and Proposition 14.1, we obtain the equivalences  $\partial f = \partial g \Leftrightarrow \text{Id} + \partial f = \text{Id} + \partial g \Leftrightarrow \partial(q + f) = \partial(q + g) \Leftrightarrow (\partial(q + f))^{-1} = (\partial(q + g))^{-1} \Leftrightarrow \partial(q + f)^* = \partial(q + g)^* \Leftrightarrow \partial(f^* \boxplus q^*) = \partial(g^* \boxplus q^*) \Leftrightarrow \partial(f^* \boxplus q) = \partial(g^* \boxplus q)$ . The functions  $f^* \boxplus q$  and  $g^* \boxplus q$  are differentiable on  $\mathcal{H}$  and their gradients coincide by Proposition 12.29 and Proposition 17.26(i). Thus, by the already verified special case, there exists  $\gamma \in \mathbb{R}$  such that  $f^* \boxplus q = g^* \boxplus q - \gamma$ . Hence  $f + q = f^{**} + q^* = (f^* \boxplus q)^* = (g^* \boxplus q - \gamma)^* = (g^* \boxplus q)^* + \gamma = g^{**} + q^* + \gamma = g + q + \gamma$ , and we conclude that  $f = g + \gamma$ .  $\square$

## 22.4 Monotone Operators on $\mathbb{R}$

We introduce a binary relation on  $\mathbb{R}^2$  via

$$(\forall \mathbf{x}_1 = (x_1, u_1) \in \mathbb{R}^2)(\forall \mathbf{x}_2 = (x_2, u_2) \in \mathbb{R}^2) \\ \mathbf{x}_1 \preceq \mathbf{x}_2 \Leftrightarrow x_1 \leq x_2 \text{ and } u_1 \leq u_2; \quad (22.17)$$

and we shall write  $\mathbf{x}_1 \prec \mathbf{x}_2$  if  $\mathbf{x}_1 \preceq \mathbf{x}_2$  and  $\mathbf{x}_1 \neq \mathbf{x}_2$ .

We leave the proof of the following results as Exercise 22.7 and Exercise 22.8. (See Section 1.3 for a discussion of order.)

**Proposition 22.16**  $(\mathbb{R}^2, \preceq)$  is directed and partially ordered, but not totally ordered.

**Proposition 22.17** Let  $A: \mathbb{R} \rightarrow 2^{\mathbb{R}}$  be such that  $\text{gra } A \neq \emptyset$ . Then  $A$  is monotone if and only if  $\text{gra } A$  is a chain in  $(\mathbb{R}^2, \preceq)$ .

**Theorem 22.18** Let  $A: \mathbb{R} \rightarrow 2^{\mathbb{R}}$  be such that  $\text{gra } A \neq \emptyset$ . Then  $A$  is monotone if and only if it is cyclically monotone.

*Proof.* It is clear that cyclic monotonicity implies monotonicity. We assume that  $A$  is  $n$ -cyclically monotone, and we shall show that  $A$  is  $(n+1)$ -cyclically monotone. To this end, let  $\mathbf{x}_1 = (x_1, u_1), \dots, \mathbf{x}_{n+1} = (x_{n+1}, u_{n+1})$  be in  $\text{gra } A$ , and set  $x_{n+2} = x_1$ . It suffices to show that

$$\sum_{i=1}^{n+1} (x_{i+1} - x_i)u_i \leq 0. \quad (22.18)$$

The  $n$ -cyclic monotonicity of  $A$  yields  $\sum_{i=1}^{n-1} (x_{i+1} - x_i)u_i + (x_1 - x_n)u_n \leq 0$ , i.e.,

$$\sum_{i=1}^{n-1} (x_{i+1} - x_i)u_i \leq -(x_1 - x_n)u_n. \quad (22.19)$$

Now define  $B: \mathbb{R} \rightarrow 2^{\mathbb{R}}$  by  $\text{gra } B = \{\mathbf{x}_1, \dots, \mathbf{x}_{n+1}\}$ . Since  $A$  is monotone, so is  $B$ . By Proposition 22.17,  $\text{gra } B$  is a chain in  $(\mathbb{R}^2, \preceq)$ . Since  $\text{gra } B$  contains only finitely many elements, it possesses a least element, which—after cyclically relabelling if necessary—we assume to be  $\mathbf{x}_{n+1}$ . After translating  $\text{gra } A$  if necessary, we assume in addition that  $\mathbf{x}_{n+1} = (x_{n+1}, u_{n+1}) = (0, 0)$ , so that  $\text{gra } B \subset \mathbb{R}_+^2$ . Using (22.19), we thus obtain

$$\sum_{i=1}^{n+1} (x_{i+1} - x_i)u_i = \sum_{i=1}^{n-1} (x_{i+1} - x_i)u_i + (0 - x_n)u_n + (x_{n+2} - 0)0 \leq \\ -(x_1 - x_n)u_n - x_n u_n = -x_1 u_n \leq 0. \quad (22.20)$$

This verifies (22.18). Therefore, by induction,  $A$  is cyclically monotone.  $\square$

In view of Example 22.12, the following result fails already in the Euclidean plane.

**Corollary 22.19** *Let  $A: \mathbb{R} \rightarrow 2^{\mathbb{R}}$  be maximally monotone. Then there exists  $f \in \Gamma_0(\mathbb{R})$  such that  $A = \partial f$ .*

*Proof.* Combine Theorem 22.18 and Theorem 22.14. □

## Exercises

**Exercise 22.1** Provide an example of a maximally monotone operator that is not paramonotone.

**Exercise 22.2** Provide an example of a maximally monotone operator that is paramonotone, but not strictly monotone.

**Exercise 22.3** Provide an example of a maximally monotone operator that is strictly monotone, but not uniformly monotone.

**Exercise 22.4** Provide an example of a maximally monotone operator that is uniformly monotone, but not strongly monotone.

**Exercise 22.5** Suppose that  $\mathcal{H}$  is finite-dimensional, and let  $A \in \mathcal{B}(\mathcal{H})$ . Show that  $A$  is strictly monotone if and only if it is strongly monotone.

**Exercise 22.6** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . Denote the Hilbert direct sum  $\mathcal{H}^n$  by  $\mathcal{H}$  and the cyclic right-shift operator on  $\mathcal{H}$  by  $\mathbf{R}$  so that  $\mathbf{R}: \mathcal{H} \rightarrow \mathcal{H}: (x_1, x_2, \dots, x_n) \mapsto (x_n, x_1, \dots, x_{n-1})$ . Show that  $A$  is  $n$ -cyclically monotone if and only if for all  $(x_1, u_1) \in \text{gra } A, \dots, (x_n, u_n) \in \text{gra } A$  we have

$$\|\mathbf{x} - \mathbf{u}\| \leq \|\mathbf{x} - \mathbf{R}\mathbf{u}\|, \quad (22.21)$$

where  $\mathbf{x} = (x_1, \dots, x_n)$  and  $\mathbf{u} = (u_1, \dots, u_n)$ ; equivalently,

$$\sum_{i=1}^n \|x_i - u_i\|^2 \leq \sum_{i=1}^n \|x_i - u_{i-1}\|^2, \quad (22.22)$$

where  $u_0 = u_n$ .

**Exercise 22.7** Prove Proposition 22.16.

**Exercise 22.8** Prove Proposition 22.17.

**Exercise 22.9** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . The *Fitzpatrick function of order  $n$*  at  $(x, u) \in \mathcal{H} \times \mathcal{H}$  is defined by

$$\sup \left( \langle x \mid u \rangle + \left( \sum_{i=1}^{n-2} \langle y_{i+1} - y_i \mid v_i \rangle \right) + \langle x - y_{n-1} \mid v_{n-1} \rangle + \langle y_1 - x \mid u \rangle \right), \quad (22.23)$$

where the supremum is taken over  $(y_1, v_1), \dots, (y_{n-1}, v_{n-1})$  in  $\text{gra } A$ , and set  $F_{A,\infty} = \sup_{n \in \{2,3,\dots\}} F_{A,n}$ . Show that  $F_{A,n}: \mathcal{H} \times \mathcal{H} \rightarrow [-\infty, +\infty]$  is lower semicontinuous and convex, and that  $F_{A,n} \geq \langle \cdot \mid \cdot \rangle$  on  $\text{gra } A$ . What is  $F_{A,2}$ ?

**Exercise 22.10** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . Show that  $A$  is  $n$ -cyclically monotone if and only if  $F_{A,n} = \langle \cdot \mid \cdot \rangle$  on  $\text{gra } A$ .

**Exercise 22.11** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ . Show that  $A$  is cyclically monotone if and only if  $F_{A,\infty} = \langle \cdot \mid \cdot \rangle$  on  $\text{gra } A$ .

**Exercise 22.12** Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $F_{\partial f, \infty} = f \oplus f^*$ .

**Exercise 22.13** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . Use Exercise 20.13 and Exercise 22.12 to determine  $F_{N_C, n}$  and  $F_{N_C, \infty}$ .

**Exercise 22.14** Let  $A \in \mathcal{B}(\mathcal{H})$  be such that  $A^* = -A$  and let  $n \in \mathbb{N}$  be such that  $n \geq 2$ . Determine  $F_{A,n}$  and  $F_{A,\infty}$ .



# Chapter 23

## Resolvents of Monotone Operators

Two quite useful single-valued, Lipschitz continuous operators can be associated with a monotone operator, namely its resolvent and its Yosida approximation. This chapter is devoted to the investigation of these operators. It exemplifies the tight interplay between firmly nonexpansive mappings and monotone operators. Indeed, firmly nonexpansive operators with full domain can be identified with maximally monotone operators via resolvents and the Minty parametrization. When specialized to subdifferential operators, resolvents become proximity operators. Numerous calculus rules for resolvents are derived. Finally, we address the problem of finding a zero of a maximally monotone operator, via the proximal-point algorithm and via approximating curves.

### 23.1 Definition and Basic Identities

**Definition 23.1** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and let  $\gamma \in \mathbb{R}_{++}$ . The *resolvent* of  $A$  is

$$J_A = (\text{Id} + A)^{-1} \quad (23.1)$$

and the *Yosida approximation* of  $A$  of index  $\gamma$  is

$$\gamma A = \frac{1}{\gamma}(\text{Id} - J_{\gamma A}). \quad (23.2)$$

The following properties follow at once from the above definition and (1.7).

**Proposition 23.2** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ , let  $\gamma \in \mathbb{R}_{++}$ , let  $x \in \mathcal{H}$ , and let  $p \in \mathcal{H}$ . Then the following hold:

- (i)  $\text{dom } J_{\gamma A} = \text{dom } \gamma A = \text{ran}(\text{Id} + \gamma A)$  and  $\text{ran } J_{\gamma A} = \text{dom } A$ .
- (ii)  $p \in J_{\gamma A} x \Leftrightarrow x \in p + \gamma A p \Leftrightarrow x - p \in \gamma A p \Leftrightarrow (p, \gamma^{-1}(x - p)) \in \text{gra } A$ .
- (iii)  $p \in \gamma A x \Leftrightarrow p \in A(x - \gamma p) \Leftrightarrow (x - \gamma p, p) \in \text{gra } A$ .

**Example 23.3** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $\gamma \in \mathbb{R}_{++}$ . Then Proposition 16.34 yields

$$J_{\gamma\partial f} = \text{Prox}_{\gamma f}. \quad (23.3)$$

In turn, it follows from Proposition 12.29 that the Yosida approximation of the subdifferential of  $f$  is the Fréchet derivative of the Moreau envelope; more precisely,

$$\gamma(\partial f) = \nabla(\gamma f). \quad (23.4)$$

**Example 23.4** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $\gamma \in \mathbb{R}_{++}$ . Setting  $f = \iota_C$  in Example 23.3 and invoking Example 12.25 and Example 16.12 yields

$$J_{N_C} = (\text{Id} + N_C)^{-1} = \text{Prox}_{\iota_C} = P_C \quad \text{and} \quad \gamma N_C = \frac{1}{\gamma}(\text{Id} - P_C). \quad (23.5)$$

**Example 23.5** Let  $\mathbf{H}$  be a real Hilbert space, let  $\mathbf{x}_0 \in \mathbf{H}$ , suppose that  $\mathcal{H} = L^2([0, T]; \mathbf{H})$ , and let  $A$  be the time-derivative operator (see Example 2.9)

$$A: \mathcal{H} \rightarrow 2^{\mathcal{H}}: x \mapsto \begin{cases} \{x'\}, & \text{if } x \in W^{1,2}([0, T]; \mathbf{H}) \text{ and } x(0) = \mathbf{x}_0; \\ \emptyset, & \text{otherwise.} \end{cases} \quad (23.6)$$

Then  $\text{dom } J_A = \mathcal{H}$  and, for every  $x \in \mathcal{H}$ ,

$$J_A x: [0, T] \rightarrow \mathbf{H}: t \mapsto e^{-t}\mathbf{x}_0 + \int_0^t e^{s-t}x(s)ds. \quad (23.7)$$

*Proof.* Let  $x \in \mathcal{H}$  and set  $y: t \mapsto e^{-t}\mathbf{x}_0 + \int_0^t e^{s-t}x(s)ds$ . As shown in Proposition 21.3 and its proof,  $A$  is maximally monotone; hence  $\text{dom } J_A = \mathcal{H}$  by Theorem 21.1, and  $y \in W^{1,2}([0, T]; \mathbf{H})$ ,  $y(0) = \mathbf{x}_0$ , and  $x(t) = y(t) + y'(t)$  a.e. on  $]0, T[$ . Thus,  $x = (\text{Id} + A)y$  and we deduce that  $y \in J_A x$ . Now let  $z \in J_A x$ , i.e.,  $x = z + Az$ . Then, by monotonicity of  $A$ ,  $0 = \langle y - z \mid x - x \rangle = \|y - z\|^2 + \langle y - z \mid Ay - Az \rangle \geq \|y - z\|^2$  and therefore  $z = y$ .  $\square$

**Proposition 23.6** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $\mu \in \mathbb{R}_{++}$ . Then the following hold:

- (i)  $\text{gra } \gamma A \subset \text{gra}(A \circ J_{\gamma A})$ .
- (ii)  $\gamma A = (\gamma \text{Id} + A^{-1})^{-1} = (J_{\gamma^{-1}A^{-1}}) \circ \gamma^{-1} \text{Id}$ .
- (iii)  $\gamma^{\mu} A = \gamma(\mu A)$ .
- (iv)  $J_{\gamma(\mu A)} = \text{Id} + \gamma/(\gamma + \mu)(J_{(\gamma + \mu)A} - \text{Id})$ .

*Proof.* Let  $x$  and  $u$  be in  $\mathcal{H}$ .

(i): We derive from (23.2) and Proposition 23.2(ii) that  $(x, u) \in \text{gra } \gamma A \Rightarrow (\exists p \in J_{\gamma A} x) \ u = \gamma^{-1}(x - p) \in Ap \Rightarrow u \in A(J_{\gamma A} x)$ .

(ii):  $u \in \gamma A x \Leftrightarrow \gamma u \in x - J_{\gamma A} x \Leftrightarrow x - \gamma u \in J_{\gamma A} x \Leftrightarrow x \in x - \gamma u + \gamma A(x - \gamma u) \Leftrightarrow u \in A(x - \gamma u) \Leftrightarrow x \in \gamma u + A^{-1}u \Leftrightarrow u \in (\gamma \text{Id} + A^{-1})^{-1}x$ . Moreover,  $x \in \gamma u + A^{-1}u \Leftrightarrow \gamma^{-1}x \in u + \gamma^{-1}A^{-1}u \Leftrightarrow u \in J_{\gamma^{-1}A^{-1}}(\gamma^{-1}x)$ .



- (iii): Let  $p \in \mathcal{H}$ . Proposition 23.2 yields  $p \in {}^{\gamma+\mu}Ax \Leftrightarrow p \in A(x - (\gamma + \mu)p) = A((x - \gamma p) - \mu p) \Leftrightarrow p \in ({}^{\mu}A)(x - \gamma p) \Leftrightarrow p \in {}^{\gamma}({}^{\mu}A)x$ .
- (iv): This follows from (iii), (23.1), and elementary manipulations.  $\square$

## 23.2 Monotonicity and Firm Nonexpansiveness

In this section, we focus on the close relationship between firmly nonexpansive mappings and monotone operators.

**Proposition 23.7** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $T: D \rightarrow \mathcal{H}$ , and set  $A = T^{-1} - \text{Id}$ . Then the following hold:*

- (i)  $T = J_A$ .
- (ii)  $T$  is firmly nonexpansive if and only if  $A$  is monotone.
- (iii)  $T$  is firmly nonexpansive and  $D = \mathcal{H}$  if and only if  $A$  is maximally monotone.

*Proof.* (i): See (23.1).

(ii): Suppose that  $T$  is firmly nonexpansive, and take  $(x, u)$  and  $(y, v)$  in  $\text{gra } A$ . Then  $x + u \in T^{-1}x$ , i.e.,  $x = T(x + u)$ . Likewise,  $y = T(y + v)$ . Hence, Proposition 4.2(v) yields

$$\begin{aligned} \langle x - y \mid u - v \rangle = \\ \langle T(x + u) - T(y + v) \mid (\text{Id} - T)(x + u) - (\text{Id} - T)(y + v) \rangle \geq 0, \end{aligned} \quad (23.8)$$

which proves the monotonicity of  $A$ . Now assume that  $A$  is monotone, and let  $x$  and  $y$  be in  $D$ . Then  $x - Tx \in A(Tx)$  and  $y - Ty \in A(Ty)$ . Hence, by monotonicity,  $\langle Tx - Ty \mid (x - Tx) - (y - Ty) \rangle \geq 0$ . By Proposition 4.2,  $T$  is firmly nonexpansive.

(iii): It follows from (ii) and Theorem 21.1 that  $A$  is maximally monotone if and only if  $\mathcal{H} = \text{ran}(\text{Id} + A) = \text{ran } T^{-1} = \text{dom } T = D$ .  $\square$

**Corollary 23.8** *Let  $T: \mathcal{H} \rightarrow \mathcal{H}$ . Then  $T$  is firmly nonexpansive if and only if it is the resolvent of a maximally monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ .*

**Proposition 23.9** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be such that  $\text{dom } A \neq \emptyset$ , set  $D = \text{ran}(\text{Id} + A)$ , and set  $T = J_A|_D$ . Then the following hold:*

- (i)  $A = T^{-1} - \text{Id}$ .
- (ii)  $A$  is monotone if and only if  $T$  is firmly nonexpansive.
- (iii)  $A$  is maximally monotone if and only if  $T$  is firmly nonexpansive and  $D = \mathcal{H}$ .

*Proof.* (i): Clear.

(ii): Assume that  $A$  is monotone, and take  $(x, u)$  and  $(y, v)$  in  $\text{gra } J_A$ . Then  $x - u \in Au$ . Likewise,  $y - v \in Av$ . Hence, by monotonicity,  $\langle u - v \mid (x - u) - (y - v) \rangle \geq 0$ , i.e.,

$$\langle x - y \mid u - v \rangle \geq \|u - v\|^2. \quad (23.9)$$

In particular, for  $x = y$ , we obtain  $u = v$ . Therefore  $T$  is single-valued and we rewrite (23.9) as  $\langle x - y \mid Tx - Ty \rangle \geq \|Tx - Ty\|^2$ . In view of Proposition 4.2(iv),  $T$  is firmly nonexpansive. The reverse statement follows from Proposition 23.7(ii).

(iii): Combine (i), (ii), and Proposition 23.7(iii).  $\square$

Further connections between monotonicity and nonexpansiveness are listed next.

**Corollary 23.10** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then the following hold:*

- (i)  $J_{\gamma A}: \mathcal{H} \rightarrow \mathcal{H}$  and  $\text{Id} - J_{\gamma A}: \mathcal{H} \rightarrow \mathcal{H}$  are firmly nonexpansive and maximally monotone.
- (ii) The reflected resolvent

$$R_{\gamma A}: \mathcal{H} \rightarrow \mathcal{H}: x \mapsto 2J_{\gamma A}x - x \quad (23.10)$$

*is nonexpansive.*

- (iii)  $\gamma A: \mathcal{H} \rightarrow \mathcal{H}$  is  $\gamma$ -cocoercive.
- (iv)  $\gamma A$  is maximally monotone.
- (v)  $\gamma A: \mathcal{H} \rightarrow \mathcal{H}$  is  $\gamma^{-1}$ -Lipschitz continuous.

*Proof.* (i): See Corollary 23.8 and Proposition 4.2 for firm nonexpansiveness, and Example 20.27 for maximal monotonicity.

(i) $\Rightarrow$ (ii): See Proposition 4.2.

(i) $\Rightarrow$ (iii): See (23.2).

(iii) $\Rightarrow$ (iv): See Example 20.28.

(iii) $\Rightarrow$ (v): Cauchy-Schwarz.  $\square$

**Proposition 23.11** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone and let  $\beta \in \mathbb{R}_{++}$ . Then  $A$  is strongly monotone with constant  $\beta$  if and only if  $J_A$  is  $(\beta + 1)$ -cocoercive, in which case  $J_A$  is Lipschitz continuous with constant  $1/(\beta + 1) \in ]0, 1[$ .*

*Proof.* Let  $x, y, u$ , and  $v$  be in  $\mathcal{H}$ . First, suppose that  $A$  is  $\beta$ -strongly monotone. Then, using Proposition 23.2(ii),  $(u, v) = (J_A x, J_A y) \Leftrightarrow (x - u, y - v) \in Au \times Av \Rightarrow \langle (x - u) - (y - v) \mid u - v \rangle \geq \beta \|u - v\|^2 \Leftrightarrow \langle x - y \mid u - v \rangle \geq (\beta + 1) \|u - v\|^2$ . This shows that  $J_A$  is  $(\beta + 1)$ -cocoercive. Conversely, suppose that  $J_A$  is  $(\beta + 1)$ -cocoercive. Then  $(u, v) \in Ax \times Ay \Leftrightarrow ((u + x) - x, (v + y) - y) \in Ax \times Ay \Leftrightarrow (x, y) = (J_A(u + x), J_A(v + y)) \Rightarrow \langle x - y \mid (u + x) - (v + y) \rangle \geq (\beta + 1) \|x - y\|^2 \Leftrightarrow \langle x - y \mid u - v \rangle \geq \beta \|x - y\|^2$ . Thus,  $A$  is  $\beta$ -strongly monotone. The last assertion follows from Cauchy-Schwarz.  $\square$

**Proposition 23.12** *Let  $\beta \in \mathbb{R}_{++}$  and let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally  $\beta$ -cohyppomonotone in the sense that  $A^{-1} + \beta \text{Id}$  is maximally monotone. Let  $\gamma \in ]\beta, +\infty[$  and set  $\lambda = 1 - \beta/\gamma$ . Then  $\text{Id} + \lambda(J_{\gamma A} - \text{Id}): \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive.*

*Proof.* Set  $B = {}^{\beta}A$ . In view of Proposition 23.6(ii) and Proposition 20.22,  $B$  is maximally monotone. Hence, by Corollary 23.10(i),  $J_{(\gamma-\beta)B}: \mathcal{H} \rightarrow \mathcal{H}$  is firmly nonexpansive. The result therefore follows from Proposition 23.6(iv), which provides  $\text{Id} + \lambda(J_{\gamma A} - \text{Id}) = J_{(\gamma-\beta)B}$ .  $\square$

The last two results of this section concern the problem of extending a (firmly) nonexpansive operator defined on  $D \subset \mathcal{H}$  to a (firmly) nonexpansive operator defined on the whole space  $\mathcal{H}$ .

**Theorem 23.13** *Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be firmly nonexpansive. Then there exists a firmly nonexpansive operator  $\tilde{T}: \mathcal{H} \rightarrow \mathcal{H}$  such that  $\tilde{T}|_D = T$  and  $\text{ran } \tilde{T} \subset \overline{\text{conv}} \text{ran } T$ .*

*Proof.* Proposition 23.7 asserts that there exists a monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\text{ran}(\text{Id} + A) = D$  and  $T = J_A$ . However, by Theorem 21.8,  $A$  admits a maximally monotone extension  $\tilde{A}: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\text{dom } \tilde{A} \subset \overline{\text{conv}} \text{dom } A$ . Now set  $\tilde{T} = J_{\tilde{A}}$ . Then it follows from Corollary 23.10(i) that  $\tilde{T}$  is firmly nonexpansive with  $\text{dom } \tilde{T} = \mathcal{H}$ . On the other hand,  $\text{ran } \tilde{T} = \text{dom}(\text{Id} + \tilde{A}) = \text{dom } \tilde{A} \subset \overline{\text{conv}} \text{dom } A = \overline{\text{conv}} \text{dom}(\text{Id} + A) = \overline{\text{conv}} \text{ran } J_A = \overline{\text{conv}} \text{ran } T$ . Finally, let  $x \in D$ . Then  $Tx = J_A x \Rightarrow x - Tx \in A(Tx) \subset \tilde{A}(Tx) \Rightarrow Tx = J_{\tilde{A}} x = \tilde{T}x$ . Thus,  $\tilde{T}|_D = T$ .  $\square$

**Corollary 23.14 (Kirszbraun–Valentine)** *Let  $D$  be a nonempty subset of  $\mathcal{H}$  and let  $T: D \rightarrow \mathcal{H}$  be a nonexpansive operator. Then there exists a nonexpansive operator  $\tilde{T}: \mathcal{H} \rightarrow \mathcal{H}$  such that  $\tilde{T}|_D = T$  and  $\text{ran } \tilde{T} \subset \overline{\text{conv}} \text{ran } T$ .*

*Proof.* Set  $R = (\text{Id} + T)/2$ . Then  $R: D \rightarrow \mathcal{H}$  is firmly nonexpansive by Proposition 4.2 and, by Theorem 23.13, it admits a firmly nonexpansive extension  $\tilde{R}: \mathcal{H} \rightarrow \mathcal{H}$ . Hence,  $\tilde{T} = P_C \circ (2\tilde{R} - \text{Id})$ , where  $C = \overline{\text{conv}} \text{ran } T$ , has the required properties.  $\square$

## 23.3 Resolvent Calculus

In this section we establish formulas for computing resolvents of transformations of maximally monotone operators.

**Proposition 23.15** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then the following hold:*

- (i) *Let  $\alpha \in \mathbb{R}_+$  and set  $B = A + \alpha \text{Id}$ . Then  $J_B = J_{(1+\alpha)^{-1}A}((1+\alpha)^{-1} \text{Id})$ .*
- (ii) *Let  $z \in \mathcal{H}$  and set  $B = z + A$ . Then  $J_B = \tau_z J_A$ .*

(iii) Let  $z \in \mathcal{H}$  and set  $B = \tau_z A$ . Then  $J_B = z + \tau_z J_A$ .

*Proof.* Let  $x$  and  $p$  be in  $\mathcal{H}$ . Then we derive from Proposition 23.2(iii) the following equivalences:

(i):  $p = J_B x \Leftrightarrow x - p \in Ap + \alpha p \Leftrightarrow x - (1 + \alpha)p \in Ap \Leftrightarrow (1 + \alpha)^{-1}x - p \in (1 + \alpha)^{-1}Ap \Leftrightarrow p = J_{(1+\alpha)^{-1}A}((1 + \alpha)^{-1}x)$ .

(ii):  $p = J_B x \Leftrightarrow x - p \in z + Ap \Leftrightarrow (x - z) - p \in Ap \Leftrightarrow p = J_A(x - z)$ .

(iii):  $p = J_B x \Leftrightarrow x - p \in A(p - z) \Leftrightarrow (x - z) - (p - z) \in A(p - z) \Leftrightarrow p - z = J_A(x - z)$ .  $\square$

**Proposition 23.16** Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces, set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ , and, for every  $i \in I$ , let  $A_i: \mathcal{H}_i \rightarrow 2^{\mathcal{H}_i}$  be maximally monotone. Set  $\mathbf{A} = \bigtimes_{i \in I} A_i$ . Then  $\mathbf{A}$  is maximally monotone and

$$J_{\mathbf{A}} = \bigtimes_{i \in I} J_{A_i}. \quad (23.11)$$

*Proof.* It is clear that  $\mathbf{A}$  is monotone. On the other hand, it follows from Theorem 21.1 that  $(\forall i \in I) \operatorname{ran}(\operatorname{Id} + A_i) = \mathcal{H}_i$ . Hence,  $\operatorname{ran}(\operatorname{Id} + \mathbf{A}) = \bigtimes_{i \in I} \operatorname{ran}(\operatorname{Id} + A_i) = \mathcal{H}$ , and  $\mathbf{A}$  is therefore maximally monotone. Now let  $\mathbf{x} = (x_i)_{i \in I}$  be an arbitrary point in  $\mathcal{H}$ , set  $\mathbf{p} = J_{\mathbf{A}} \mathbf{x}$ , and let  $(p_i)_{i \in I} \in \mathcal{H}$ . We derive from Proposition 23.2(ii) that

$$\begin{aligned} (\forall i \in I) \quad p_i = J_{A_i} x_i &\Leftrightarrow (\forall i \in I) \quad x_i - p_i \in A_i p_i \\ &\Leftrightarrow \mathbf{x} - (p_i)_{i \in I} \in (A_i p_i)_{i \in I} = \mathbf{A}(p_i)_{i \in I} \\ &\Leftrightarrow \mathbf{p} = (p_i)_{i \in I}, \end{aligned} \quad (23.12)$$

which establishes (23.11).  $\square$

Given a maximally monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  and  $\gamma \in \mathbb{R}_{++}$ , no simple formula is known to express  $J_{\gamma A}$  in terms of  $J_A$ . However, a simple formula relates the graphs of these two resolvents (the proof is left as Exercise 23.9).

**Proposition 23.17** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: (x, u) \mapsto (\gamma x + (1 - \gamma)u, u)$ . Then  $\operatorname{gra} J_{\gamma A} = L(\operatorname{gra} J_A)$ .

The next result relates the resolvent of an operator to that of its inverse.

**Proposition 23.18** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then

$$\operatorname{Id} = J_{\gamma A} + \gamma J_{\gamma^{-1}A^{-1}} \circ \gamma^{-1} \operatorname{Id}. \quad (23.13)$$

In particular,

$$J_{A^{-1}} = \operatorname{Id} - J_A. \quad (23.14)$$

*Proof.* Since  $J_{\gamma A}$  and  $J_{A^{-1}/\gamma}$  are single-valued, (23.13) follows from (23.2) and Proposition 23.6(ii).  $\square$

**Remark 23.19** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x \in \mathcal{H}$ . In view of Example 23.3 and Corollary 16.24, setting  $A = \partial f$  in (23.13) yields Theorem 14.3(ii), namely  $x = \text{Prox}_{\gamma f} x + \gamma \text{Prox}_{\gamma^{-1}f^*}(\gamma^{-1}x)$ .

The next result characterizes cocoercive operators.

**Proposition 23.20** *Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  and let  $\gamma \in \mathbb{R}_{++}$ . Then  $T$  is  $\gamma$ -cocoercive if and only if it is the Yosida approximation of index  $\gamma$  of a maximally monotone operator from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ .*

*Proof.* The Yosida approximation of index  $\gamma$  of a maximally monotone operator is  $\gamma$ -cocoercive by Corollary 23.10(iii). Conversely, suppose that  $T$  is  $\gamma$ -cocoercive. Then it follows from Corollary 23.8 and (23.14) that there exists a maximally monotone operator  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  such that  $\gamma T = J_B = \text{Id} - J_{B^{-1}}$ . Now set  $A = \gamma^{-1}B^{-1}$ . Then Proposition 20.22 asserts that  $A$  is maximally monotone, and (23.2) yields  $T = {}^{\gamma}A$ .  $\square$

**Proposition 23.21** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and let  $x, y$ , and  $v$  be in  $\mathcal{H}$ . Then*

$$(y, v) = (J_{\gamma A}x, {}^{\gamma}Ax) \quad \Leftrightarrow \quad \begin{cases} (y, v) \in \text{gra } A, \\ x = y + \gamma v. \end{cases} \quad (23.15)$$

*Proof.* Using Proposition 23.2(ii), (23.1), and (23.2), we obtain

$$\begin{cases} y = J_{\gamma A}x, \\ v = {}^{\gamma}Ax, \end{cases} \quad \Leftrightarrow \quad \begin{cases} (y, x - y) \in \text{gra } \gamma A, \\ v = (x - y)/\gamma, \end{cases} \quad \Leftrightarrow \quad \begin{cases} (y, v) \in \text{gra } A, \\ x = y + \gamma v, \end{cases} \quad (23.16)$$

as required.  $\square$

**Remark 23.22** Here are some consequences of Proposition 23.21.

- (i) Set  $R: \mathcal{H} \rightarrow \text{gra } A: x \mapsto (J_{\gamma A}x, {}^{\gamma}Ax)$ . Then  $R$  is a bijection and, more precisely, a Lipschitz homeomorphism from  $\mathcal{H}$  to  $\text{gra } A$ , viewed as a subset of  $\mathcal{H} \oplus \mathcal{H}$ . Indeed, for every  $x$  and  $y$  in  $\mathcal{H}$ , Corollary 23.10(i)&(v) yields  $\|Rx - Ry\|^2 = \|J_{\gamma A}x - J_{\gamma A}y\|^2 + \|{}^{\gamma}Ax - {}^{\gamma}Ay\|^2 \leq \|x - y\|^2 + \gamma^{-2}\|x - y\|^2$ . Hence,  $R$  is  $\sqrt{1 + 1/\gamma^2}$ -Lipschitz continuous. Conversely, if  $(x, u)$  and  $(y, v)$  are in  $\text{gra } A$  then, by Cauchy–Schwarz,

$$\begin{aligned} \|R^{-1}(x, u) - R^{-1}(y, v)\|^2 &= \|(x - y) + \gamma(u - v)\|^2 \\ &\leq (\|x - y\| + \gamma\|u - v\|)^2 \\ &\leq (1 + \gamma^2)(\|x - y\|^2 + \|u - v\|^2) \\ &= (1 + \gamma^2)\|(x, u) - (y, v)\|^2. \end{aligned} \quad (23.17)$$

Therefore,  $R^{-1}$  is  $\sqrt{1 + \gamma^2}$ -Lipschitz continuous.

- (ii) Setting  $\gamma = 1$  in (i), we obtain via (23.2) and (23.14) the *Minty parametrization*

$$\mathcal{H} \rightarrow \text{gra } A: x \mapsto (J_A x, J_{A^{-1}} x) = (J_A x, x - J_A x) \quad (23.18)$$

of  $\text{gra } A$ .

We now investigate resolvents of composite operators.

**Proposition 23.23** *Let  $\mathcal{K}$  be a real Hilbert space, suppose that  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  is such that  $LL^*$  is invertible, let  $A: \mathcal{K} \rightarrow 2^{\mathcal{K}}$  be maximally monotone, and set  $B = L^*AL$ . Then the following hold:*

- (i)  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is maximally monotone.
- (ii)  $J_B = \text{Id} - L^* \circ (LL^* + A^{-1})^{-1} \circ L$ .
- (iii) Suppose that  $LL^* = \mu \text{Id}$  for some  $\mu \in \mathbb{R}_{++}$ . Then  $J_B = \text{Id} - L^* \circ {}^{\mu}A \circ L$ .

*Proof.* (i): Since  $\mathcal{K} = L(L^*(\mathcal{K})) \subset L(\mathcal{H}) = \text{ran } L$ , we have  $\text{ran } L = \mathcal{K}$  and therefore  $\text{cone}(\text{ran } L - \text{dom } A) = \overline{\text{span}}(\text{ran } L - \text{dom } A)$ . Thus, as will be seen in Theorem 24.5,  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is maximally monotone.

(ii): It follows from (i) and Corollary 23.10(i) that  $J_B$  is single-valued with domain  $\mathcal{H}$ . Now set  $T = (LL^* + A^{-1})^{-1}$ . Since  $LL^*$  is invertible, it is strictly monotone by Fact 2.18(v), and so is  $LL^* + A^{-1}$ . As a result,  $T$  is single-valued on  $\text{dom } T = \text{ran}(LL^* + A^{-1})$ . On the other hand, by Example 24.13,  $LL^*$  is  $3^*$  monotone. Hence, it follows from Corollary 24.4(i) and Corollary 24.22(ii) that  $\text{dom } T = \mathcal{H}$ . Now let  $x \in \mathcal{H}$ , note that  $v = T(Lx) = (LL^* + A^{-1})^{-1}(Lx)$  is well defined, and set  $p = x - L^*v$ . Then  $Lx \in LL^*v + A^{-1}v$  and hence  $Lp = L(x - L^*v) \in A^{-1}v$ . In turn,  $v \in A(Lp)$  and, therefore,  $x - p = L^*v \in L^*(A(Lp)) = Bp$ . Thus,  $p = J_B x$ , as claimed.

(iii): This follows from (ii) and Proposition 23.6(ii).  $\square$

**Corollary 23.24** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\rho \in \mathbb{R} \setminus \{0\}$ , and set  $B = \rho A(\rho \cdot)$ . Then  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is maximally monotone and*

$$J_B = \rho^{-1} J_{\rho^2 A}(\rho \cdot). \quad (23.19)$$

*Proof.* Apply Proposition 23.23(iii) to  $\mathcal{K} = \mathcal{H}$  and  $L = \rho \text{Id}$ .  $\square$

**Corollary 23.25** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $L \in \mathcal{B}(\mathcal{H})$  be an invertible operator such that  $L^{-1} = L^*$ , and set  $B = L^* \circ A \circ L$ . Then  $B$  is maximally monotone and  $J_B = L^* \circ J_A \circ L$ .*

*Proof.* Apply Proposition 23.23(iii) to  $\mathcal{K} = \mathcal{H}$  and  $\mu = 1$ .  $\square$

**Corollary 23.26** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and set  $B = -A^{\vee}$ . Then  $B$  is maximally monotone and  $J_B = -(J_A)^{\vee}$ .*

*Proof.* Apply Corollary 23.25 to  $L = -\text{Id}$ .  $\square$

**Proposition 23.27** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and set  $B = \gamma^{-1}\text{Id} - \gamma A = \gamma^{-1}J_{\gamma A}$ . Then  $B: \mathcal{H} \rightarrow \mathcal{H}$  is maximally monotone and*

$$J_B = \text{Id} - \frac{1}{\gamma} J_{\frac{\gamma^2}{\gamma+1}A} \circ \left( \frac{\gamma}{\gamma+1} \text{Id} \right). \quad (23.20)$$

*Proof.* It follows from Corollary 23.10(i) that  $B = \gamma^{-1}J_{\gamma A}$  is maximally monotone and single-valued with domain  $\mathcal{H}$ . Hence, for every  $x$  and  $p$  in  $\mathcal{H}$ , Proposition 23.2(ii) yields

$$\begin{aligned} p = J_B x &\Leftrightarrow \gamma(x - p) = \gamma B p = J_{\gamma A} p \\ &\Leftrightarrow p - \gamma(x - p) \in \gamma A(\gamma(x - p)) \\ &\Leftrightarrow \frac{\gamma x}{\gamma + 1} - \gamma(x - p) \in \frac{\gamma^2}{\gamma + 1} A(\gamma(x - p)) \\ &\Leftrightarrow \gamma(x - p) = J_{\frac{\gamma^2}{\gamma+1}A} \left( \frac{\gamma x}{\gamma + 1} \right) \\ &\Leftrightarrow p = x - \frac{1}{\gamma} J_{\frac{\gamma^2}{\gamma+1}A} \left( \frac{\gamma x}{\gamma + 1} \right), \end{aligned} \quad (23.21)$$

which gives (23.20).  $\square$

**Proposition 23.28** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , let  $\lambda \in \mathbb{R}_{++}$ , and let  $x \in \mathcal{H}$ . Then the following hold:*

- (i)  $J_{\gamma A} x = J_{\lambda \gamma A}(\lambda x + (1 - \lambda)J_{\gamma A} x)$ .
- (ii) Suppose that  $\lambda \leq 1$ . Then  $\|J_{\lambda \gamma A} x - x\| \leq (2 - \lambda)\|J_{\gamma A} x - x\|$ .
- (iii)  $\|J_{\gamma A} x - J_{\lambda \gamma A} x\| \leq |1 - \lambda| \|J_{\gamma A} x - x\|$ .

*Proof.* Set  $\mu = \lambda\gamma$ .

(i): We have  $x \in (\text{Id} + \gamma A)(\text{Id} + \gamma A)^{-1}x = (\text{Id} + \gamma A)J_{\gamma A} x$ . Hence  $x - J_{\gamma A} x \in \gamma A(J_{\gamma A} x)$  and therefore  $\lambda(x - J_{\gamma A} x) \in \mu A(J_{\gamma A} x)$ . In turn,  $\lambda x + (1 - \lambda)J_{\gamma A} x \in J_{\gamma A} x + \mu A(J_{\gamma A} x) = (\text{Id} + \mu A)(J_{\gamma A} x)$  and therefore  $J_{\gamma A} x = J_{\mu A}(\lambda x + (1 - \lambda)J_{\gamma A} x)$ .

(ii): By (i) and Corollary 23.10(i),

$$\begin{aligned} \|J_{\mu A} x - x\| &\leq \|J_{\mu A} x - J_{\gamma A} x\| + \|J_{\gamma A} x - x\| \\ &= \|J_{\mu A} x - J_{\mu A}(\lambda x + (1 - \lambda)J_{\gamma A} x)\| + \|J_{\gamma A} x - x\| \\ &\leq \|x - \lambda x - (1 - \lambda)J_{\gamma A} x\| + \|J_{\gamma A} x - x\| \\ &\leq (2 - \lambda)\|J_{\gamma A} x - x\|. \end{aligned} \quad (23.22)$$

(iii): By (i) and Corollary 23.10(i), we have

$$\begin{aligned} \|J_{\gamma A} x - J_{\mu A} x\| &= \|J_{\mu A}(\lambda x + (1 - \lambda)J_{\gamma A} x) - J_{\mu A} x\| \\ &\leq \|\lambda x + (1 - \lambda)J_{\gamma A} x - x\| \\ &= |1 - \lambda| \|J_{\gamma A} x - x\|, \end{aligned} \quad (23.23)$$

which concludes the proof.  $\square$

Using Example 23.3, we derive easily properties of proximity operators from those of resolvents.

**Proposition 23.29** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $x \in \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Then the following hold:*

- (i) *Set  $g = f + (\alpha/2)\|\cdot\|^2 + \langle \cdot | u \rangle + \beta$ , where  $u \in \mathcal{H}$ ,  $\alpha \in \mathbb{R}_+$ , and  $\beta \in \mathbb{R}$ . Then  $\text{Prox}_g x = \text{Prox}_{(\alpha+1)^{-1}f}((\alpha+1)^{-1}(x-u))$ .*
- (ii) *Set  $g = \tau_z f$ , where  $z \in \mathcal{H}$ . Then  $\text{Prox}_g x = z + \text{Prox}_f(x-z)$ .*
- (iii) *Set  $g = f \circ L$ , where  $L \in \mathcal{B}(\mathcal{H})$  is an invertible operator such that  $L^{-1} = L^*$ . Then  $\text{Prox}_g x = L^*(\text{Prox}_f(Lx))$ .*
- (iv) *Set  $g = f(\rho \cdot)$ , where  $\rho \in \mathbb{R} \setminus \{0\}$ . Then  $\text{Prox}_g x = \rho^{-1} \text{Prox}_{\rho^2 f}(\rho x)$ .*
- (v) *Set  $g = f^\vee$ . Then  $\text{Prox}_g x = -\text{Prox}_f(-x)$ .*
- (vi) *Set  $g = \gamma f$ . Then  $\text{Prox}_g x = x + (\gamma+1)^{-1}(\text{Prox}_{(\gamma+1)f} x - x)$ .*
- (vii) *Set  $g = (2\gamma)^{-1}\|\cdot\|^2 - \gamma f$ . Then*

$$\text{Prox}_g x = x - \frac{1}{\gamma} \text{Prox}_{\frac{\gamma^2}{\gamma+1}f} \left( \frac{\gamma}{\gamma+1} x \right).$$

- (viii) *Set  $g = \gamma f^*$ . Then  $\text{Prox}_g x = x - \gamma \text{Prox}_{\gamma^{-1}f}(\gamma^{-1}x)$ .*

*Proof.* These properties are specializations to  $A = \partial f$  and  $B = \partial g$  of some of the above results.

(i): Proposition 23.15(i)&(ii) (note that Corollary 16.38(iii) and Proposition 17.26(i) imply that  $\partial g = \partial f + \alpha \text{Id} + u$ ).

(ii): Proposition 23.15(iii).

(iii): Corollary 23.25 (note that, since  $\text{ran } L = \mathcal{H}$ , Corollary 16.42(i) yields  $\partial g = L^* \circ (\partial f) \circ L$ ).

(iv): Corollary 23.24.

(v): Corollary 23.26.

(vi): Proposition 23.6(iv) and (23.4).

(vii): It follows from Example 13.12 that  $\varphi = (\gamma/2)\|\cdot\|^2 - \gamma f(\gamma \cdot) \in \Gamma_0(\mathcal{H})$  and therefore that  $g = \varphi(\gamma^{-1} \cdot) \in \Gamma_0(\mathcal{H})$ . Furthermore, (23.4) yields  $\nabla g = \gamma^{-1} \text{Id} - \gamma(\partial f)$ . Hence, the result follows from Proposition 23.27.

(viii): Remark 23.19.  $\square$

**Proposition 23.30** *Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces, set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ , and, for every  $i \in I$ , let  $f_i \in \Gamma_0(\mathcal{H}_i)$  and  $x_i \in \mathcal{H}_i$ . Set  $\mathbf{x} = (x_i)_{i \in I}$  and  $\mathbf{f} = \bigoplus_{i \in I} f_i$ . Then  $\text{Prox}_{\mathbf{f}} \mathbf{x} = (\text{Prox}_{f_i} x_i)_{i \in I}$ .*

*Proof.* It is clear that  $\mathbf{f} \in \Gamma_0(\mathcal{H})$ . The result therefore follows from Proposition 23.16, where  $(\forall i \in I) A_i = \partial f_i$ .  $\square$

**Proposition 23.31** *Let  $(\phi_i)_{i \in I}$  be a totally ordered family of functions in  $\Gamma_0(\mathbb{R})$  such that  $(\forall i \in I) \phi_i \geq \phi_i(0) = 0$ . Suppose that  $\mathcal{H} = \ell^2(I)$ , set  $f = \bigoplus_{i \in I} \phi_i$ , and let  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ . Then the following hold:*



- (i)  $f \in \Gamma_0(\mathcal{H})$ .
- (ii)  $(\forall (\xi_i)_{i \in I} \in \mathcal{H}) \operatorname{Prox}_f(\xi_i)_{i \in I} = (\operatorname{Prox}_{\phi_i} \xi_i)_{i \in I}$ .
- (iii)  $\operatorname{Prox}_f$  is weakly sequentially continuous.

*Proof.* (i): For every  $i \in I$ , define  $\varphi_i: \mathcal{H} \rightarrow ]-\infty, +\infty]: (\xi_j)_{j \in I} \mapsto \phi_i(\xi_i)$ . Then the family  $(\varphi_i)_{i \in I}$  lies in  $\Gamma_0(\mathcal{H})$  and, since  $f(0) = 0$ , it follows from Corollary 9.4(ii) that  $f = \sum_{i \in I} \varphi_i \in \Gamma_0(\mathcal{H})$ .

(ii): Let  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ , and set  $p = (\pi_i)_{i \in I} = (\operatorname{Prox}_{\phi_i} \xi_i)_{i \in I}$ . For every  $i \in I$ , since  $0 \in \operatorname{Argmin} \phi_i$ , it follows from Theorem 16.2 that  $0 - 0 \in \partial \phi_i(0)$  and therefore from (16.30) that  $\operatorname{Prox}_{\phi_i} 0 = 0$ . Hence, we derive from Proposition 12.27 that

$$(\forall i \in I) \quad |\pi_i|^2 = |\operatorname{Prox}_{\phi_i} \xi_i - \operatorname{Prox}_{\phi_i} 0|^2 \leq |\xi_i - 0|^2 = |\xi_i|^2. \quad (23.24)$$

This shows that  $p \in \mathcal{H}$ . Now let  $y = (\eta_i)_{i \in I} \in \mathcal{H}$ . It follows from Proposition 12.26 that

$$(\forall i \in I) \quad (\eta_i - \pi_i)(\xi_i - \pi_i) + \phi_i(\pi_i) \leq \phi_i(\eta_i). \quad (23.25)$$

Summing over  $I$ , we obtain  $\langle y - p \mid x - p \rangle + f(p) \leq f(y)$ . In view of (12.25), we conclude that  $p = \operatorname{Prox}_f x$ .

(iii): Let  $(x_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{H}$  and let  $x \in \mathcal{H}$  be such that  $x_n \rightharpoonup x$ . Then Proposition 2.40 asserts that  $(x_n)_{n \in \mathbb{N}}$  is bounded. On the other hand, arguing as in (23.24), since  $0 \in \operatorname{Argmin} f$ ,  $\operatorname{Prox}_f 0 = 0$  and, by nonexpansiveness of  $\operatorname{Prox}_f$ , we obtain  $(\forall n \in \mathbb{N}) \|\operatorname{Prox}_f x_n\| \leq \|x_n\|$ . Hence,

$$(\operatorname{Prox}_f x_n)_{n \in \mathbb{N}} \text{ is bounded.} \quad (23.26)$$

Now set  $x = (\xi_i)_{i \in I}$  and  $(\forall n \in \mathbb{N}) x_n = (\xi_{i,n})_{i \in I}$ . Denoting the standard unit vectors in  $\mathcal{H}$  by  $(e_i)_{i \in I}$ , we obtain  $(\forall i \in I) \xi_{i,n} = \langle x_n \mid e_i \rangle \rightarrow \langle x \mid e_i \rangle = \xi_i$ . Since, by Proposition 12.27, the operators  $(\operatorname{Prox}_{\phi_i})_{i \in I}$  are continuous, (ii) yields  $(\forall i \in I) \langle \operatorname{Prox}_f x_n \mid e_i \rangle = \operatorname{Prox}_{\phi_i} \xi_{i,n} \rightarrow \operatorname{Prox}_{\phi_i} \xi_i = \langle \operatorname{Prox}_f x \mid e_i \rangle$ . It then follows from (23.26) and Proposition 2.40 that  $\operatorname{Prox}_f x_n \rightharpoonup \operatorname{Prox}_f x$ .  $\square$

**Proposition 23.32** *Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $LL^* = \mu \operatorname{Id}$  for some  $\mu \in \mathbb{R}_{++}$ , let  $f \in \Gamma_0(\mathcal{K})$ , and let  $x \in \mathcal{H}$ . Then  $f \circ L \in \Gamma_0(\mathcal{H})$  and  $\operatorname{Prox}_{f \circ L} x = x + \mu^{-1} L^*(\operatorname{Prox}_{\mu f}(Lx) - Lx)$ .*

*Proof.* Since  $\operatorname{ran} L = \mathcal{K}$  by Fact 2.18(v) and Fact 2.19, Corollary 16.42(i) yields  $\partial(f \circ L) = L^* \circ (\partial f) \circ L$ . Hence, the result follows by setting  $A = \partial f$  in Proposition 23.23(iii) and using (23.2).  $\square$

**Corollary 23.33** *Suppose that  $u$  is a nonzero vector in  $\mathcal{H}$ , let  $\phi \in \Gamma_0(\mathbb{R})$ , set  $g = \phi(\langle \cdot \mid u \rangle)$ , and let  $x \in \mathcal{H}$ . Then*

$$\operatorname{Prox}_g x = x + \frac{\operatorname{Prox}_{\|u\|^2 \phi} \langle x \mid u \rangle - \langle x \mid u \rangle}{\|u\|^2} u. \quad (23.27)$$

*Proof.* Apply Proposition 23.32 to  $\mathcal{K} = \mathbb{R}$ ,  $f = \phi$ , and  $L = \langle \cdot | u \rangle$ .  $\square$

**Proposition 23.34** *Suppose that  $(e_k)_{k \in \mathbb{N}}$  is an orthonormal basis of  $\mathcal{H}$ , let  $(\phi_k)_{k \in \mathbb{N}}$  be a family of functions in  $\Gamma_0(\mathbb{R})$  such that  $(\forall k \in \mathbb{N}) \phi_k \geq \phi_k(0) = 0$ , and set*

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \sum_{k \in \mathbb{N}} \phi_k(\langle x | e_k \rangle). \quad (23.28)$$

*Then  $f \in \Gamma_0(\mathcal{H})$  and*

$$(\forall x \in \mathcal{H}) \quad \text{Prox}_f x = \sum_{k \in \mathbb{N}} (\text{Prox}_{\phi_k} \langle x | e_k \rangle) e_k. \quad (23.29)$$

*Proof.* Set  $L: \mathcal{H} \rightarrow \ell^2(\mathbb{N}): x \mapsto (\langle x | e_k \rangle)_{k \in \mathbb{N}}$ . Then  $L$  is linear, bounded, and invertible with  $L^{-1} = L^*: \ell^2(\mathbb{N}) \rightarrow \mathcal{H}: (\xi_k)_{k \in \mathbb{N}} \mapsto \sum_{k \in \mathbb{N}} \xi_k e_k$ . Now set  $\varphi: \ell^2(\mathbb{N}) \rightarrow ]-\infty, +\infty]: (\xi_k)_{k \in \mathbb{N}} \mapsto \sum_{k \in \mathbb{N}} \phi_k(\xi_k)$ . Then  $f = \varphi \circ L$  and we derive from Proposition 23.32 that  $\text{Prox}_{f \circ L} = L^* \circ \text{Prox}_\varphi \circ L$ . In turn, using Proposition 23.31(i)&(ii), we obtain the announced results.  $\square$

## 23.4 Zeros of Monotone Operators

In this section we study the properties of the zeros of monotone operators (see (1.8)) and describe a basic algorithm to construct them iteratively.

**Proposition 23.35** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be strictly monotone. Then  $\text{zer } A$  is at most a singleton.*

*Proof.* Suppose that  $x$  and  $y$  are distinct points in  $\text{zer } A$ . Then  $0 \in Ax$ ,  $0 \in Ay$ , and (22.2) yields  $0 = \langle x - y | 0 - 0 \rangle > 0$ , which is impossible.  $\square$

**Proposition 23.36** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and suppose that one of the following holds:*

- (i)  $A^{-1}$  is locally bounded everywhere.
- (ii)  $\lim_{\|x\| \rightarrow +\infty} \inf \|Ax\| = +\infty$ .
- (iii)  $\text{dom } A$  is bounded.

*Then  $\text{zer } A \neq \emptyset$ .*

*Proof.* The conclusion follows from (i), (ii), and (iii) via Corollary 21.19, Corollary 21.20, and Corollary 21.21, respectively.  $\square$

**Corollary 23.37** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and suppose that one of the following holds:*

- (i)  $A$  is uniformly monotone with a supercoercive modulus.
- (ii)  $A$  is strongly monotone.

Then  $\text{zer } A$  is a singleton.

*Proof.* (i): In view of Proposition 22.8,  $\text{zer } A \neq \emptyset$ . Furthermore, since  $A$  is strictly monotone, we derive from Proposition 23.35 that  $\text{zer } A$  is a singleton.

(ii) $\Rightarrow$ (i): Definition 22.1.  $\square$

**Proposition 23.38** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\text{Fix } J_{\gamma A} = \text{zer } A = \text{zer } {}^{\gamma}A$ .*

*Proof.* Let  $x \in \mathcal{H}$ . Then Proposition 23.2(ii) yields  $0 \in Ax \Leftrightarrow x - x \in \gamma Ax \Leftrightarrow x = J_{\gamma A} x \Leftrightarrow {}^{\gamma}Ax = 0$ .  $\square$

**Proposition 23.39** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Then  $\text{zer } A$  is closed and convex.*

*Proof.* This is a consequence of Proposition 20.31 since  $\text{zer } A = A^{-1}0$  and  $A^{-1}$  is maximally monotone.  $\square$

Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\text{zer } A = \text{zer}(\gamma A)$  and, in view of Proposition 23.38, a zero of  $A$  can be approximated iteratively by suitable resolvent iterations. Such algorithms are known as *proximal-point algorithms*.

**Example 23.40** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = J_{\gamma A} x_n. \quad (23.30)$$

Then the following hold:

- (i) Suppose that  $\text{zer } A \neq \emptyset$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer } A$ .
- (ii) Suppose that  $A$  is strongly monotone with constant  $\beta \in \mathbb{R}_{++}$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly (actually linearly) to the unique point in  $\text{zer } A$ .

*Proof.* Proposition 23.38 yields  $\text{Fix } J_{\gamma A} = \text{zer}(\gamma A) = \text{zer } A$ .

(i): In view of Corollary 23.10(i), the result is an application of Example 5.17 with  $T = J_{\gamma A}$ .

(ii): Proposition 23.11 asserts that  $J_{\gamma A}$  is Lipschitz continuous with constant  $1/(\beta\gamma + 1) \in ]0, 1[$ . Therefore, the result follows by setting  $T = J_{\gamma A}$  in Theorem 1.48(i)&(iii).  $\square$

A finer proximal-point algorithm is described in the following theorem.

**Theorem 23.41 (proximal-point algorithm)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator such that  $\text{zer } A \neq \emptyset$ , let  $(\gamma_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  such that  $\sum_{n \in \mathbb{N}} \gamma_n^2 = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = J_{\gamma_n A} x_n. \quad (23.31)$$

Then the following hold:

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer } A$ .
- (ii) Suppose that  $A$  is uniformly monotone on every bounded subset of  $\mathcal{H}$  (see Remark 22.2). Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique point in  $\text{zer } A$ .

*Proof.* Set  $(\forall n \in \mathbb{N}) \ u_n = (x_n - x_{n+1})/\gamma_n$ . Then  $(\forall n \in \mathbb{N}) \ u_n \in Ax_{n+1}$  and  $x_{n+1} - x_{n+2} = \gamma_{n+1}u_{n+1}$ . Hence, by monotonicity and Cauchy–Schwarz,

$$\begin{aligned}
 (\forall n \in \mathbb{N}) \quad 0 &\leq \langle x_{n+1} - x_{n+2} \mid u_n - u_{n+1} \rangle / \gamma_{n+1} \\
 &= \langle u_{n+1} \mid u_n - u_{n+1} \rangle \\
 &= \langle u_{n+1} \mid u_n \rangle - \|u_{n+1}\|^2 \\
 &\leq \|u_{n+1}\|(\|u_n\| - \|u_{n+1}\|),
 \end{aligned} \tag{23.32}$$

which implies that  $(\|u_n\|)_{n \in \mathbb{N}}$  converges. Now let  $z \in \text{zer } A$ . Since Proposition 23.38 asserts that  $z \in \bigcap_{n \in \mathbb{N}} \text{Fix } J_{\gamma_n A}$ , we deduce from (23.31), Corollary 23.10(i), and (4.1) that

$$\begin{aligned}
 (\forall n \in \mathbb{N}) \quad \|x_{n+1} - z\|^2 &= \|J_{\gamma_n A} x_n - J_{\gamma_n A} z\|^2 \\
 &\leq \|x_n - z\|^2 - \|x_n - J_{\gamma_n A} x_n\|^2 \\
 &= \|x_n - z\|^2 - \gamma_n^2 \|u_n\|^2.
 \end{aligned} \tag{23.33}$$

Thus,  $(x_n)_{n \in \mathbb{N}}$  is bounded and Fejér monotone with respect to  $\text{zer } A$ , and  $\sum_{n \in \mathbb{N}} \gamma_n^2 \|u_n\|^2 < +\infty$ . In turn, since  $\sum_{n \in \mathbb{N}} \gamma_n^2 = +\infty$ , we have  $\liminf \|u_n\| = 0$  and therefore  $u_n \rightarrow 0$  since  $(\|u_n\|)_{n \in \mathbb{N}}$  converges.

(i): Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ , say  $x_{k_n+1} \rightharpoonup x$ . Since  $0 \leftarrow u_{k_n} \in Ax_{k_n+1}$ , Proposition 20.33(ii) yields  $0 \in Ax$  and we derive the result from Theorem 5.5.

(ii): The assumptions imply that  $A$  is strictly monotone. Hence, by Proposition 23.35,  $\text{zer } A$  reduces to a singleton. Now let  $x$  be the weak limit in (i). Then  $0 \in Ax$  and  $(\forall n \in \mathbb{N}) \ u_n \in Ax_{n+1}$ . Hence, since  $(x_n)_{n \in \mathbb{N}}$  lies in a bounded set, there exists an increasing function  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \langle x_{n+1} - x \mid u_n \rangle \geq \phi(\|x_{n+1} - x\|). \tag{23.34}$$

Since  $u_n \rightarrow 0$  and  $x_{n+1} \rightharpoonup x$ , Lemma 2.41(iii) yields  $\|x_{n+1} - x\| \rightarrow 0$ .  $\square$

## 23.5 Asymptotic Behavior

In view of Proposition 20.31 and Theorem 3.14, the following notion is well defined.

**Definition 23.42** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \text{dom } A$ . Then  ${}^0Ax$  denotes the element in  $Ax$  of minimal norm.

**Proposition 23.43** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , let  $\mu \in \mathbb{R}_{++}$ , and let  $x \in \text{dom } A$ . Then the following hold:*

- (i)  $\|\gamma Ax\| \leq \inf \|Ax\|$ .
- (ii)  $\|\gamma^{+\mu} Ax\| \leq \|\gamma Ax\| \leq \|^0 Ax\|$ .

*Proof.* (i): Set  $p = \gamma Ax$  and let  $u \in Ax$ . Proposition 23.2(iii) and the monotonicity of  $A$  yield  $\langle u - p \mid p \rangle = \langle u - p \mid x - (x - \gamma p) \rangle / \gamma \geq 0$ . Hence, by Cauchy–Schwarz,  $\|p\| \leq \|u\|$ .

(ii): By (i),  $\|^{\mu} Ax\| \leq \|^0 Ax\|$ . Applying this inequality to  $\gamma A$ , which is maximally monotone by Corollary 23.10(iv), and, using Proposition 23.6(iii), we get  $\|\gamma^{+\mu} Ax\| = \|^{\mu}(\gamma A)x\| \leq \|^0(\gamma A)x\| = \|\gamma Ax\|$ .  $\square$

We now present a central perturbation result.

**Theorem 23.44** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator and let  $x \in \mathcal{H}$ . Then the inclusions*

$$(\forall \gamma \in ]0, 1[) \quad 0 \in Ax_{\gamma} + \gamma(x_{\gamma} - x) \quad (23.35)$$

*define a unique curve  $(x_{\gamma})_{\gamma \in ]0, 1[}$ . Moreover, exactly one of the following holds:*

- (i)  $\text{zer } A \neq \emptyset$  and  $x_{\gamma} \rightarrow P_{\text{zer } A} x$  as  $\gamma \downarrow 0$ .
- (ii)  $\text{zer } A = \emptyset$  and  $\|x_{\gamma}\| \rightarrow +\infty$  as  $\gamma \downarrow 0$ .

*Proof.* It follows from (23.35), Proposition 23.2(ii), and Corollary 23.8 that, for every  $\gamma \in ]0, 1[$ ,

$$x_{\gamma} = J_{A/\gamma} x \quad (23.36)$$

is well defined. We now proceed in two steps.

(a)  $\text{zer } A \neq \emptyset \Rightarrow x_{\gamma} \rightarrow P_{\text{zer } A} x$  as  $\gamma \downarrow 0$ : Set  $x_0 = P_{\text{zer } A} x$ , which is well defined and characterized by

$$x_0 \in \text{zer } A \quad \text{and} \quad (\forall z \in \text{zer } A) \quad \langle z - x_0 \mid x - x_0 \rangle \leq 0 \quad (23.37)$$

by virtue of Proposition 23.39 and Theorem 3.14. Now let  $z \in \text{zer } A$ . Then Proposition 23.38 yields  $(\forall \gamma \in ]0, 1[) \quad z = J_{A/\gamma} z$ . Therefore, it follows from (23.36) and Corollary 23.10(i) that

$$(\forall \gamma \in ]0, 1[) \quad \langle z - x_{\gamma} \mid z - x \rangle \geq \|z - x_{\gamma}\|^2. \quad (23.38)$$

Thus, by Cauchy–Schwarz,  $(x_{\gamma})_{\gamma \in ]0, 1[}$  is bounded. Now let  $(\gamma_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1[$  such that  $\gamma_n \downarrow 0$  as  $n \rightarrow +\infty$ . Then it is enough to show that  $x_{\gamma_n} \rightarrow x_0$ . To this end, let  $y$  be a weak sequential cluster point of  $(x_{\gamma_n})_{n \in \mathbb{N}}$ , say  $x_{\gamma_{k_n}} \rightharpoonup y$ . Since  $(x_{\gamma_{k_n}})_{n \in \mathbb{N}}$  is bounded, the sequence  $(\gamma_{k_n}(x - x_{\gamma_{k_n}}), x_{\gamma_{k_n}})_{n \in \mathbb{N}}$ , which lies in  $\text{gra } A$  by virtue of (23.35), converges to  $(0, y)$  in  $\mathcal{H}^{\text{strong}} \times \mathcal{H}^{\text{weak}}$ . Hence, it follows from Proposition 20.33(i) that  $y \in \text{zer } A$ . In turn, we derive from (23.38) that  $0 \leftarrow \langle y - x \mid y - x_{\gamma_{k_n}} \rangle \geq \|y - x_{\gamma_{k_n}}\|^2$  and therefore that  $x_{\gamma_{k_n}} \rightarrow y$ . However, (23.38) yields  $0 \geq$

$\|z - x_{\gamma_{k_n}}\|^2 - \langle z - x_{\gamma_{k_n}} | z - x \rangle = \langle z - x_{\gamma_{k_n}} | x - x_{\gamma_{k_n}} \rangle \rightarrow \langle z - y | x - y \rangle$  and, in view of the characterization (23.37), we obtain  $y = x_0$ . Altogether,  $x_0$  is the only weak sequential cluster point of the bounded sequence  $(x_{\gamma_n})_{n \in \mathbb{N}}$ , and it follows from Lemma 2.38 that  $x_{\gamma_n} \rightharpoonup x_0$ . Invoking (23.38), we obtain  $\|x_{\gamma_n} - x_0\|^2 \leq \langle x - x_0 | x_{\gamma_n} - x_0 \rangle \rightarrow 0$  and therefore  $x_{\gamma_n} \rightarrow x_0$ .

(b)  $\|x_\gamma\| \not\rightarrow +\infty$  as  $\gamma \downarrow 0 \Rightarrow \text{zer } A \neq \emptyset$ : Take a sequence  $(\gamma_n)_{n \in \mathbb{N}}$  in  $]0, 1[$  such that  $\gamma_n \downarrow 0$  and  $(x_{\gamma_n})_{n \in \mathbb{N}}$  is bounded. Then (23.35) yields  $0 \leftarrow \gamma_n(x - x_{\gamma_n}) \in Ax_{\gamma_n}$ . Furthermore,  $(x_{\gamma_n})_{n \in \mathbb{N}}$  possesses a weak sequential cluster point  $y$ , and Proposition 20.33(ii) forces  $y \in \text{zer } A$ .  $\square$

We now revisit the approximating curve investigated in Proposition 4.20 and analyze it with different tools.

**Corollary 23.45** *Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be nonexpansive and let  $x \in \mathcal{H}$ . Then the equations*

$$(\forall \gamma \in ]0, 1[) \quad x_\gamma = \gamma x + (1 - \gamma)Tx_\gamma \quad (23.39)$$

*define a unique curve  $(x_\gamma)_{\gamma \in ]0, 1[}$ . Moreover, exactly one of the following holds:*

- (i)  $\text{Fix } T \neq \emptyset$  and  $x_\gamma \rightarrow P_{\text{Fix } T} x$  as  $\gamma \downarrow 0$ .
- (ii)  $\text{Fix } T = \emptyset$  and  $\|x_\gamma\| \rightarrow +\infty$  as  $\gamma \downarrow 0$ .

*Proof.* Set  $A = \text{Id} - T$ . Then  $\text{zer } A = \text{Fix } T$ ,  $A$  is maximally monotone by Example 20.26, and (23.39) becomes

$$(\forall \gamma \in ]0, 1[) \quad 0 = Ax_\gamma + \frac{\gamma}{1 - \gamma}(x_\gamma - x). \quad (23.40)$$

Since  $\lim_{\gamma \downarrow 0} \gamma/(1 - \gamma) = 0$ , the conclusion follows from Theorem 23.44.  $\square$

**Corollary 23.46** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \mathcal{H}$ . Then exactly one of the following holds:*

- (i)  $x \in \text{dom } A$ , and  ${}^\gamma Ax \rightarrow {}^0 Ax$  and  $\|{}^\gamma Ax\| \uparrow \|{}^0 Ax\|$  as  $\gamma \downarrow 0$ .
- (ii)  $x \notin \text{dom } A$  and  $\|{}^\gamma Ax\| \uparrow +\infty$  as  $\gamma \downarrow 0$ .

*Proof.* Set  $B = A^{-1} - x$ . Then  $\text{zer } B = Ax$ . Moreover,  $B$  is maximally monotone and, by Theorem 23.44, the inclusions

$$(\forall \gamma \in ]0, 1[) \quad 0 \in Bx_\gamma + \gamma(x_\gamma - 0) \quad (23.41)$$

define a unique curve  $(x_\gamma)_{\gamma \in ]0, 1[}$ . However, it follows from Proposition 23.6(ii) that  $(\forall \gamma \in ]0, 1[) \quad 0 \in Bx_\gamma + \gamma x_\gamma \Leftrightarrow 0 \in A^{-1}x_\gamma - x + \gamma x_\gamma \Leftrightarrow x \in (\gamma \text{Id} + A^{-1})x_\gamma \Leftrightarrow x_\gamma = {}^\gamma Ax$ .

(i): Suppose that  $x \in \text{dom } A$ . Then  $\text{zer } B \neq \emptyset$ , and it follows from Theorem 23.44(i) that  ${}^\gamma Ax = x_\gamma \rightarrow P_{\text{zer } B} 0 = {}^0 Ax$  as  $\gamma \downarrow 0$ . In turn, we derive from Proposition 23.43(ii) that  $\|{}^\gamma Ax\| \uparrow \|{}^0 Ax\|$  as  $\gamma \downarrow 0$ .

(ii): Suppose that  $x \notin \text{dom } A$ . Then  $\text{zer } B = \emptyset$ , and the assertion therefore follows from Theorem 23.44(ii).  $\square$

**Theorem 23.47** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $x \in \mathcal{H}$ . Set

$$(\forall \gamma \in \mathbb{R}_{++}) \quad x_\gamma = J_{\gamma A} x. \quad (23.42)$$

Then  $x_\gamma \rightarrow P_{\overline{\text{dom } A}} x$  as  $\gamma \downarrow 0$ , and exactly one of the following holds:

- (i)  $\text{zer } A \neq \emptyset$  and  $x_\gamma \rightarrow P_{\text{zer } A} x$  as  $\gamma \uparrow +\infty$ .
- (ii)  $\text{zer } A = \emptyset$  and  $\|x_\gamma\| \rightarrow +\infty$  as  $\gamma \uparrow +\infty$ .

*Proof.* Set  $D = \overline{\text{dom } A}$  and  $Z = \text{zer } A$ . It follows from Corollary 21.12 that  $D$  is nonempty, closed, and convex. Hence, by Theorem 3.14,  $P_D$  is well defined. Now let  $\gamma \in ]0, 1[$  and let  $(y, v) \in \text{gra } A$ . It follows from Proposition 23.2(ii) and the monotonicity of  $A$  that  $\langle x_\gamma - y \mid x - x_\gamma - \gamma v \rangle \geq 0$ , which yields

$$\|x_\gamma - y\|^2 \leq \langle x_\gamma - y \mid x - y \rangle + \gamma \|x_\gamma - y\| \|v\|. \quad (23.43)$$

As a result,  $\|x_\gamma - y\| \leq \|x - y\| + \gamma \|v\|$ , and  $(x_\gamma)_{\gamma \in ]0, 1[}$  is therefore bounded. Now let  $(\gamma_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, 1[$  such that  $\gamma_n \downarrow 0$  as  $n \rightarrow +\infty$ . Let  $z$  be a weak sequential cluster point of  $(x_{\gamma_n})_{n \in \mathbb{N}}$ , say  $x_{\gamma_{k_n}} \rightharpoonup z$ . Note that  $z \in D$  since  $(x_{\gamma_{k_n}})_{n \in \mathbb{N}}$  lies in  $\text{dom } A \subset D$  and since  $D$  is weakly sequentially closed by Theorem 3.32. On the other hand, since  $(x_{\gamma_{k_n}})_{n \in \mathbb{N}}$  is bounded, Lemma 2.35 and (23.43) yield  $\|z - y\|^2 \leq \varliminf \|x_{\gamma_{k_n}} - y\|^2 \leq \langle z - y \mid x - y \rangle$  and therefore  $\langle x - z \mid y - z \rangle \leq 0$ . Since  $y$  is an arbitrary point in  $\text{dom } A$ , this inequality holds for every  $y \in D$  and, in view of Theorem 3.14, it follows that  $z = P_D x$  is the only weak sequential cluster point of the bounded sequence  $(x_{\gamma_n})_{n \in \mathbb{N}}$ . Hence, by Lemma 2.38,  $x_{\gamma_n} \rightharpoonup P_D x$ . It follows from (23.43) that

$$\varlimsup \|x_{\gamma_n} - y\|^2 \leq \langle P_D x - y \mid x - y \rangle. \quad (23.44)$$

Now set  $f: \mathcal{H} \rightarrow \mathbb{R}: z \mapsto \varlimsup \|x_{\gamma_n} - z\|^2$ . Then  $f \geq 0$  and it follows from Example 8.17 that  $f$  is continuous. Now let  $(y_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{dom } A$  that converges strongly to  $P_D x$ . Then, using (23.44), we obtain  $0 \leq f(P_D x) = \lim f(y_n) \leq \lim \langle P_D x - y_n \mid x - y_n \rangle = 0$ . Hence  $0 = f(P_D x) = \varlimsup \|x_{\gamma_n} - P_D x\|^2$  and therefore  $x_{\gamma_n} \rightarrow P_D x$ . Thus,  $x_\gamma \rightarrow P_D x$  as  $\gamma \downarrow 0$ .

(i)&(ii): Apply Theorem 23.44 since (23.42) is equivalent to  $(\forall \gamma \in \mathbb{R}_{++}) 0 \in Ax_\gamma + \gamma^{-1}(x_\gamma - x)$ . □

## Exercises

**Exercise 23.1** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and at most single-valued. Define the *Cayley transform*  $C_A: \mathcal{H} \rightarrow \mathcal{H}$  of  $A$  by  $C_A = (\text{Id} - A)(\text{Id} + A)^{-1}$ . Show that  $C_A = 2J_A - \text{Id}$  and determine  $C_{C_A}$ . What happens when  $A$  is linear with  $\text{dom } A = \mathcal{H}$ ?

**Exercise 23.2** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and let  $\mu \in \mathbb{R}_{++}$ . Show that

$$(\forall x \in \mathcal{H}) \quad \|J_{\gamma A}x - J_{\mu A}x\| \leq |\gamma - \mu| \|\gamma Ax\|. \quad (23.45)$$

**Exercise 23.3** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and let  $\gamma \in \mathbb{R}_{++}$ . Show that

$$\text{Id} = \gamma \text{Id} \circ \gamma A + \gamma^{-1}(A^{-1}) \circ \gamma^{-1} \text{Id} \quad (23.46)$$

and deduce that  $\text{Id} = {}^1A + {}^1(A^{-1})$ .

**Exercise 23.4** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and set  $T = \text{Id}|_D$ . Determine  $\tilde{T}$ , where  $\tilde{T}$  is as in the conclusion of Theorem 23.13.

**Exercise 23.5** Let  $D$  be a nonempty subset of  $\mathcal{H}$  and set  $T = \text{Id}|_D$ . Determine  $\text{Fix } \tilde{T}$  and  $\text{ran } \tilde{T}$ , where  $\tilde{T}$  is as in the conclusion of Corollary 23.14. Give examples in which (i)  $\tilde{T}$  is unique, and (ii)  $\tilde{T}$  is not unique.

**Exercise 23.6** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Show that  $\text{dom}(A) + \text{ran}(A) = \mathcal{H}$ .

**Exercise 23.7** Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $\text{dom}(\partial f) + \text{dom}(\partial f^*) = \mathcal{H}$ . Compare to Exercise 15.3.

**Exercise 23.8** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. Prove the following:

- (i)  $J_A$  is injective if and only if  $A$  is at most single-valued.
- (ii)  $J_A$  is surjective if and only if  $\text{dom } A = \mathcal{H}$ .

**Exercise 23.9** Prove Proposition 23.17.

**Exercise 23.10** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\gamma \in \mathbb{R}_{++}$ , and set  $L: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: (x, u) \mapsto (x + \gamma u, x)$ . Show that  $\text{gra } J_{\gamma A} = L(\text{gra } A)$ .

**Exercise 23.11** Give the details of the proof of Corollary 23.33.

**Exercise 23.12** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator and let  $x \in \mathcal{H}$ . Use Corollary 23.37(ii) to show that the inclusions (23.35) define a unique curve  $(x_{\gamma})_{\gamma \in ]0,1[}$ .

**Exercise 23.13** Without using Theorem 23.44, show directly that the equations (23.39) define a unique curve  $(x_{\gamma})_{\gamma \in ]0,1[}$ .

**Exercise 23.14** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator such that  $\text{zer } A \neq \emptyset$ , and let  $x \in \mathcal{H}$ . Show that  $\gamma Ax \rightarrow 0$  as  $\gamma \uparrow +\infty$ .

**Exercise 23.15** Let  $A \in \mathcal{B}(\mathcal{H})$  be such that  $\text{ran } A$  is closed. Show that

$${}^0(A^{-1}) = A^{\dagger}|_{\text{ran } A}, \quad (23.47)$$

where  $A^{-1}$  is the set-valued inverse.



# Chapter 24

## Sums of Monotone Operators

The sum of two monotone operators is monotone. However, maximal monotonicity of the sum of two maximally monotone operators is not automatic and requires additional assumptions. In this chapter, we provide flexible sufficient conditions for maximality. Under additional assumptions, conclusions can be drawn about the range of the sum, which is important for deriving surjectivity results. Consideration is also given to the parallel sum of two set-valued operators.

Throughout this chapter,  $F_A$  designates the Fitzpatrick function of a monotone operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  (see Definition 20.42).

### 24.1 Maximal Monotonicity of a Sum

As the following example shows, the sum of two maximally monotone operators may not be maximally monotone.

**Example 24.1** Suppose that  $\mathcal{H} = \mathbb{R}^2$ , and set  $C = B((-1, 0); 1)$  and  $D = B((1, 0); 1)$ . Then  $N_C$  and  $N_D$  are maximally monotone by Example 20.41, with  $(\text{dom } N_C) \cap (\text{dom } N_D) \neq \emptyset$ . However, it follows from Corollary 21.21 and the fact that  $\text{ran}(N_C + N_D) = \mathbb{R} \times \{0\}$  that  $N_C + N_D$  is not maximally monotone.

**Theorem 24.2** Set  $Q_1: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (x, u) \mapsto x$ , and let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that

$$0 \in \text{sri } Q_1(\text{dom } F_A - \text{dom } F_B). \quad (24.1)$$

Then  $A + B$  is maximally monotone.

*Proof.* Proposition 20.48 and Proposition 20.47(iv) imply that, for every  $(x, u_1, u_2)$  in  $\mathcal{H}^3$ ,

$$\langle x \mid u_1 + u_2 \rangle \leq F_A(x, u_1) + F_B(x, u_2) \leq F_A^*(u_1, x) + F_B^*(u_2, x). \quad (24.2)$$

Thus the function

$$F: \mathcal{H} \times \mathcal{H} \rightarrow ]-\infty, +\infty] : (x, u) \mapsto (F_A(x, \cdot) \square F_B(x, \cdot))(u) \quad (24.3)$$

is proper and convex, and it satisfies

$$F \geq \langle \cdot \mid \cdot \rangle. \quad (24.4)$$

Corollary 15.8 and (24.2) yield

$$\begin{aligned} (\forall (u, x) \in \mathcal{H} \times \mathcal{H}) \quad F^*(u, x) &= (F_A^*(\cdot, x) \square F_B^*(\cdot, x))(u) \\ &\geq F(x, u) \\ &\geq \langle x \mid u \rangle. \end{aligned} \quad (24.5)$$

Now fix  $(x, u) \in \mathcal{H} \times \mathcal{H}$  and assume that  $F^*(u, x) = \langle x \mid u \rangle$ . Then (24.5) and Proposition 20.51(iii) guarantee the existence of  $u_1$  and  $u_2$  in  $\mathcal{H}$  such that  $u_1 + u_2 = u$  and  $\langle x \mid u \rangle = F^*(u, x) = F_A^*(u_1, x) + F_B^*(u_2, x) \geq \langle x \mid u_1 \rangle + \langle x \mid u_2 \rangle = \langle x \mid u \rangle$ . It follows that  $F_A^*(u_1, x) = \langle x \mid u_1 \rangle$  and that  $F_B^*(u_2, x) = \langle x \mid u_2 \rangle$ . By (20.40),  $(x, u) = (x, u_1 + u_2) \in \text{gra}(A + B)$ . Now assume that  $(x, u) \in \text{gra}(A + B)$ . Then there exist  $u_1 \in Ax$  and  $u_2 \in Bx$  such that  $u = u_1 + u_2$ . Hence  $F_A^*(u_1, x) = \langle x \mid u_1 \rangle$  and  $F_B^*(u_2, x) = \langle x \mid u_2 \rangle$ , by (20.40). This and (24.5) imply that  $\langle x \mid u \rangle = \langle x \mid u_1 \rangle + \langle x \mid u_2 \rangle = F_A^*(u_1, x) + F_B^*(u_2, x) \geq (F_A^*(\cdot, x) \square F_B^*(\cdot, x))(u) = F^*(u, x) \geq \langle x \mid u \rangle$ . Therefore,  $F^*(u, x) = \langle x \mid u \rangle$ . Altogether, we have shown that

$$\{(x, u) \in \mathcal{H} \times \mathcal{H} \mid F^*(u, x) = \langle x \mid u \rangle\} = \text{gra}(A + B). \quad (24.6)$$

Combining (24.5), (24.6), (24.4), and Theorem 20.38(ii), we conclude that  $A + B$  is maximally monotone.  $\square$

**Theorem 24.3** *Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that*

$$\text{cone}(\text{dom } A - \text{dom } B) = \overline{\text{span}}(\text{dom } A - \text{dom } B). \quad (24.7)$$

*Then  $A + B$  is maximally monotone.*

*Proof.* Using Proposition 21.11 and its notation, we obtain

$$\begin{aligned} \text{cone}(\text{dom } A - \text{dom } B) &\subset \text{cone}(Q_1(\text{dom } F_A) - Q_1(\text{dom } F_B)) \\ &\subset \overline{\text{span}}(Q_1(\text{dom } F_A) - Q_1(\text{dom } F_B)) \\ &\subset \overline{\text{span}}(\overline{\text{dom } A} - \overline{\text{dom } B}) \\ &= \overline{\text{span}}(\text{dom } A - \text{dom } B) \\ &= \text{cone}(\text{dom } A - \text{dom } B). \end{aligned} \quad (24.8)$$

Hence (24.1) holds and the result follows from Theorem 24.2.  $\square$

**Corollary 24.4** *Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that one of the following holds:*

- (i)  $\text{dom } B = \mathcal{H}$ .
- (ii)  $\text{dom } A \cap \text{int dom } B \neq \emptyset$ .
- (iii)  $0 \in \text{int}(\text{dom } A - \text{dom } B)$ .
- (iv)  $(\forall x \in \mathcal{H})(\exists \varepsilon \in \mathbb{R}_{++}) [0, \varepsilon] \subset \text{dom } A - \text{dom } B$ .
- (v)  $\text{dom } A$  and  $\text{dom } B$  are convex and

$$0 \in \text{sri}(\text{dom } A - \text{dom } B). \quad (24.9)$$

Then  $A + B$  is maximally monotone.

*Proof.* (i) $\Rightarrow$ (ii) $\Rightarrow$ (iii) $\Rightarrow$ (iv) $\Rightarrow$ (24.7): Clear.

(v): By (6.8), (24.7) is equivalent to (24.9).  $\square$

**Theorem 24.5** *Let  $\mathcal{K}$  be a real Hilbert space, let  $A: \mathcal{K} \rightarrow 2^{\mathcal{K}}$  be maximally monotone, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and suppose that  $\text{cone}(\text{ran } L - \text{dom } A) = \overline{\text{span}}(\text{ran } L - \text{dom } A)$ . Then  $L^*AL$  is maximally monotone.*

*Proof.* We work in  $\mathcal{H} \oplus \mathcal{K}$ . Using Proposition 20.23, we see that the operators

$$B = N_{\text{gra } L} \text{ and } C: \mathcal{H} \times \mathcal{K} \rightarrow 2^{\mathcal{H} \times \mathcal{K}}: (x, y) \mapsto \{0\} \times Ay \quad (24.10)$$

are maximally monotone with  $\text{dom } B = \text{gra } L$  and  $\text{dom } C = \mathcal{H} \times \text{dom } A$ . Hence  $\text{dom}(B + C) \subset \text{gra } L$ . We claim that

$$\begin{aligned} (\forall (x, u) \in \mathcal{H} \times \mathcal{H})(\forall v \in \mathcal{K}) \\ (u, v) \in (B + C)(x, Lx) \Leftrightarrow u + L^*v \in L^*(A(Lx)). \end{aligned} \quad (24.11)$$

Let  $(x, u) \in \mathcal{H} \times \mathcal{H}$  and  $v \in \mathcal{K}$ . Then  $(B + C)(x, Lx) = N_{\text{gra } L}(x, Lx) + (\{0\} \times A(Lx))$  and hence

$$\begin{aligned} (u, v) \in (B + C)(x, Lx) &\Leftrightarrow (u, v) \in N_{\text{gra } L}(x, Lx) + (\{0\} \times A(Lx)) \\ &\Leftrightarrow (u, v) \in \{(L^*w, -w) \mid w \in \mathcal{K}\} + (\{0\} \times A(Lx)) \\ &\Leftrightarrow (\exists w \in \mathcal{K}) \quad u = L^*w \quad \text{and} \quad v + w \in A(Lx) \\ &\Leftrightarrow u + L^*v \in L^*(A(Lx)). \end{aligned} \quad (24.12)$$

This verifies (24.11). Since  $\text{dom } B - \text{dom } C = (\text{gra } L) - (\mathcal{H} \times \text{dom } A) = \mathcal{H} \times (\text{ran } L - \text{dom } A)$  and  $\text{cone}(\text{ran } L - \text{dom } A) = \overline{\text{span}}(\text{ran } L - \text{dom } A)$ , we deduce that  $\text{cone}(\text{dom } B - \text{dom } C) = \overline{\text{span}}(\text{dom } B - \text{dom } C)$ . Hence, Theorem 24.3 implies that  $B + C$  is maximally monotone. Now let  $(z, w)$  be monotonically related to  $\text{gra}(L^*AL)$ , i.e.,

$$(\forall x \in \mathcal{H}) \quad \inf \langle x - z \mid L^*(A(Lx)) - w \rangle \geq 0. \quad (24.13)$$

Now take  $(x, u) \in \mathcal{H} \times \mathcal{H}$  and  $v \in \mathcal{K}$  such that  $((x, Lx), (u, v)) \in \text{gra}(B + C)$ . Then, by (24.11),  $u + L^*v \in L^*(A(Lx))$ . Moreover, (24.13) yields

$$\begin{aligned} 0 &\leq \langle x - z \mid u + L^*v - w \rangle \\ &= \langle x - z \mid u - w \rangle + \langle Lx - Lz \mid v \rangle \\ &= \langle (x, Lx) - (z, Lz) \mid (u, v) - (w, 0) \rangle. \end{aligned} \quad (24.14)$$

The maximal monotonicity of  $B + C$  implies that  $((z, Lz), (w, 0)) \in \text{gra}(B + C)$ . In view of (24.11), this is equivalent to  $w \in L^*(A(Lz))$ . Appealing to (20.16), we conclude that  $L^*AL$  is maximally monotone.  $\square$

**Remark 24.6** We have proved the composition result (Theorem 24.5) by applying the sum result (Theorem 24.3) in a suitable product space. It is also possible to proceed in the opposite direction. Indeed, let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Then  $A \times B: \mathcal{H} \times \mathcal{H} \rightarrow 2^{\mathcal{H} \times \mathcal{H}}$  is maximally monotone. Set  $L: \mathcal{H} \rightarrow \mathcal{H} \times \mathcal{H}: x \mapsto (x, x)$ . Then  $L^*: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (u, v) \mapsto u + v$  and hence  $L^*(A \times B)L = A + B$ . Therefore, maximal monotonicity of the composition  $L^*(A \times B)L$  implies maximal monotonicity of the sum  $A + B$ .

## 24.2 3\* Monotone Operators

**Definition 24.7** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone. Then  $A$  is 3\* monotone if

$$\text{dom } A \times \text{ran } A \subset \text{dom } F_A. \quad (24.15)$$

**Proposition 24.8** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and 3\* monotone. Then  $\overline{\text{dom } F_A} = \overline{\text{dom } A} \times \overline{\text{ran } A}$ .

*Proof.* Combine (24.15) and Corollary 21.12.  $\square$

**Example 24.9** Let  $f \in \Gamma_0(\mathcal{H})$ . Then  $\partial f$  is 3\* monotone.

*Proof.* Using Example 20.46, we obtain  $\text{dom } \partial f \times \text{ran } \partial f = \text{dom } \partial f \times \text{dom } \partial f^* \subset \text{dom } f \times \text{dom } f^* \subset \text{dom } F_{\partial f}$ .  $\square$

**Example 24.10** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . Then  $N_C$  is 3\* monotone.

*Proof.* Apply Example 24.9 to  $f = \iota_C$  and use Example 16.12.  $\square$

**Example 24.11** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and uniformly monotone with a supercoercive modulus  $\phi$ . Then  $A$  is 3\* monotone.

*Proof.* Fix  $(x, u) \in \text{gra } A$  and  $w \in \mathcal{H}$ , and set  $\gamma = \|u - w\|$  and  $\psi: \mathbb{R}_+ \rightarrow [-\infty, +\infty[ : t \mapsto \gamma t - \phi(t)$ . Since  $\lim_{t \rightarrow +\infty} \phi(t)/t = +\infty$ , we

can find  $\tau \in \mathbb{R}_+$  such that  $\psi(t) < 0 = \psi(0)$  whenever  $t > \tau$ . Thus,  $\sup_{t \in \mathbb{R}_+} \psi(t) = \sup_{t \in [0, \tau]} \psi(t) \leq \gamma\tau < +\infty$ . Therefore, (20.32), (22.3), and Cauchy-Schwarz yield

$$\begin{aligned} F_A(x, w) - \langle x \mid w \rangle &= \sup_{(y, v) \in \text{gra } A} (\langle x - y \mid v - u \rangle + \langle x - y \mid u - w \rangle) \\ &\leq \sup_{y \in \text{dom } A} \psi(\|x - y\|) \\ &< +\infty. \end{aligned} \quad (24.16)$$

In other words,  $(x, w) \in \text{dom } F_A$  and hence

$$(\text{dom } A) \times \mathcal{H} \subset \text{dom } F_A. \quad (24.17)$$

Therefore,  $A$  is 3\* monotone.  $\square$

**Proposition 24.12** *Let  $A \in \mathcal{B}(\mathcal{H})$  be monotone. Then the following are equivalent for some  $\beta \in \mathbb{R}_{++}$ :*

- (i)  $A$  is 3\* monotone.
- (ii)  $A$  is  $\beta$ -cocoercive.
- (iii)  $A^*$  is  $\beta$ -cocoercive.
- (iv)  $A^*$  is 3\* monotone.

*Proof.*

(i) $\Rightarrow$ (ii): Set

$$f: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto F_A(x, 0). \quad (24.18)$$

Then  $f \in \Gamma_0(\mathcal{H})$  by Proposition 20.47(i)&(ii). Since  $A$  is 3\* monotone, we see that  $\mathcal{H} \times \{0\} \subset \text{dom } A \times \text{ran } A \subset \text{dom } F_A$ . Hence  $\text{dom } f = \mathcal{H}$ . By Corollary 8.30(ii),  $0 \in \text{cont } f$  and thus there exist  $\rho \in \mathbb{R}_{++}$  and  $\mu \in \mathbb{R}_{++}$  such that  $(\forall x \in B(0; \rho)) f(x) = F_A(x, 0) = \sup_{y \in \mathcal{H}} (\langle x \mid Ay \rangle - \langle y \mid Ay \rangle) \leq \mu$ . Now let  $x \in B(0; \rho)$  and  $y \in \mathcal{H}$ . Then

$$\begin{aligned} (\forall t \in \mathbb{R}) \quad 0 &\leq \mu + \langle ty \mid A(ty) \rangle - \langle x \mid A(ty) \rangle \\ &= \mu + t^2 \langle y \mid Ay \rangle - t \langle x \mid Ay \rangle. \end{aligned} \quad (24.19)$$

This implies that

$$(\forall x \in B(0; \rho)) (\forall y \in \mathcal{H}) \quad \langle x \mid Ay \rangle^2 \leq 4\mu \langle y \mid Ay \rangle. \quad (24.20)$$

Now let  $y \in \mathcal{H} \setminus \ker A$  and set  $x = (\rho/\|Ay\|)Ay \in B(0; \rho)$ . By (24.20), we arrive at  $4\mu \langle y \mid Ay \rangle \geq (\rho/\|Ay\|)^2 \|Ay\|^4$ , i.e.,  $\langle y \mid Ay \rangle \geq \|Ay\|^2 \rho^2 / (4\mu)$ . The last inequality is also true when  $y \in \ker A$ ; hence we set  $\beta = \rho^2 / (4\mu)$ , and we deduce that, for every  $y \in \mathcal{H}$ ,  $\langle y \mid \beta Ay \rangle \geq \|\beta Ay\|^2$ . Therefore, since  $A$  is linear, we deduce that  $A$  is  $\beta$ -cocoercive.

(ii) $\Leftrightarrow$ (iii): Clear from Corollary 4.3.

(ii) $\Rightarrow$ (i) & (iii) $\Rightarrow$ (i): Fix  $x$  and  $y$  in  $\mathcal{H}$ , and take  $z \in \mathcal{H}$ . Then

$$\begin{aligned}
& \langle x \mid Az \rangle + \langle z \mid Ay \rangle - \langle z \mid Az \rangle \\
&= \left( \langle x \mid Az \rangle - \frac{1}{2} \langle z \mid Az \rangle \right) + \left( \langle A^* z \mid y \rangle - \frac{1}{2} \langle z \mid A^* z \rangle \right) \\
&\leq \left( \|x\| \|Az\| - \frac{1}{2} \beta \|Az\|^2 \right) + \left( \|A^* z\| \|y\| - \frac{1}{2} \beta \|A^* z\|^2 \right) \\
&\leq \frac{1}{2\beta} (\|x\|^2 + \|y\|^2), \tag{24.21}
\end{aligned}$$

where (24.21) was obtained by computing the maximum of the quadratic functions  $t \mapsto \|x\|t - (1/2)\beta t^2$  and  $t \mapsto \|y\|t - (1/2)\beta t^2$ . It follows from (24.21) that

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad F_A(x, Ay) \leq \frac{1}{2\beta} (\|x\|^2 + \|y\|^2). \tag{24.22}$$

Hence  $\mathcal{H} \times \text{ran } A \subset \text{dom } F_A$ .

(iii) $\Leftrightarrow$ (iv): Apply the already established equivalence (i) $\Leftrightarrow$ (ii) to  $A^*$ .  $\square$

**Example 24.13** Let  $\mathcal{K}$  be a real Hilbert space and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Then  $LL^*$  is  $3^*$  monotone.

*Proof.* Combine Example 24.9, Example 2.48, and Proposition 17.26(i). Alternatively, combine Corollary 18.17 and Proposition 24.12.  $\square$

**Example 24.14** Let  $N$  be a strictly positive integer, and let  $R: \mathcal{H}^N \rightarrow \mathcal{H}^N: (x_1, x_2, \dots, x_N) \mapsto (x_N, x_1, \dots, x_{N-1})$  be the *cyclic right-shift operator* in  $\mathcal{H}^N$ . Then  $\text{Id} - R$  is  $3^*$  monotone.

*Proof.* By Proposition 4.2 and Proposition 24.12,  $-R$  is nonexpansive  $\Leftrightarrow 2((1/2)(\text{Id} - R)) - \text{Id}$  is nonexpansive  $\Leftrightarrow (1/2)(\text{Id} - R)$  is firmly nonexpansive  $\Rightarrow \text{Id} - R$  is  $3^*$  monotone.  $\square$

**Proposition 24.15** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then the following hold:

- (i)  $A$  is  $3^*$  monotone if and only if  $A^{-1}$  is  $3^*$  monotone.
- (ii)  $A$  is  $3^*$  monotone if and only if  $\gamma A$  is  $3^*$  monotone.

*Proof.* (i): By Proposition 20.47(vi),  $A$  is  $3^*$  monotone  $\Leftrightarrow \text{dom } A \times \text{ran } A \subset \text{dom } F_A \Leftrightarrow \text{ran } A \times \text{dom } A \subset \text{dom } F_A^\top \Leftrightarrow \text{dom } A^{-1} \times \text{ran } A^{-1} \subset \text{dom } F_{A^{-1}} \Leftrightarrow A^{-1}$  is  $3^*$  monotone.

(ii): By Proposition 20.47(vii),  $A$  is  $3^*$  monotone  $\Leftrightarrow \text{dom } A \times \text{ran } A \subset \text{dom } F_A \Leftrightarrow \text{dom } A \times \gamma \text{ran } A \subset (\text{Id} \times \gamma \text{Id})(\text{dom } F_A) \Leftrightarrow \text{dom } (\gamma A) \times \text{ran } (\gamma A) \subset \text{dom } F_{\gamma A} \Leftrightarrow \gamma A$  is  $3^*$  monotone.  $\square$

**Example 24.16** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be monotone and let  $\gamma \in \mathbb{R}_{++}$ . Then  $J_{\gamma A}$  and  ${}^\gamma A$  are  $3^*$  monotone.

*Proof.* Since  $\text{Id} + \gamma A$  is strongly monotone, Example 24.11 implies that  $\text{Id} + \gamma A$  is  $3^*$  monotone. Therefore, by Proposition 24.15(i),  $J_{\gamma A} = (\text{Id} + \gamma A)^{-1}$  is  $3^*$  monotone. It follows from (23.14) that  $\text{Id} - J_{\gamma A} = J_{(\gamma A)^{-1}}$  is  $3^*$  monotone. In view of Proposition 24.15(ii),  $\gamma A = \gamma^{-1}(\text{Id} - J_{\gamma A})$  is also  $3^*$  monotone.  $\square$

**Proposition 24.17** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Then  $(\forall x \in \mathcal{H})(\forall u \in \mathcal{H}) F_{A+B}(x, u) \leq (F_A(x, \cdot) \square F_B(x, \cdot))(u)$ .*

*Proof.* Fix  $(x, u) \in \mathcal{H} \times \mathcal{H}$  and suppose that  $(u_1, u_2) \in \mathcal{H} \times \mathcal{H}$  satisfies  $u = u_1 + u_2$ . Take  $(y, v_1) \in \text{gra } A$  and  $(y, v_2) \in \text{gra } B$ . Then  $\langle x | v_1 + v_2 \rangle + \langle y | u \rangle - \langle y | v_1 + v_2 \rangle = (\langle x | v_1 \rangle + \langle y | u_1 \rangle - \langle y | v_1 \rangle) + (\langle x | v_2 \rangle + \langle y | u_2 \rangle - \langle y | v_2 \rangle) \leq F_A(x, u_1) + F_B(x, u_2)$  and hence  $F_{A+B}(x, u) \leq F_A(x, u_1) + F_B(x, u_2)$ . In turn, this implies that  $F_{A+B}(x, u) \leq \inf_{u_1+u_2=u} F_A(x, u_1) + F_B(x, u_2) = (F_A(x, \cdot) \square F_B(x, \cdot))(u)$ .  $\square$

**Proposition 24.18** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $(\text{dom } A \cap \text{dom } B) \times \mathcal{H} \subset \text{dom } F_B$ . Then  $A + B$  is  $3^*$  monotone.*

*Proof.* Let  $(x, v) \in \text{dom}(A + B) \times \mathcal{H} = (\text{dom } A \cap \text{dom } B) \times \mathcal{H}$  and take  $u \in Ax$ . By Proposition 20.47(i),  $F_A(x, u) = \langle x | u \rangle < +\infty$ . Thus, by Proposition 24.17,  $F_{A+B}(x, v) \leq (F_A(x, \cdot) \square F_B(x, \cdot))(v) \leq F_A(x, u) + F_B(x, v - u) < +\infty$ . Hence  $\text{dom}(A + B) \times \text{ran}(A + B) = (\text{dom } A \cap \text{dom } B) \times \mathcal{H} \subset \text{dom } F_{A+B}$ , and therefore  $A + B$  is  $3^*$  monotone.  $\square$

## 24.3 The Brézis–Haraux Theorem

**Theorem 24.19 (Simons)** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A + B$  is maximally monotone. Suppose that*

$$(\forall u \in \text{ran } A)(\forall v \in \text{ran } B)(\exists x \in \mathcal{H}) \\ (x, u) \in \text{dom } F_A \text{ and } (x, v) \in \text{dom } F_B. \quad (24.23)$$

*Then  $\overline{\text{ran}}(A + B) = \overline{\text{ran } A + \text{ran } B}$  and  $\text{int } \text{ran}(A + B) = \text{int}(\text{ran } A + \text{ran } B)$ .*

*Proof.* We set  $Q_2: \mathcal{H} \times \mathcal{H} \rightarrow \mathcal{H}: (z, w) \mapsto w$ . Fix  $u \in \text{ran } A$  and  $v \in \text{ran } B$ , and let  $x \in \mathcal{H}$  be such that  $(x, u) \in \text{dom } F_A$  and  $(x, v) \in \text{dom } F_B$ . Proposition 24.17 implies that

$$F_{A+B}(x, u + v) \leq F_A(x, u) + F_B(x, v) < +\infty, \quad (24.24)$$

and thus that  $u + v \in Q_2(\text{dom } F_{A+B})$ . Hence  $\text{ran } A + \text{ran } B \subset Q_2(\text{dom } F_{A+B})$  and therefore

$$\overline{\text{ran } A + \text{ran } B} \subset \overline{Q_2(\text{dom } F_{A+B})} \quad \text{and} \\ \text{int}(\text{ran } A + \text{ran } B) \subset \text{int } Q_2(\text{dom } F_{A+B}). \quad (24.25)$$

Since  $A + B$  is maximally monotone, Corollary 21.12 yields

$$\overline{Q_2(\operatorname{dom} F_{A+B})} = \overline{\operatorname{ran}}(A + B) \quad \text{and} \\ \operatorname{int} Q_2(\operatorname{dom} F_{A+B}) = \operatorname{int} \operatorname{ran}(A + B). \quad (24.26)$$

Altogether,

$$\overline{\operatorname{ran} A + \operatorname{ran} B} \subset \overline{\operatorname{ran}}(A + B) \quad \text{and} \\ \operatorname{int}(\operatorname{ran} A + \operatorname{ran} B) \subset \operatorname{int} \operatorname{ran}(A + B). \quad (24.27)$$

The results follow since the reverse inclusions in (24.27) always hold.  $\square$

**Theorem 24.20 (Brézis–Haraux)** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A+B$  is maximally monotone and one of the following conditions is satisfied:*

- (i)  $A$  and  $B$  are  $3^*$  monotone.
- (ii)  $\operatorname{dom} A \subset \operatorname{dom} B$  and  $B$  is  $3^*$  monotone.

*Then  $\overline{\operatorname{ran}}(A + B) = \overline{\operatorname{ran} A + \operatorname{ran} B}$  and  $\operatorname{int} \operatorname{ran}(A + B) = \operatorname{int}(\operatorname{ran} A + \operatorname{ran} B)$ .*

*Proof.* Let  $u \in \operatorname{ran} A$  and  $v \in \operatorname{ran} B$ . In view of Theorem 24.19, it suffices to show that there exists  $x \in \mathcal{H}$  such that  $(x, u) \in \operatorname{dom} F_A$  and  $(x, v) \in \operatorname{dom} F_B$ .

(i): For every  $x \in \operatorname{dom} A \cap \operatorname{dom} B$ ,  $(x, u) \in \operatorname{dom} A \times \operatorname{ran} A \subset \operatorname{dom} F_A$  and  $(x, v) \in \operatorname{dom} B \times \operatorname{ran} B \subset \operatorname{dom} F_B$ .

(ii): Since  $u \in \operatorname{ran} A$ , there exists  $x \in \operatorname{dom} A$  such that  $(x, u) \in \operatorname{gra} A$ . Proposition 20.47(i) yields  $(x, u) \in \operatorname{dom} F_A$ . Furthermore, since  $x \in \operatorname{dom} A$  and  $\operatorname{dom} A \subset \operatorname{dom} B$ , we have  $(x, v) \in \operatorname{dom} B \times \operatorname{ran} B \subset \operatorname{dom} F_B$ .  $\square$

**Example 24.21** It follows from Example 24.10 that, in Example 24.1,  $N_C$  and  $N_D$  are  $3^*$  monotone. However,  $\overline{\operatorname{ran}}(N_C + N_D) = \mathbb{R} \times \{0\} \neq \mathbb{R}^2 = \overline{\operatorname{ran} N_C + \operatorname{ran} N_D}$ . The assumption that  $A + B$  be maximally monotone in Theorem 24.20 is therefore critical.

**Corollary 24.22** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A + B$  is maximally monotone,  $A$  or  $B$  is surjective, and one of the following conditions is satisfied:*

- (i)  $A$  and  $B$  are  $3^*$  monotone.
- (ii)  $\operatorname{dom} A \subset \operatorname{dom} B$  and  $B$  is  $3^*$  monotone.

*Then  $A + B$  is surjective.*

**Corollary 24.23** *Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A + B$  is maximally monotone and  $B$  is uniformly monotone with a supercoercive modulus  $\phi$ . Suppose, in addition, that  $A$  is  $3^*$  monotone or that  $\operatorname{dom} A \subset \operatorname{dom} B$ . Then the following hold:*

- (i)  $\operatorname{ran}(A + B) = \mathcal{H}$ .



(ii)  $\text{zer}(A + B)$  is a singleton.

*Proof.* (i): Proposition 22.8(i) and Example 24.11 imply that  $B$  is surjective and  $3^*$  monotone. We then deduce from Corollary 24.22 that  $\text{ran}(A+B) = \mathcal{H}$ .

(ii): Since  $A$  is monotone and  $B$  is strictly monotone,  $A + B$  is strictly monotone. Hence, the inclusion  $0 \in Ax + Bx$  has at most one solution by Proposition 23.35. Existence follows from (i).  $\square$

## 24.4 Parallel Sum

**Definition 24.24** Let  $A$  and  $B$  be operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . The *parallel sum* of  $A$  and  $B$  is

$$A \square B = (A^{-1} + B^{-1})^{-1}. \quad (24.28)$$

Some elementary properties are collected in the next result.

**Proposition 24.25** Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Then the following hold:

- (i)  $(A \square B)x = \bigcup_{y \in \mathcal{H}} (Ay \cap B(x - y))$ .
- (ii)  $u \in (A \square B)x \Leftrightarrow (\exists y \in \mathcal{H}) \ y \in A^{-1}u \text{ and } x - y \in B^{-1}u$ .
- (iii)  $\text{dom}(A \square B) = \text{ran}(A^{-1} + B^{-1})$ .
- (iv)  $\text{ran}(A \square B) = \text{ran } A \cap \text{ran } B$ .
- (v) Suppose that  $A$  and  $B$  are monotone. Then  $A \square B$  is monotone.

**Proposition 24.26** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be at most single-valued and let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be linear. Then

$$A \square B = A(A + B)^{-1}B. \quad (24.29)$$

*Proof.* Let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Proposition 24.25(ii) yields

$$\begin{aligned} (x, u) \in \text{gra}(A \square B) &\Leftrightarrow (\exists y \in \mathcal{H}) \ y \in A^{-1}u \text{ and } x - y \in B^{-1}u \\ &\Leftrightarrow (\exists y \in \mathcal{H}) \ u = Ay = Bx - By \\ &\Leftrightarrow (\exists y \in \mathcal{H}) \ u = Ay \text{ and } (A + B)y = Bx \\ &\Leftrightarrow (\exists y \in \mathcal{H}) \ u = Ay \text{ and } y \in (A + B)^{-1}Bx \\ &\Leftrightarrow (x, u) \in \text{gra}(A(A + B)^{-1}B). \end{aligned} \quad (24.30)$$

Hence (24.29) holds.  $\square$

**Proposition 24.27** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{H})$ , and suppose that  $0 \in \text{sri}(\text{dom } f^* - \text{dom } g^*)$ . Then

$$(\partial f) \square (\partial g) = \partial(f \square g). \quad (24.31)$$

*Proof.* Using (24.28), Corollary 16.24, Corollary 16.38(i), Proposition 15.1, and Proposition 15.7(i), we obtain  $(\partial f) \square (\partial g) = ((\partial f)^{-1} + (\partial g)^{-1})^{-1} = (\partial f^* + \partial g^*)^{-1} = (\partial(f^* + g^*))^{-1} = \partial(f^* + g^*)^* = \partial(f \square g)^\vee = \partial(f \square g)$ .  $\square$

**Proposition 24.28** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A + B$  is maximally monotone. Then*

$$J_A \square J_B = J_{(1/2)(A+B)} \circ \frac{1}{2}\text{Id}. \quad (24.32)$$

*Proof.* Take  $(x, u) \in \text{gra}(J_A \square J_B)$ . By Proposition 24.25(i), there exists  $y \in \mathcal{H}$  such that  $u \in J_A y$  and  $u \in J_B(x - y)$ . Since  $J_A$  and  $J_B$  are single-valued, we have  $u = J_A y = J_B(x - y)$ . By Proposition 23.2(ii),  $(u, y - u) \in \text{gra } A$  and  $(u, x - y - u) \in \text{gra } B$ . Now let  $(z, w) \in \text{gra}(A + B)$ . Then there exist  $w_A \in Az$  and  $w_B \in Bz$  such that  $w = w_A + w_B$ . By monotonicity of  $A$  and  $B$ , we have  $\langle u - z \mid y - u - w_A \rangle \geq 0$  and  $\langle u - z \mid x - y - u - w_B \rangle \geq 0$ . Adding the last two inequalities yields  $\langle u - z \mid x - 2u - w \rangle \geq 0$ . Since  $A + B$  is maximally monotone, we deduce that  $(u, x - 2u) = (u, 2(x/2 - u)) \in \text{gra}(A + B)$ . In turn, by Proposition 23.2(ii),  $u = J_{(1/2)(A+B)}(x/2)$ .  $\square$

**Corollary 24.29** *Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $A + B$  is maximally monotone. Then  $J_{A+B} = (J_A \square J_B) \circ 2\text{Id}$ .*

**Corollary 24.30** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $\text{dom } f \cap \text{dom } g \neq \emptyset$  and  $\partial f + \partial g = \partial(f + g)$ . Then*

$$\text{Prox}_{f+g} = (\text{Prox}_f \square \text{Prox}_g) \circ 2\text{Id}. \quad (24.33)$$

**Corollary 24.31** *Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$  such that  $C \cap D \neq \emptyset$  and  $N_C + N_D = N_{C \cap D}$ . Then*

$$P_{C \cap D} = (P_C \square P_D) \circ 2\text{Id}. \quad (24.34)$$

**Proposition 24.32** *Let  $C$  and  $D$  be closed linear subspaces of  $\mathcal{H}$  such that  $C + D$  is closed. Then the following hold:*

- (i)  $P_{C \cap D} = 2P_C(P_C + P_D)^{-1}P_D$ .
- (ii) Suppose that  $\text{ran}(P_C + P_D)$  is closed. Then  $P_{C \cap D} = 2P_C(P_C + P_D)^\dagger P_D$ .

*Proof.* Since  $C + D$  is closed, it follows from Corollary 15.35 that  $C^\perp + D^\perp = C^\perp + D^\perp = (C \cap D)^\perp$ . Hence,  $N_C + N_D = N_{C \cap D}$ .

(i): Using Corollary 24.31, the linearity of  $P_{C \cap D}$ , and Proposition 24.26, we see that  $P_{C \cap D} = 2(P_C \square P_D) = 2P_D(P_C + P_D)^{-1}P_D$ .

(ii): Since, by Corollary 3.22 and Corollary 20.25 the projectors  $P_C$  and  $P_D$  are continuous, maximally monotone, and self-adjoint, so is their sum  $P_C + P_D$ . Furthermore, by Proposition 4.8 and Proposition 24.12,  $P_C$  and  $P_D$  are  $3^*$  monotone. Theorem 24.20 therefore yields

$$\text{ran}(P_C + P_D) = \overline{\text{ran}}(P_C + P_D) = \overline{\text{ran } P_C + \text{ran } P_D} = \overline{C + D} = C + D. \quad (24.35)$$

In turn, Exercise 3.13 implies that

$$\begin{aligned}(P_C + P_D)^\dagger &= P_{\text{ran}(P_C + P_D)^*} \circ (P_C + P_D)^{-1} \circ P_{\text{ran}(P_C + P_D)} \\ &= P_{C+D}(P_C + P_D)^{-1}P_{C+D}.\end{aligned}\quad (24.36)$$

On the other hand, by Exercise 4.9, we have  $P_C P_{C+D} = P_C$  and  $P_{C+D} P_D = P_D$ . Altogether,

$$\begin{aligned}P_C(P_C + P_D)^\dagger P_D &= P_C(P_{C+D}(P_C + P_D)^{-1}P_{C+D})P_D \\ &= (P_C P_{C+D})(P_C + P_D)^{-1}(P_{C+D} P_D) \\ &= P_C(P_C + P_D)^{-1}P_D.\end{aligned}\quad (24.37)$$

In view of (i), the proof is therefore complete.  $\square$

**Corollary 24.33 (Anderson–Duffin)** *Suppose that  $\mathcal{H}$  has finite dimension, and let  $C$  and  $D$  be linear subspaces of  $\mathcal{H}$ . Then*

$$P_{C \cap D} = 2P_C(P_C + P_D)^\dagger P_D. \quad (24.38)$$

## Exercises

**Exercise 24.1** Show that Theorem 24.5 fails if  $\text{cone}(\text{ran } L - \text{dom } A) \neq \overline{\text{span}}(\text{ran } L - \text{dom } A)$ .

**Exercise 24.2** Is the upper bound for  $F_{A+B}$  provided in Proposition 24.17 sharp when  $A$  and  $B$  are operators in  $\mathcal{B}(\mathcal{H})$  such that  $A^* = -A$  and  $B^* = -B$ ?

**Exercise 24.3** Is the upper bound for  $F_{A+B}$  provided in Proposition 24.17 sharp when  $A$  and  $B$  are operators in  $\mathcal{B}(\mathcal{H})$  such that  $A$  is self-adjoint and monotone, and  $B^* = -B$ ?

**Exercise 24.4** Is the upper bound for  $F_{A+B}$  provided in Proposition 24.17 sharp when  $V$  is a closed linear subspace of  $\mathcal{H}$ ,  $A = P_V$ , and  $B = P_{V^\perp}$ ?

**Exercise 24.5** Is the upper bound for  $F_{A+B}$  provided in Proposition 24.17 sharp when  $\mathcal{H} = \mathbb{R}$ ,  $A = P_K$ , and  $B = P_{K^\perp}$ , where  $K = \mathbb{R}_+$ ?

**Exercise 24.6** An operator  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  is *angle bounded* with constant  $\beta \in \mathbb{R}_{++}$  if, for all  $(x, u)$ ,  $(y, v)$ , and  $(z, w)$  in  $\mathcal{H} \times \mathcal{H}$ ,

$$\left. \begin{aligned} (x, u) &\in \text{gra } A, \\ (y, v) &\in \text{gra } A, \\ (z, w) &\in \text{gra } A \end{aligned} \right\} \Rightarrow \langle y - z \mid w - u \rangle \leq \beta \langle x - y \mid u - v \rangle. \quad (24.39)$$

Let  $f \in \Gamma_0(\mathcal{H})$ . Show that  $\partial f$  is angle bounded with constant 1.

**Exercise 24.7** Show that every angle bounded monotone operator is  $3^*$  monotone.

**Exercise 24.8** Provide an example of a maximally monotone operator that is not  $3^*$  monotone.

**Exercise 24.9** Provide an example of maximally monotone operators  $A$  and  $B$  such that  $A + B$  is maximally monotone,  $\text{ran}(A + B) = \{0\}$ , and  $\text{ran } A = \text{ran } B = \mathcal{H}$ . Conclude that the assumptions (i) and (ii) of Theorem 24.20 are critical.

**Exercise 24.10** Let  $R$  be as in Example 24.14. Prove that

$$\ker(\text{Id} - R) = \{(x, \dots, x) \in \mathcal{H}^N \mid x \in \mathcal{H}\} \quad (24.40)$$

and that

$$\text{ran}(\text{Id} - R) = (\ker(\text{Id} - R))^\perp = \left\{ (x_1, \dots, x_N) \in \mathcal{H}^N \mid \sum_{i=1}^N x_i = 0 \right\}. \quad (24.41)$$

**Exercise 24.11** Let  $A$  and  $B$  be monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $\text{dom}(A \square B) \neq \emptyset$ . Let  $(x, u_1)$  and  $(x, u_2)$  be in  $\text{gra}(A \square B)$ , and let  $y_1$  and  $y_2$  be such that  $u_1 \in Ay_1 \cap B(x - y_1)$  and  $u_2 \in Ay_2 \cap B(x - y_2)$ . Show that  $\langle y_1 - y_2 \mid u_1 - u_2 \rangle = 0$ .

**Exercise 24.12** Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, and let  $\gamma \in \mathbb{R}_{++}$ . Show that  $A \square (\gamma^{-1} \text{Id}) = \gamma A$ .

**Exercise 24.13** Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Use Lemma 2.13(i) and Exercise 24.11 to show that, for every  $x \in \mathcal{H}$ ,  $(A \square B)x$  is convex.

**Exercise 24.14** Let  $A$  and  $B$  be strictly positive self-adjoint surjective operators in  $\mathcal{B}(\mathcal{H})$ . Set  $q_A: x \mapsto (1/2) \langle x \mid Ax \rangle$ . Show that  $A + B$  is surjective and observe that the surjectivity assumption in Exercise 12.12 is therefore superfluous. Furthermore, use Exercise 12.12 to deduce that  $q_A \square q_B = q_A \square q_B$ .

**Exercise 24.15** Let  $A \in \mathcal{B}(\mathcal{H})$  be positive and self-adjoint, and set  $q_A: x \mapsto (1/2) \langle x \mid Ax \rangle$ . Show that  $q_A$  is convex and Fréchet differentiable, with  $\nabla q_A = A$  and  $q_A^* \circ A = q_A$ . Moreover, prove that  $\text{ran } A \subset \text{dom } q_A^* \subset \overline{\text{ran } A}$ .

**Exercise 24.16** Let  $A \in \mathcal{B}(\mathcal{H})$  be positive and self-adjoint, and let  $B \in \mathcal{B}(\mathcal{H})$  be such that  $B^* = -B$ . Suppose that  $\text{ran } A$  is closed and set  $C = A + B$ . Show that  $C$  is  $3^*$  monotone if and only if  $\text{ran } B \subset \text{ran } A$ .

# Chapter 25

## Zeros of Sums of Monotone Operators

Properties of the zeros of a single monotone operator were discussed in Section 23.4. In this chapter, we first characterize the zeros of sums of monotone operators and then present basic algorithms to construct such zeros iteratively. These are called splitting algorithms in the sense that they involve the operators individually.

### 25.1 Characterizations

The solutions to problems in various areas of nonlinear analysis can be modeled as the zeros of the sum of two monotone operators. These zeros can be characterized as follows, where we use the notation (23.10).

**Proposition 25.1** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ , let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Then the following hold:*

- (i)  $\text{zer}(A + B) = \text{dom}(A \cap (-B))$ .
- (ii) *Suppose that  $A$  and  $B$  are monotone. Then*

$$\text{zer}(A + B) = J_{\gamma B}(\text{Fix } R_{\gamma A} R_{\gamma B}). \quad (25.1)$$

- (iii) *Let  $C$  be a closed affine subspace of  $\mathcal{H}$ , set  $V = C - C$ , and suppose that  $A = N_C$ . Then  $\text{zer}(A + B) = \{x \in C \mid V^{\perp} \cap Bx \neq \emptyset\}$ .*
- (iv) *Suppose that  $A$  is monotone and that  $B$  is at most single-valued. Then*

$$\text{zer}(A + B) = \text{Fix } J_{\gamma A} \circ (\text{Id} - \gamma B). \quad (25.2)$$

*Proof.* Let  $x \in \mathcal{H}$ .

- (i):  $0 \in Ax + Bx \Leftrightarrow (\exists u \in Bx) -u \in Ax \Leftrightarrow Ax \cap (-Bx) \neq \emptyset$ .
- (ii): We have

$$0 \in Ax + Bx \Leftrightarrow (\exists y \in \mathcal{H}) \quad x - y \in \gamma Ax \quad \text{and} \quad y - x \in \gamma Bx$$

$$\begin{aligned}
&\Leftrightarrow (\exists y \in \mathcal{H}) \quad 2x - y \in (\text{Id} + \gamma A)x \quad \text{and} \quad x = J_{\gamma B}y \\
&\Leftrightarrow (\exists y \in \mathcal{H}) \quad x = J_{\gamma A}(R_{\gamma B}y) \quad \text{and} \quad x = J_{\gamma B}y \\
&\Leftrightarrow (\exists y \in \mathcal{H}) \quad y = 2x - R_{\gamma B}y = R_{\gamma A}(R_{\gamma B}y) \quad \text{and} \quad x = J_{\gamma B}y \\
&\Leftrightarrow (\exists y \in \text{Fix } R_{\gamma A}R_{\gamma B}) \quad x = J_{\gamma B}y.
\end{aligned} \tag{25.3}$$

(iii): It follows from Example 6.42 that  $x \in \text{zer}(N_C + B) \Leftrightarrow 0 \in N_Cx + Bx \Leftrightarrow (\exists u \in Bx) -u \in N_Cx \Leftrightarrow [x \in C \text{ and } (\exists u \in Bx) u \in V^\perp]$ .

(iv): We have  $0 \in Ax + Bx \Leftrightarrow -Bx \in Ax \Leftrightarrow x - \gamma Bx \in x + \gamma Ax \Leftrightarrow x \in ((\text{Id} + \gamma A)^{-1} \circ (\text{Id} - \gamma B))x \Leftrightarrow x = (J_{\gamma A} \circ (\text{Id} - \gamma B))x$ .  $\square$

**Proposition 25.2** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $B: \mathcal{H} \rightarrow \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Then  $\text{Fix } J_A \circ (\text{Id} + \gamma(B - \text{Id})) = \text{Fix } J_{\gamma^{-1}A} \circ B$ .*

*Proof.* For every  $x \in \mathcal{H}$ , we have  $x \in \text{Fix } J_{\gamma^{-1}A} \circ B \Leftrightarrow \gamma(Bx - x) \in Ax \Leftrightarrow (\gamma B + (1 - \gamma)\text{Id})x \in (\text{Id} + A)x \Leftrightarrow x = (\text{Id} + A)^{-1}(\gamma B + (1 - \gamma)\text{Id})x \Leftrightarrow x \in \text{Fix } J_A(\text{Id} + \gamma(B - \text{Id}))$ .  $\square$

The following asymptotic result complements Proposition 25.1(iii).

**Proposition 25.3** *Let  $V$  be a closed linear subspace of  $\mathcal{H}$ , let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $(x_n, u_n)_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } B$ , and let  $(x, u) \in \mathcal{H} \times \mathcal{H}$ . Suppose that*

$$x_n \rightharpoonup x, \quad u_n \rightharpoonup u, \quad P_{V^\perp}x_n \rightarrow 0, \quad \text{and} \quad P_V u_n \rightarrow 0. \tag{25.4}$$

*Then the following hold:*

- (i)  $x \in \text{zer}(N_V + B)$ .
- (ii)  $(x, u) \in (V \times V^\perp) \cap \text{gra } B$ .
- (iii)  $\langle x_n \mid u_n \rangle \rightarrow \langle x \mid u \rangle = 0$ .

*Proof.* (ii)&(iii): This follows from Proposition 20.50.

(i): Combine (ii) and Proposition 25.1(iii).  $\square$

In order to study the zeros of an arbitrary finite sum of maximally monotone operators, we need a few technical facts.

**Proposition 25.4** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , and let  $(A_i)_{i \in I}$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Furthermore, set*

$$\begin{cases} \mathcal{H} = \bigoplus_{i \in I} \mathcal{H}, \\ \mathcal{D} = \{(x, \dots, x) \in \mathcal{H} \mid x \in \mathcal{H}\}, \\ j: \mathcal{H} \rightarrow \mathcal{D}: x \mapsto (x, \dots, x), \\ \mathcal{B} = \bigtimes_{i \in I} A_i. \end{cases} \tag{25.5}$$

*Then the following hold for every  $\mathbf{x} = (x_i)_{i \in I} \in \mathcal{H}$ :*

- (i)  $D^\perp = \{u \in \mathcal{H} \mid \sum_{i \in I} u_i = 0\}$ .
- (ii)  $N_D x = \begin{cases} \{u \in \mathcal{H} \mid \sum_{i \in I} u_i = 0\}, & \text{if } x \in D; \\ \emptyset, & \text{otherwise.} \end{cases}$
- (iii)  $P_D x = j((1/m) \sum_{i \in I} x_i)$ .
- (iv)  $P_{D^\perp} x = (x_i - (1/m) \sum_{j \in I} x_j)_{i \in I}$ .
- (v)  $J_\gamma B x = (J_\gamma A_i x_i)_{i \in I}$ .
- (vi)  $j(\text{zer} \sum_{i \in I} A_i) = \text{zer} (N_D + B)$ .

*Proof.* (i): This follows from (2.6).

(ii): Combine (i) and Example 6.42.

(iii): Set  $p = (1/m) \sum_{i \in I} x_i$  and  $\mathbf{p} = j(p)$ , and let  $\mathbf{y} = j(y)$ , where  $y \in \mathcal{H}$ . Then  $\mathbf{p} \in D$ ,  $\mathbf{y}$  is an arbitrary point in  $D$ , and  $\langle x - \mathbf{p} \mid \mathbf{y} \rangle = \sum_{i \in I} \langle x_i - p \mid y \rangle = \langle \sum_{i \in I} x_i - mp \mid y \rangle = 0$ . Hence, by Corollary 3.22(i),  $\mathbf{p} = P_D x$ .

(iv): Corollary 3.22(v).

(v): Proposition 23.16.

(vi): Let  $x \in \mathcal{H}$ . Then (ii) implies that

$$\begin{aligned}
 0 \in \sum_{i \in I} A_i x &\Leftrightarrow \left( \exists (u_i)_{i \in I} \in \prod_{i \in I} A_i x \right) \sum_{i \in I} u_i = 0 \\
 &\Leftrightarrow (\exists \mathbf{u} \in B j(x)) \quad -\mathbf{u} \in D^\perp = N_D j(x) \\
 &\Leftrightarrow j(x) \in \text{zer}(N_D + B) \subset D,
 \end{aligned} \tag{25.6}$$

and we have obtained the announced identity.  $\square$

**Corollary 25.5** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , and let  $(A_i)_{i \in I}$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . For every  $i \in I$ , let  $(x_{i,n}, u_{i,n})_{n \in \mathbb{N}}$  be a sequence in  $\text{gra } A_i$  and let  $(x_i, u_i) \in \mathcal{H} \times \mathcal{H}$ . Suppose that*

$$\sum_{i \in I} u_{i,n} \rightarrow 0 \quad \text{and} \quad (\forall i \in I) \quad \begin{cases} x_{i,n} \rightharpoonup x_i, \\ u_{i,n} \rightharpoonup u_i, \\ mx_{i,n} - \sum_{j \in I} x_{j,n} \rightarrow 0. \end{cases} \tag{25.7}$$

*Then there exists  $x \in \text{zer} \sum_{i \in I} A_i$  such that the following hold:*

- (i)  $x = x_1 = \dots = x_m$ .
- (ii)  $\sum_{i \in I} u_i = 0$ .

- (iii)  $(\forall i \in I) (x, u_i) \in \text{gra } A_i$ .
- (iv)  $\sum_{i \in I} \langle x_{i,n} \mid u_{i,n} \rangle \rightarrow \langle x \mid \sum_{i \in I} u_i \rangle = 0$ .

*Proof.* Define  $\mathcal{H}$ ,  $D$ , and  $B$  as in Proposition 25.4, set  $\mathbf{x} = (x_i)_{i \in I}$  and  $\mathbf{u} = (u_i)_{i \in I}$ , and note that, by Proposition 23.16,  $B$  is maximally monotone. Now set  $(\forall n \in \mathbb{N}) \mathbf{x}_n = (x_{i,n})_{i \in I}$  and  $\mathbf{u}_n = (u_{i,n})_{i \in I}$ . Then  $(\mathbf{x}_n, \mathbf{u}_n)_{n \in \mathbb{N}}$  lies in  $\text{gra } B$  and we derive from (25.7), Proposition 25.4(iii), and Proposition 25.4(iv), that  $\mathbf{x}_n \rightharpoonup \mathbf{x}$ ,  $\mathbf{u}_n \rightharpoonup \mathbf{u}$ ,  $P_{D^\perp} \mathbf{x}_n \rightarrow 0$ , and  $P_D \mathbf{u}_n \rightarrow 0$ . The assertions therefore follow from Proposition 25.3 and Proposition 25.4(i)&(vi).  $\square$

## 25.2 Douglas–Rachford Splitting

We now turn our attention to the central problem of finding a zero of the sum of two maximally monotone operators  $A$  and  $B$ , i.e.,

$$\text{find } x \in \mathcal{H} \text{ such that } 0 \in Ax + Bx. \quad (25.8)$$

When  $A + B$  is maximally monotone, one can approach this problem via Theorem 23.41. Naturally, this approach is numerically viable only in those cases in which it is easy to compute  $J_{\gamma(A+B)}$ . A more widely applicable alternative is to devise an *operator splitting* algorithm, in which the operators  $A$  and  $B$  are employed in separate steps.

The following algorithm is referred to as the *Douglas–Rachford* algorithm because, in a special linear case, it is akin to a method proposed by Douglas and Rachford for solving certain matrix equations. Its main step consists in alternating the reflected resolvents defined in (23.10).

**Theorem 25.6 (Douglas–Rachford algorithm)** *Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $\text{zer}(A + B) \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = J_{\gamma B} x_n, \\ z_n = J_{\gamma A}(2y_n - x_n), \\ x_{n+1} = x_n + \lambda_n(z_n - y_n). \end{cases} \quad (25.9)$$

*Then there exists  $x \in \text{Fix } R_{\gamma A} R_{\gamma B}$  such that the following hold:*

- (i)  $J_{\gamma B} x \in \text{zer}(A + B)$ .
- (ii)  $(y_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (iii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to  $x$ .
- (iv)  $(y_n)_{n \in \mathbb{N}}$  converges weakly to  $J_{\gamma B} x$ .
- (v)  $(z_n)_{n \in \mathbb{N}}$  converges weakly to  $J_{\gamma B} x$ .
- (vi) *Suppose that  $A = N_C$ , where  $C$  is a closed affine subspace of  $\mathcal{H}$ . Then  $(P_C x_n)_{n \in \mathbb{N}}$  converges weakly to  $J_{\gamma B} x$ .*



(vii) Suppose that one of the following holds (see Remark 22.2):

- (a)  $A$  is uniformly monotone on every nonempty bounded subset of  $\text{dom } A$ .
- (b)  $B$  is uniformly monotone on every nonempty bounded subset of  $\text{dom } B$ .

Then  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique point in  $\text{zer}(A + B)$ .

*Proof.* Set  $T = R_{\gamma A} R_{\gamma B}$ . Since  $R_{\gamma A}$  and  $R_{\gamma B}$  are nonexpansive by Corollary 23.10(ii),  $T$  is nonexpansive. Moreover, since Proposition 25.1(ii) states that  $J_{\gamma B}(\text{Fix } T) = \text{zer}(A + B)$ , which is nonempty, we have  $\text{Fix } T \neq \emptyset$ . We also note that, using Proposition 4.21(i), we can rewrite (25.9) as

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \mu_n(Tx_n - x_n), \quad \text{where} \quad \mu_n = \frac{\lambda_n}{2}. \quad (25.10)$$

It will be convenient to set

$$(\forall n \in \mathbb{N}) \quad v_n = x_n - y_n \quad \text{and} \quad w_n = 2y_n - x_n - z_n \quad (25.11)$$

and to observe that (25.9) yields

$$(\forall n \in \mathbb{N}) \quad \begin{cases} (z_n, w_n) \in \text{gra } \gamma A, \\ (y_n, v_n) \in \text{gra } \gamma B, \\ v_n + w_n = y_n - z_n. \end{cases} \quad (25.12)$$

(i): Proposition 25.1(ii).

(ii): This follows from (25.10) and Theorem 5.14(ii).

(iii): This follows from (25.10) and Theorem 5.14(iii).

(iv): By Corollary 23.10(i),  $(\forall n \in \mathbb{N}) \quad \|y_n - y_0\| = \|J_{\gamma B}x_n - J_{\gamma B}x_0\| \leq \|x_n - x_0\|$ . Hence, since  $(x_n)_{n \in \mathbb{N}}$  is bounded by (iii), so is  $(y_n)_{n \in \mathbb{N}}$ . Now let  $y$  be a weak sequential cluster point of  $(y_n)_{n \in \mathbb{N}}$ , say  $y_{k_n} \rightharpoonup y$ . It follows from (25.11), (ii), and (iii) that

$$y_{k_n} \rightharpoonup y, \quad z_{k_n} \rightharpoonup y, \quad v_{k_n} \rightharpoonup x - y, \quad \text{and} \quad w_{k_n} \rightharpoonup y - x. \quad (25.13)$$

In turn, (25.12), (ii), and Corollary 25.5 yield  $y \in \text{zer}(\gamma A + \gamma B) = \text{zer}(A + B)$ ,  $(y, y - x) \in \text{gra } \gamma A$ , and  $(y, x - y) \in \text{gra } \gamma B$ . Hence,

$$y = J_{\gamma B}x \quad \text{and} \quad y \in \text{dom } A. \quad (25.14)$$

Thus,  $J_{\gamma B}x$  is the unique weak sequential cluster point of  $(y_n)_{n \in \mathbb{N}}$  and, appealing to Lemma 2.38, we conclude that  $y_n \rightharpoonup J_{\gamma B}x$ .

(v): Combine (ii) and (iv).

(vi): On the one hand, (iii), Proposition 4.11(i), and Example 23.4 yield  $P_Cx_n \rightharpoonup P_Cx = J_{\gamma A}x$ . On the other hand, by Proposition 4.21(iii)&(iv),

$x \in \text{Fix} T \Leftrightarrow x \in \text{Fix}(P_C R_{\gamma B} + \text{Id} - J_{\gamma B}) \Leftrightarrow P_C x = J_{\gamma B} x$ . Altogether,  $P_C x_n \rightharpoonup J_{\gamma B} x$ .

(vii): Set  $y = J_{\gamma B} x$ . As seen in (i) and (iii),  $y \in \text{zer}(A + B)$  and  $x_n \rightharpoonup x$ . Since  $A$  or  $B$  is strictly monotone, so is  $A + B$ . Hence Proposition 23.35 yields  $\text{zer}(A + B) = \{y\}$ . It follows from (25.11), (iii), (iv), and (v) that

$$y_n \rightharpoonup y, \quad z_n \rightharpoonup y, \quad v_n \rightharpoonup x - y, \quad \text{and} \quad w_n \rightharpoonup y - x. \quad (25.15)$$

We therefore deduce from (25.12), (25.15), (ii), and Corollary 25.5 that

$$(y, y - x) \in \text{gra } \gamma A \quad \text{and} \quad (y, x - y) \in \text{gra } \gamma B. \quad (25.16)$$

(vii)(a): Set  $C = \{y\} \cup \{z_n\}_{n \in \mathbb{N}}$ . Since (v) and (25.12) imply that  $(z_n)_{n \in \mathbb{N}}$  is a bounded sequence in  $\text{dom } A$ , and since  $y \in \text{dom } A$  by (25.14),  $C$  is a bounded subset of  $\text{dom } A$ . Hence, it follows from (25.12), (25.16), and (22.5) that there exists an increasing function  $\phi_A : \mathbb{R}_+ \rightarrow [0, +\infty]$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \gamma \phi_A(\|z_n - y\|) \leq \langle z_n - y \mid w_n - y + x \rangle. \quad (25.17)$$

On the other hand, by (25.12), (25.16), and the monotonicity of  $\gamma B$ ,

$$(\forall n \in \mathbb{N}) \quad 0 \leq \langle y_n - y \mid v_n - x + y \rangle. \quad (25.18)$$

Thus,

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad \gamma \phi_A(\|z_n - y\|) &\leq \langle z_n - y \mid w_n - y + x \rangle + \langle y_n - y \mid v_n - x + y \rangle \\ &= \langle z_n - y_n \mid w_n - y + x \rangle + \langle y_n - y \mid w_n - y + x \rangle \\ &\quad + \langle y_n - y \mid v_n - x + y \rangle \\ &= \langle z_n - y_n \mid w_n - y + x \rangle + \langle y_n - y \mid w_n + v_n \rangle \\ &= \langle z_n - y_n \mid w_n - y + x \rangle + \langle y_n - y \mid y_n - z_n \rangle \\ &= \langle z_n - y_n \mid w_n - y_n + x \rangle. \end{aligned} \quad (25.19)$$

However, it follows from (ii) that  $z_n - y_n \rightarrow 0$  and from (25.15) that  $w_n - y_n + x \rightharpoonup 0$ . Thus, it follows from Lemma 2.41(iii) that  $\phi_A(\|z_n - y\|) \rightarrow 0$  and therefore  $z_n \rightarrow y$ . In turn,  $y_n \rightarrow y$ .

(vii)(b): We argue as in (vii)(a), except that the roles of  $A$  and  $B$  are interchanged. Thus, (25.19) becomes

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad \gamma \phi_B(\|y_n - y\|) &\leq \langle y_n - y \mid v_n - x + y \rangle + \langle z_n - y \mid w_n - y + x \rangle \\ &= \langle y_n - z_n \mid v_n - x + y \rangle + \langle z_n - y \mid v_n - x + y \rangle \\ &\quad + \langle z_n - y \mid w_n - y + x \rangle \\ &= \langle y_n - z_n \mid v_n - x + y \rangle + \langle z_n - y \mid v_n + w_n \rangle \\ &= \langle y_n - z_n \mid v_n + z_n - x \rangle. \end{aligned} \quad (25.20)$$

Consequently, since  $y_n - z_n \rightarrow 0$  and  $v_n + z_n - x \rightarrow 0$ , we get  $\phi_B(\|y_n - y\|) \rightarrow 0$ , hence  $y_n \rightarrow y$  and  $z_n \rightarrow y$ .  $\square$

By recasting the Douglas–Rachford algorithm (25.9) in a product space, we obtain a parallel splitting algorithm for finding a zero of a finite sum of maximally monotone operators.

**Proposition 25.7 (parallel splitting algorithm)** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , let  $(A_i)_{i \in I}$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$  such that  $\text{zer} \sum_{i \in I} A_i \neq \emptyset$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $(x_{i,0})_{i \in I} \in \mathcal{H}^m$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} p_n = \frac{1}{m} \sum_{i \in I} x_{i,n}, \\ (\forall i \in I) \quad y_{i,n} = J_{\gamma A_i} x_{i,n}, \\ q_n = \frac{1}{m} \sum_{i \in I} y_{i,n}, \\ (\forall i \in I) \quad x_{i,n+1} = x_{i,n} + \lambda_n(2q_n - p_n - y_{i,n}). \end{cases} \quad (25.21)$$

Then  $(p_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer} \sum_{i \in I} A_i$ .

*Proof.* Define  $\mathcal{H}$ ,  $\mathbf{D}$ ,  $\mathbf{j}$ , and  $\mathbf{B}$  as in (25.5), and set  $\mathbf{A} = N_{\mathbf{D}}$ . It follows from Example 20.41, Example 23.4, Proposition 23.16, and Proposition 25.4(iii)&(v) that  $\mathbf{A}$  and  $\mathbf{B}$  are maximally monotone, with

$$(\forall \mathbf{x} \in \mathcal{H}) \quad J_{\gamma \mathbf{A}} \mathbf{x} = \mathbf{j} \left( \frac{1}{m} \sum_{i \in I} x_i \right) \quad \text{and} \quad J_{\gamma \mathbf{B}} \mathbf{x} = (J_{\gamma A_i} x_i)_{i \in I}. \quad (25.22)$$

Moreover, Proposition 25.4(vi) yields

$$\mathbf{j} \left( \text{zer} \sum_{i \in I} A_i \right) = \text{zer} (\mathbf{A} + \mathbf{B}). \quad (25.23)$$

Now set  $(\forall n \in \mathbb{N}) \quad \mathbf{x}_n = (x_{i,n})_{i \in I}$ ,  $\mathbf{p}_n = \mathbf{j}(p_n)$ ,  $\mathbf{y}_n = (y_{i,n})_{i \in I}$ , and  $\mathbf{q}_n = \mathbf{j}(q_n)$ . Then it follows from (25.22) that (25.21) can be rewritten as

$$(\forall n \in \mathbb{N}) \quad \begin{cases} \mathbf{p}_n = P_{\mathbf{D}} \mathbf{x}_n, \\ \mathbf{y}_n = J_{\gamma \mathbf{B}} \mathbf{x}_n, \\ \mathbf{q}_n = P_{\mathbf{D}} \mathbf{y}_n, \\ \mathbf{x}_{n+1} = \mathbf{x}_n + \lambda_n(2\mathbf{q}_n - \mathbf{p}_n - \mathbf{y}_n). \end{cases} \quad (25.24)$$

In turn, since  $J_{\gamma \mathbf{A}} = P_{\mathbf{D}}$  is linear, this is equivalent to

$$(\forall n \in \mathbb{N}) \quad \begin{cases} \mathbf{p}_n = J_{\gamma \mathbf{A}} \mathbf{x}_n, \\ \mathbf{y}_n = J_{\gamma \mathbf{B}} \mathbf{x}_n, \\ \mathbf{x}_{n+1} = \mathbf{x}_n + \lambda_n(J_{\gamma \mathbf{A}}(2\mathbf{y}_n - \mathbf{x}_n) - \mathbf{y}_n). \end{cases} \quad (25.25)$$

Hence, we derive from Theorem 25.6(vi) and (25.23) that  $(p_n)_{n \in \mathbb{N}}$  converges weakly to a point  $j(x)$ , where  $x \in \text{zer} \sum_{i \in I} A_i$ . Thus,  $p_n = j^{-1}(p_n) \rightharpoonup x$ .  $\square$

### 25.3 Forward–Backward Splitting

We focus on the case when  $B$  is single-valued in (25.8). The algorithm described next is known as a *forward–backward* algorithm. It alternates an explicit step using the operator  $B$  with an implicit resolvent step involving the operator  $A$ , i.e., in the language of numerical analysis, a forward step with a backward step.

**Theorem 25.8 (forward–backward algorithm)** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\beta \in \mathbb{R}_{++}$ , let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive, let  $\gamma \in ]0, 2\beta[$ , and set  $\delta = \min\{1, \beta/\gamma\} + 1/2$ . Furthermore, let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, \delta]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(\delta - \lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Suppose that  $\text{zer}(A + B) \neq \emptyset$  and set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ x_{n+1} = x_n + \lambda_n(J_{\gamma A}y_n - x_n). \end{cases} \quad (25.26)$$

*Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer}(A + B)$ .
- (ii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and let  $x \in \text{zer}(A + B)$ . Then  $(Bx_n)_{n \in \mathbb{N}}$  converges strongly to  $Bx$ .
- (iii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and that one of the following holds:
  - (a)  $A$  is uniformly monotone on every nonempty bounded subset of  $\text{dom } A$ .
  - (b)  $B$  is uniformly monotone on every nonempty bounded subset of  $\mathcal{H}$ .

*Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique point in  $\text{zer}(A + B)$ .*

*Proof.* Set  $T = J_{\gamma A} \circ (\text{Id} - \gamma B)$  and  $\alpha = 1/\delta$ . On the one hand, Corollary 23.8 and Remark 4.24(iii) imply that  $J_{\gamma A}$  is  $1/2$ -averaged. On the other hand, Proposition 4.33 implies that  $\text{Id} - \gamma B$  is  $\gamma/(2\beta)$ -averaged. Hence, Proposition 4.32 implies that  $T$  is  $\alpha$ -averaged. Moreover, by Proposition 25.1(iv),  $\text{Fix } T = \text{zer}(A + B)$ . We also deduce from (25.26) that  $(x_n)_{n \in \mathbb{N}}$  is generated by (5.15).

- (i): The claim follows from Proposition 5.15(iii).
- (ii): Let  $x \in \text{zer}(A + B) = \text{Fix } T$  and set  $\varepsilon = \inf_{n \in \mathbb{N}} \lambda_n$ . We derive from the above and Proposition 4.25(iii) that, for every  $n \in \mathbb{N}$ ,

$$\begin{aligned} \|Tx_n - x\|^2 &= \|J_{\gamma A}(x_n - \gamma Bx_n) - J_{\gamma A}(x - \gamma Bx)\|^2 \\ &\leq \|(\text{Id} - \gamma B)x_n - (\text{Id} - \gamma B)x\|^2 \end{aligned}$$

$$\leq \|x_n - x\|^2 - \gamma(2\beta - \gamma)\|Bx_n - Bx\|^2 \quad (25.27)$$

and, therefore, using Corollary 2.14 and the inequality  $\delta > 1$ , that

$$\begin{aligned} \|x_{n+1} - x\|^2 &= \|(1 - \lambda_n)(x_n - x) + \lambda_n(Tx_n - x)\|^2 \\ &= (1 - \lambda_n)\|x_n - x\|^2 + \lambda_n\|Tx_n - x\|^2 \\ &\quad - \lambda_n(1 - \lambda_n)\|Tx_n - x_n\|^2 \\ &\leq \|x_n - x\|^2 - \gamma(2\beta - \gamma)\lambda_n\|Bx_n - Bx\|^2 \\ &\quad - \lambda_n(1 - \lambda_n)\|Tx_n - x_n\|^2 \\ &\leq \|x_n - x\|^2 - \gamma(2\beta - \gamma)\varepsilon\|Bx_n - Bx\|^2 \\ &\quad + \delta(\delta - 1)\|Tx_n - x_n\|^2. \end{aligned} \quad (25.28)$$

Thus, using Proposition 5.15(i)&(ii) and Proposition 5.4(ii), we obtain

$$\begin{aligned} \gamma(2\beta - \gamma)\varepsilon\|Bx_n - Bx\|^2 &\leq (\|x_n - x\|^2 - \|x_{n+1} - x\|^2) \\ &\quad + \delta(\delta - 1)\|Tx_n - x_n\|^2 \\ &\rightarrow 0, \end{aligned} \quad (25.29)$$

and the conclusion follows.

(iii): As seen in (i), there exists  $x \in \text{zer}(A + B)$  such that  $x_n \rightharpoonup x$ .

(iii)(a): Set  $(\forall n \in \mathbb{N}) \ z_n = J_{\gamma A}y_n$ . Then

$$-\gamma Bx \in \gamma Ax \quad \text{and} \quad (\forall n \in \mathbb{N}) \quad x_n - \gamma Bx_n - z_n = y_n - z_n \in \gamma Az_n. \quad (25.30)$$

On the other hand, we derive from Proposition 5.15(ii) that

$$z_n - x_n = Tx_n - x_n \rightarrow 0. \quad (25.31)$$

Thus,  $z_n \rightharpoonup x$  and hence  $C = \{x\} \cup \{z_n\}_{n \in \mathbb{N}}$  is a bounded subset of  $\text{dom } A$ . In turn, it follows from (22.5) and (25.30) that there exists an increasing function  $\phi_A: \mathbb{R}_+ \rightarrow [0, +\infty]$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \langle z_n - x \mid x_n - z_n - \gamma(Bx_n - Bx) \rangle \geq \gamma\phi_A(\|z_n - x\|). \quad (25.32)$$

However,  $z_n - x \rightharpoonup 0$ , and it follows from (25.31) and (ii) that  $x_n - z_n - \gamma(Bx_n - Bx) \rightarrow 0$ . We thus derive from (25.32) that  $\phi_A(\|z_n - x\|) \rightarrow 0$ , which forces  $z_n \rightarrow x$ . In view of (25.31), we conclude that  $x_n \rightarrow x$ .

(iii)(b): Since  $x_n \rightharpoonup x$ ,  $C = \{x\} \cup \{x_n\}_{n \in \mathbb{N}}$  is a bounded subset of  $\mathcal{H}$ . Hence, it follows from (22.5) that there exists an increasing function  $\phi_B: \mathbb{R}_+ \rightarrow [0, +\infty]$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \langle x_n - x \mid Bx_n - Bx \rangle \geq \gamma\phi_B(\|x_n - x\|). \quad (25.33)$$

Thus, (ii) yields  $\phi_B(\|x_n - x\|) \rightarrow 0$  and therefore  $x_n \rightarrow x$ .  $\square$

We conclude this section with two instances of linear convergence of the forward–backward algorithm.

**Proposition 25.9** *Let  $D$  be a nonempty closed subset of  $\mathcal{H}$ , let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and such that  $\text{dom } A \subset D$ , let  $B: D \rightarrow \mathcal{H}$ , let  $\alpha \in \mathbb{R}_{++}$ , and let  $\beta \in \mathbb{R}_{++}$ . Suppose that one of the following holds:*

- (i)  *$A$  is  $\alpha$ -strongly monotone,  $B$  is  $\beta$ -cocoercive, and  $\gamma \in ]0, 2\beta[$ .*
- (ii)  *$\alpha \leq \beta$ ,  $B$  is  $\alpha$ -strongly monotone and  $\beta$ -Lipschitz continuous, and  $\gamma \in ]0, 2\alpha/\beta^2[$ .*

Let  $x_0 \in D$  and set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n \\ x_{n+1} = J_{\gamma A} y_n. \end{cases} \quad (25.34)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges linearly to the unique point in  $\text{zer}(A + B)$ .

*Proof.* Set  $T = J_{\gamma A} \circ (\text{Id} - \gamma B)$  and note that, by Proposition 25.1(iv),  $\text{Fix } T = \text{zer}(A + B)$ . Since Proposition 23.2(i) asserts that  $\text{ran } J_{\gamma A} = \text{dom } A$ ,  $T$  is a well-defined operator from  $D$  to  $D$ , and (25.34) reduces to  $(\forall n \in \mathbb{N}) x_{n+1} = Tx_n$ . Since  $D$ , as a closed subset of  $\mathcal{H}$ , is a complete metric space, in view of Theorem 1.48, it is enough to show that in both cases  $T$  is Lipschitz continuous with a constant in  $[0, 1[$ .

(i): Set  $\tau = 1/(\alpha\gamma + 1)$ . On the one hand, we derive from Proposition 23.11 that  $J_{\gamma A}$  is Lipschitz continuous with constant  $\tau$ . On the other hand, we derive from Proposition 4.33 and Remark 4.24(i) that  $\text{Id} - \gamma B$  is nonexpansive. Altogether,  $T$  is Lipschitz continuous with constant  $\tau \in ]0, 1[$ .

(ii): Observe that  $\gamma(2\alpha - \gamma\beta^2) \in ]0, 1]$  and set  $\tau = \sqrt{1 - \gamma(2\alpha - \gamma\beta^2)}$ . On the one hand, we derive from Corollary 23.10(i) that  $J_{\gamma A}$  is nonexpansive. On the other hand, for every  $x$  and  $y$  in  $D$ , we have

$$\begin{aligned} \|(\text{Id} - \gamma B)x - (\text{Id} - \gamma B)y\|^2 &= \|x - y\|^2 - 2\gamma \langle x - y \mid Bx - By \rangle \\ &\quad + \gamma^2 \|Bx - By\|^2 \\ &\leq \|x - y\|^2 - 2\gamma\alpha \|x - y\|^2 + \gamma^2\beta^2 \|x - y\|^2 \\ &= (1 - \gamma(2\alpha - \gamma\beta^2)) \|x - y\|^2. \end{aligned} \quad (25.35)$$

Altogether,  $T$  is Lipschitz continuous with constant  $\tau \in [0, 1[$ . □

## 25.4 Tseng's Splitting Algorithm

In this section, we once again consider the case when  $B$  is single-valued in (25.8), but relax the cocoercivity condition imposed by the forward–backward algorithm in Theorem 25.8 at the expense of additional computations. More

precisely, the algorithm presented below involves at each iteration two forward steps, a backward step, and a projection step.

**Theorem 25.10 (Tseng's algorithm)** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and such that  $\text{dom } A \subset D$ , and let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a monotone operator that is single-valued on  $D$ . Suppose that  $A + B$  is maximally monotone, and that  $C$  is a closed convex subset of  $D$  such that  $C \cap \text{zer}(A + B) \neq \emptyset$  and  $B$  is  $1/\beta$ -Lipschitz continuous relative to  $C \cup \text{dom } A$ , for some  $\beta \in \mathbb{R}_{++}$ . Let  $x_0 \in C$ , let  $\gamma \in ]0, \beta[$ , and set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ z_n = J_{\gamma A} y_n, \\ r_n = z_n - \gamma Bz_n, \\ x_{n+1} = P_C(x_n - y_n + r_n). \end{cases} \quad (25.36)$$

Then the following hold:

- (i)  $(x_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (ii)  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to a point in  $C \cap \text{zer}(A + B)$ .
- (iii) Suppose that  $A$  or  $B$  is uniformly monotone on every nonempty bounded subset of  $\text{dom } A$ . Then  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique point in  $C \cap \text{zer}(A + B)$ .

*Proof.* Suppose that, for some  $n \in \mathbb{N}$ ,  $x_n \in C$ . Then  $x_n \in D$  and  $y_n$  is therefore well defined. In turn, we derive from Proposition 23.2(i) that  $z_n \in \text{ran } J_{\gamma A} = \text{dom } A \subset D$ , which makes  $r_n$  well defined. Finally,  $x_{n+1} = P_C(r_n + x_n - y_n) \in C$ . This shows by induction that the sequences  $(x_n)_{n \in \mathbb{N}}$ ,  $(y_n)_{n \in \mathbb{N}}$ ,  $(z_n)_{n \in \mathbb{N}}$ , and  $(r_n)_{n \in \mathbb{N}}$  are well defined. Let us set

$$(\forall n \in \mathbb{N}) \quad u_n = \gamma^{-1}(x_n - z_n) + Bz_n - Bx_n. \quad (25.37)$$

Note that (25.36) yields

$$(\forall n \in \mathbb{N}) \quad u_n = \gamma^{-1}(y_n - z_n) + Bz_n \in Az_n + Bz_n. \quad (25.38)$$

Now let  $z \in C \cap \text{zer}(A + B)$  and let  $n \in \mathbb{N}$ . We first note that

$$z = P_C z \quad \text{and} \quad (z, -\gamma Bz) \in \text{gra } \gamma A. \quad (25.39)$$

On the other hand, by Proposition 23.2(ii) and (25.36),  $(z_n, y_n - z_n) \in \text{gra } \gamma A$ . Hence, by (25.39) and monotonicity of  $\gamma A$ ,  $\langle z_n - z \mid z_n - y_n - \gamma Bz \rangle \leq 0$ . However, by monotonicity of  $B$ ,  $\langle z_n - z \mid \gamma Bz - \gamma Bz_n \rangle \leq 0$ . Upon adding these two inequalities, we obtain

$$\langle z_n - z \mid z_n - y_n - \gamma Bz_n \rangle \leq 0. \quad (25.40)$$

In turn, we derive from (25.36) that

$$2\gamma \langle z_n - z \mid Bx_n - Bz_n \rangle = 2 \langle z_n - z \mid z_n - y_n - \gamma Bz_n \rangle$$

$$\begin{aligned}
& + 2 \langle z_n - z \mid \gamma Bx_n + y_n - z_n \rangle \\
& \leq 2 \langle z_n - z \mid \gamma Bx_n + y_n - z_n \rangle \\
& = 2 \langle z_n - z \mid x_n - z_n \rangle \\
& = \|x_n - z\|^2 - \|z_n - z\|^2 - \|x_n - z_n\|^2 \quad (25.41)
\end{aligned}$$

and, therefore, from Proposition 4.8 that

$$\begin{aligned}
\|x_{n+1} - z\|^2 &= \|P_C(r_n + x_n - y_n) - P_C z\|^2 \\
&\leq \|r_n + x_n - y_n - z\|^2 \\
&= \|(z_n - z) + \gamma(Bx_n - Bz_n)\|^2 \\
&= \|z_n - z\|^2 + 2\gamma \langle z_n - z \mid Bx_n - Bz_n \rangle + \gamma^2 \|Bx_n - Bz_n\|^2 \\
&\leq \|x_n - z\|^2 - \|x_n - z_n\|^2 + \gamma^2 \|Bx_n - Bz_n\|^2 \\
&\leq \|x_n - z\|^2 - (1 - \gamma^2/\beta^2) \|x_n - z_n\|^2. \quad (25.42)
\end{aligned}$$

This shows that

$$(x_n)_{n \in \mathbb{N}} \text{ is Fejér monotone with respect to } C \cap \text{zer}(A + B). \quad (25.43)$$

(i): An immediate consequence of (25.42).

(ii): It follows from (i), the relative Lipschitz continuity of  $B$ , and (25.37) that

$$Bz_n - Bx_n \rightarrow 0 \quad \text{and} \quad u_n \rightarrow 0. \quad (25.44)$$

Now let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ , say  $x_{k_n} \rightharpoonup x$ . Let us show that  $x \in C \cap \text{zer}(A + B)$ . Since  $(x_n)_{n \in \mathbb{N}}$  lies in  $C$ , which is weakly sequentially closed by Theorem 3.32, we have  $x \in C$  and it remains to show that  $(x, 0) \in \text{gra}(A + B)$ . It follows from (i) that  $z_{k_n} \rightharpoonup x$ , and from (25.44) that  $u_{k_n} \rightarrow 0$ . Altogether,  $(z_{k_n}, u_{k_n})_{n \in \mathbb{N}}$  lies in  $\text{gra}(A + B)$  by (25.38), and it satisfies

$$z_{k_n} \rightharpoonup x \quad \text{and} \quad u_{k_n} \rightarrow 0. \quad (25.45)$$

Since  $A + B$  is maximally monotone, it follows from Proposition 20.33(ii) that  $(x, 0) \in \text{gra}(A + B)$ . In view of (25.43), Theorem 5.5, and (i), the assertions are proved.

(iii): Since  $A + B$  is strictly monotone, it follows from Proposition 23.35 that  $\text{zer}(A + B)$  is a singleton. As shown in (ii), there exists  $x \in C \cap \text{zer}(A + B) \subset \text{dom } A$  such that

$$z_n \rightharpoonup x. \quad (25.46)$$

The assumptions imply that  $A + B$  is uniformly monotone on  $\{x\} \cup \{z_n\}_{n \in \mathbb{N}} \subset \text{dom } A$ . Hence, since  $0 \in (A + B)x$ , it follows from (25.38) and (22.5) that there exists an increasing function  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \langle z_n - x \mid u_n \rangle \geq \gamma \phi(\|z_n - x\|). \quad (25.47)$$



We therefore deduce from (25.46) and (25.44) that  $\phi(\|z_n - x\|) \rightarrow 0$ , which implies that  $z_n \rightarrow x$ . In turn, (i) yields  $x_n \rightarrow x$ .  $\square$

**Remark 25.11** Here are a few observations concerning Theorem 25.10.

- (i) Sufficient conditions for the maximal monotonicity of  $A + B$  are discussed in Theorem 24.3 and Corollary 24.4.
- (ii) The set  $C$  can be used to impose constraints on the zeros of  $A + B$ .
- (iii) If  $\text{dom } A$  is closed, then it follows from Corollary 21.12 that it is convex and we can therefore choose  $C = \text{dom } A$ .
- (iv) If  $\text{dom } B = \mathcal{H}$ , we can choose  $C = \mathcal{H}$ . In this case, it follows that  $B$  is continuous and single-valued on  $\mathcal{H}$ , hence maximally monotone by Corollary 20.25. In turn, Corollary 24.4(i) implies that  $A + B$  is maximally monotone and (25.36) reduces to the *forward-backward-forward algorithm*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ z_n = J_{\gamma A} y_n, \\ x_{n+1} = x_n - y_n + z_n - \gamma Bz_n. \end{cases} \quad (25.48)$$

## 25.5 Variational Inequalities

**Definition 25.12** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone. The associated *variational inequality* problem is to

$$\text{find } x \in \mathcal{H} \text{ such that } (\exists u \in Bx)(\forall y \in \mathcal{H}) \quad \langle x - y \mid u \rangle + f(x) \leq f(y). \quad (25.49)$$

Here are a few examples of variational inequalities (additional examples will arise in the context of minimization problems in Section 26.1 and Section 26.2).

**Example 25.13** In Definition 25.12, let  $z \in \mathcal{H}$  and set  $B: x \mapsto x - z$ . Then we obtain the variational inequality problem

$$\text{find } x \in \mathcal{H} \text{ such that } (\forall y \in \mathcal{H}) \quad \langle x - y \mid x - z \rangle + f(x) \leq f(y). \quad (25.50)$$

As seen in Proposition 12.26, the solution to this problem is  $\text{Prox}_f z$ .

**Example 25.14** In Definition 25.12, set  $f = \iota_C$ , where  $C$  is a nonempty closed convex subset of  $\mathcal{H}$ , and let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be maximally monotone. Then we obtain the classical variational inequality problem

$$\text{find } x \in C \text{ such that } (\forall y \in C) \quad \langle x - y \mid Bx \rangle \leq 0. \quad (25.51)$$

In particular, if  $B: x \mapsto x - z$ , where  $z \in \mathcal{H}$ , we recover the variational inequality that characterizes the projection of  $z$  onto  $C$  (see Theorem 3.14).

**Example 25.15 (complementarity problem)** In Example 25.14, set  $C = K$ , where  $K$  is a nonempty closed convex cone in  $\mathcal{H}$ . Then we obtain the *complementarity problem*

$$\text{find } x \in K \text{ such that } x \perp Bx \text{ and } Bx \in K^\oplus. \quad (25.52)$$

*Proof.* If  $x \in K$ , then  $\{x/2, 2x\} \subset K$  and the condition  $\sup_{y \in K} \langle x - y \mid Bx \rangle \leq 0$  implies that  $\langle x - x/2 \mid Bx \rangle \leq 0$  and  $\langle x - 2x \mid Bx \rangle \leq 0$ , hence  $\langle x \mid Bx \rangle = 0$ . It therefore reduces to  $\sup_{y \in K} \langle -y \mid Bx \rangle \leq 0$ , i.e., by Definition 6.21, to  $Bx \in K^\oplus$ .  $\square$

**Remark 25.16** In view of (16.1), the variational inequality problem (25.49) can be recast as that of finding a point in  $\text{zer}(A + B)$ , where  $A = \partial f$  is maximally monotone by Theorem 20.40.

It follows from Remark 25.16 that we can apply the splitting algorithms considered earlier in this chapter to solve variational inequalities. We start with an application of the Douglas–Rachford algorithm.

**Proposition 25.17 (Douglas–Rachford algorithm)** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, and suppose that the variational inequality*

$$\text{find } x \in \mathcal{H} \text{ such that } (\exists u \in Bx)(\forall y \in \mathcal{H}) \langle x - y \mid u \rangle + f(x) \leq f(y) \quad (25.53)$$

*admits at least one solution. Let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = J_{\gamma B} x_n, \\ z_n = \text{Prox}_{\gamma f}(2y_n - x_n), \\ x_{n+1} = x_n + \lambda_n(z_n - y_n). \end{cases} \quad (25.54)$$

*Then there exists  $x \in \mathcal{H}$  such that the following hold:*

- (i)  $J_{\gamma B} x$  is a solution to (25.53).
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to  $x$ .
- (iii)  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to  $J_{\gamma B} x$ .
- (iv) Suppose that one of the following holds:
  - (a)  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ .
  - (b)  $B$  is uniformly monotone on every nonempty bounded subset of  $\text{dom } B$ .

*Then  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique solution to (25.53).*

*Proof.* Apply items (i), (iii), (iv), (v), and (vii) of Theorem 25.6 to  $A = \partial f$ , and use Example 23.3 and Example 22.4.  $\square$

Next, we consider the case when  $B$  is single-valued, with an application of the forward–backward algorithm (linear convergence results can be derived from Proposition 25.9 in a similar fashion).

**Proposition 25.18 (forward–backward algorithm)** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $\beta \in \mathbb{R}_{++}$ , let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive, let  $\gamma \in ]0, 2\beta[$ , and set  $\delta = \min\{1, \beta/\gamma\} + 1/2$ . Suppose that the variational inequality*

$$\text{find } x \in \mathcal{H} \text{ such that } (\forall y \in \mathcal{H}) \quad \langle x - y \mid Bx \rangle + f(x) \leq f(y) \quad (25.55)$$

*admits at least one solution. Furthermore, let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, \delta]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(\delta - \lambda_n) = +\infty$ , let  $x_0 \in \mathcal{H}$ , and set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ x_{n+1} = x_n + \lambda_n (\text{Prox}_{\gamma f} y_n - x_n). \end{cases} \quad (25.56)$$

*Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a solution to (25.55).
- (ii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and let  $x$  be a solution to (25.55). Then  $(Bx_n)_{n \in \mathbb{N}}$  converges strongly to  $Bx$ .
- (iii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and that one of the following holds:

- (a)  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ .
- (b)  $B$  is uniformly monotone on every nonempty bounded subset of  $\mathcal{H}$ .

*Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique point in  $\text{zer}(A + B)$ .*

*Proof.* This is an application of Theorem 25.8 to  $A = \partial f$ , Example 23.3, and Example 22.4.  $\square$

**Example 25.19** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $\beta \in \mathbb{R}_{++}$ , and let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive. Suppose that the variational inequality

$$\text{find } x \in C \text{ such that } (\forall y \in C) \quad \langle x - y \mid Bx \rangle \leq 0 \quad (25.57)$$

admits at least one solution. Let  $x_0 \in \mathcal{H}$  and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = P_C(x_n - \beta Bx_n). \quad (25.58)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a solution  $x$  to (25.57) and, moreover,  $(Bx_n)_{n \in \mathbb{N}}$  converges strongly to  $Bx$ .

*Proof.* Apply Proposition 25.18 to  $f = \iota_C$ ,  $\gamma = \beta$ , and  $\lambda_n \equiv 1$ , and use Example 12.25.  $\square$

In instances in which  $B$  is not cocoercive on  $\mathcal{H}$ , one can turn to Tseng's splitting algorithm of Section 25.4. Here is an illustration.

**Example 25.20** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $\beta \in \mathbb{R}_{++}$ , and let  $B: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be a maximally monotone operator that is single-valued and  $\beta$ -Lipschitz continuous relative to  $C$ . Suppose that

$$\text{cone}(C - \text{dom } B) = \overline{\text{span}}(C - \text{dom } B), \quad (25.59)$$

and that the variational inequality

$$\text{find } x \in C \text{ such that } (\forall y \in C) \langle x - y \mid Bx \rangle \leq 0 \quad (25.60)$$

admits at least one solution. Let  $x_0 \in C$ , let  $\gamma \in ]0, 1/\beta[$ , and set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ z_n = P_C y_n, \\ r_n = z_n - \gamma Bz_n, \\ x_{n+1} = P_C(x_n - y_n + r_n). \end{cases} \quad (25.61)$$

Then the following hold:

- (i)  $(x_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (ii)  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to a solution to (25.60).
- (iii) Suppose that  $B$  is uniformly monotone on every nonempty bounded subset of  $C$ . Then  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique solution to (25.60).

*Proof.* It follows from Example 20.41, (25.59), and Theorem 24.3 that the operator  $N_C + B$  is maximally monotone. Hence, the results follow from Theorem 25.10 applied to  $D = C$  and  $A = N_C$ , and by invoking Example 20.41 and Example 23.4.  $\square$

## Exercises

**Exercise 25.1** Let  $C$  and  $D$  be nonempty closed convex subsets of  $\mathcal{H}$ . Show that  $C \cap D = P_D(\text{Fix}(2P_C - \text{Id}) \circ (2P_D - \text{Id}))$ .

**Exercise 25.2** Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ . Use Exercise 24.13 to show that  $\text{zer}(A + B)$  is convex.

**Exercise 25.3** Let  $A$  and  $B$  be operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ .

- (i) Show that  $\text{zer}(A + B) \neq \emptyset \Leftrightarrow \text{zer}(A^{-1} - (B^{-1})^\vee) \neq \emptyset$ .
- (ii) Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$ , and set  $A = \partial f$  and  $B = \partial g$ . Do we recover Proposition 15.13 from (i)?

**Exercise 25.4** In the setting of Theorem 25.6, suppose that  $\mathcal{H}$  is finite-dimensional. Show that  $(y_n)_{n \in \mathbb{N}}$  converges to a point in  $\text{zer}(A + B)$  without using items (iv) and (v).

**Exercise 25.5** In the setting of Theorem 25.6, suppose that  $B$  is affine and continuous. Show that  $(y_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer}(A + B)$  without using items (iv) and (v).

**Exercise 25.6** Let  $A$  and  $B$  be maximally monotone operators from  $\mathcal{H}$  to  $2^{\mathcal{H}}$ , let  $\gamma \in \mathbb{R}_{++}$ , and suppose that the problem

$$\text{find } x \in \mathcal{H} \quad \text{such that} \quad 0 \in Ax + {}^{\mu}Bx \quad (25.62)$$

has at least one solution. Devise an algorithm using the resolvent of  $A$  and  $B$  separately to solve (25.62), and establish rigorously a convergence result.

**Exercise 25.7** Deduce the strong and weak convergence results of Example 23.40 from Theorem 25.6.

**Exercise 25.8** In Theorem 25.6 make the additional assumption that the operators  $A$  and  $B$  are odd and that  $(\forall n \in \mathbb{N}) \lambda_n = 1$ . Prove the following:

- (i)  $J_{\gamma A}$  and  $J_{\gamma B}$  are odd.
- (ii) The convergence of the sequences  $(x_n)_{n \in \mathbb{N}}$ ,  $(y_n)_{n \in \mathbb{N}}$ , and  $(z_n)_{n \in \mathbb{N}}$  defined in (25.9) is strong.

**Exercise 25.9** Deduce the convergence results of Example 23.40 from Theorem 25.8 and Proposition 25.9.

**Exercise 25.10** In Theorem 25.8(i), make the additional assumption that  $\text{int zer}(A + B) \neq \emptyset$ . Show that the sequence produced by the forward-backward algorithm (25.26) converges strongly to a zero of  $A + B$ .

**Exercise 25.11** In the setting of Proposition 25.9(ii), show that  $B$  is coco-ercive.



# Chapter 26

## Fermat's Rule in Convex Optimization

Fermat's rule (Theorem 16.2) provides a simple characterization of the minimizers of a function as the zeros of its subdifferential. This chapter explores various consequences of this fact.

Throughout,  $\mathcal{K}$  is a real Hilbert space.

### 26.1 General Characterizations of Minimizers

Let us first provide characterizations of the minimizers of a function in  $\Gamma_0(\mathcal{H})$ .

**Proposition 26.1** *Let  $f \in \Gamma_0(\mathcal{H})$ . Then*

$$\operatorname{Argmin} f = \operatorname{zer} \partial f = \partial f^*(0) = \operatorname{Fix} \operatorname{Prox}_f = \operatorname{zer} \operatorname{Prox}_{f^*}. \quad (26.1)$$

*Proof.* Let  $x \in \mathcal{H}$ . Then  $x \in \operatorname{Argmin} f \Leftrightarrow 0 \in \partial f(x)$  (by Theorem 16.2)  $\Leftrightarrow x - 0 = x \in \partial f^*(0)$  (by Corollary 16.24)  $\Leftrightarrow \operatorname{Prox}_{f^*} x = 0$  (by (16.30))  $\Leftrightarrow \operatorname{Prox}_f x = x$  (by (14.6)).  $\square$

The next theorem is an application of Fermat's rule to the minimization of the sum of two convex functions satisfying a constraint qualification.

**Theorem 26.2** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g \in \Gamma_0(\mathcal{K})$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $\bar{x} \in \mathcal{H}$ . Suppose that one of the following holds:*

- (a)  $0 \in \operatorname{sri}(\operatorname{dom} g - L(\operatorname{dom} f))$  (see Proposition 6.19 for special cases).
- (b)  $\mathcal{K}$  is finite-dimensional,  $g$  is polyhedral, and  $\operatorname{dom} g \cap \operatorname{ri} L(\operatorname{dom} f) \neq \emptyset$ .
- (c)  $\mathcal{H}$  and  $\mathcal{K}$  are finite-dimensional,  $f$  and  $g$  are polyhedral, and  $\operatorname{dom} g \cap L(\operatorname{dom} f) \neq \emptyset$ .

*Then the following are equivalent:*

- (i)  $\bar{x}$  is a solution to the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(Lx). \quad (26.2)$$

- (ii)  $\bar{x} \in \text{zer}(\partial f + L^* \circ (\partial g) \circ L)$ .
- (iii)  $(\exists v \in \partial g(L\bar{x})) -L^*v \in \partial f(\bar{x})$ .
- (iv)  $(\exists v \in \partial g(L\bar{x}))(\forall y \in \mathcal{H}) \langle \bar{x} - y \mid L^*v \rangle + f(\bar{x}) \leq f(y)$ .

Moreover, if  $g$  is Gâteaux differentiable at  $L\bar{x}$ , each of items (i)–(iv) is also equivalent to each of the following:

- (v)  $-L^*(\nabla g(L\bar{x})) \in \partial f(\bar{x})$ .
- (vi)  $(\forall y \in \mathcal{H}) \langle \bar{x} - y \mid L^*(\nabla g(L\bar{x})) \rangle + f(\bar{x}) \leq f(y)$ .
- (vii)  $(\forall \gamma \in \mathbb{R}_{++}) \bar{x} = \text{Prox}_{\gamma f}(\bar{x} - \gamma L^*(\nabla g(L\bar{x})))$ .

*Proof.* (i) $\Leftrightarrow$ (ii): It follows from Proposition 26.1 and Theorem 16.37 that  $\text{Argmin}(f + g \circ L) = \text{zer } \partial(f + g \circ L) = \text{zer}(\partial f + L^* \circ (\partial g) \circ L)$ .

(ii) $\Leftrightarrow$ (iii):  $\bar{x} \in \text{zer}(\partial f + L^* \circ (\partial g) \circ L) \Leftrightarrow 0 \in \partial f(\bar{x}) + L^*(\partial g(L\bar{x})) \Leftrightarrow (\exists v \in \partial g(L\bar{x})) -L^*v \in \partial f(\bar{x})$ .

(iii) $\Leftrightarrow$ (iv): Definition 16.1.

Now assume that  $g$  is Gâteaux differentiable at  $L\bar{x}$ .

(iii) $\Leftrightarrow$ (v): Proposition 17.26(i).

(iv) $\Leftrightarrow$ (vi): Proposition 17.26(i).

(v) $\Leftrightarrow$ (vii): Let  $\gamma \in \mathbb{R}_{++}$ . Then (16.30) yields  $-L^*(\nabla g(L\bar{x})) \in \partial f(\bar{x}) \Leftrightarrow (\bar{x} - \gamma L^*(\nabla g(L\bar{x}))) - \bar{x} \in \partial(\gamma f)(\bar{x}) \Leftrightarrow \bar{x} = \text{Prox}_{\gamma f}(\bar{x} - \gamma L^*(\nabla g(L\bar{x})))$ .  $\square$

**Corollary 26.3** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$ , let  $\bar{x} \in \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Suppose that one of the following holds:*

- (a)  $0 \in \text{sri}(\text{dom } g - \text{dom } f)$  (see Proposition 6.19 for special cases).
- (b)  $\mathcal{H}$  is finite-dimensional,  $g$  is polyhedral, and  $\text{dom } g \cap \text{ri dom } f \neq \emptyset$ .
- (c)  $\mathcal{H}$  is finite-dimensional,  $f$  and  $g$  are polyhedral, and  $\text{dom } f \cap \text{dom } g \neq \emptyset$ .

Then the following are equivalent:

- (i)  $\bar{x}$  is a solution to the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + g(x). \quad (26.3)$$

- (ii)  $\bar{x} \in \text{zer}(\partial f + \partial g)$ .
- (iii)  $\bar{x} \in \text{Prox}_{\gamma g}(\text{Fix}(2 \text{Prox}_{\gamma f} - \text{Id}) \circ (2 \text{Prox}_{\gamma g} - \text{Id}))$ .
- (iv)  $(\exists u \in \partial g(\bar{x})) -u \in \partial f(\bar{x})$ .
- (v)  $(\exists u \in \partial g(\bar{x}))(\forall y \in \mathcal{H}) \langle \bar{x} - y \mid u \rangle + f(\bar{x}) \leq f(y)$ .

Moreover, if  $g$  is Gâteaux differentiable at  $\bar{x}$ , each of items (i)–(v) is also equivalent to each of the following:

- (vi)  $-\nabla g(\bar{x}) \in \partial f(\bar{x})$ .
- (vii)  $(\forall y \in \mathcal{H}) \langle \bar{x} - y \mid \nabla g(\bar{x}) \rangle + f(\bar{x}) \leq f(y)$ .
- (viii)  $\bar{x} = \text{Prox}_{\gamma f}(\bar{x} - \gamma \nabla g(\bar{x}))$ .



*Proof.* Applying Theorem 26.2 to  $\mathcal{K} = \mathcal{H}$  and  $L = \text{Id}$  yields all the results, except the equivalences involving (iii). The equivalence (ii)  $\Leftrightarrow$  (iii) follows from Theorem 20.40, Proposition 25.1(ii), and Example 23.3.  $\square$

**Remark 26.4** Condition (v) in Corollary 26.3 is an instance of the variational inequality featured in Definition 25.12.

## 26.2 Abstract Constrained Minimization Problems

The problem under consideration in this section is the characterization of the minimizers of a function  $f \in \Gamma_0(\mathcal{H})$  over a closed convex set.

**Proposition 26.5** *Let  $C$  be a closed convex subset of  $\mathcal{H}$ , let  $f \in \Gamma_0(\mathcal{H})$ , let  $\bar{x} \in \mathcal{H}$ , and let  $\gamma \in \mathbb{R}_{++}$ . Suppose that one of the following holds:*

- (a)  $0 \in \text{sri}(C - \text{dom } f)$  (see Proposition 6.19 for special cases).
- (b)  $\mathcal{H}$  is finite-dimensional,  $C$  is polyhedral, and  $C \cap \text{ri}(\text{dom } f) \neq \emptyset$ .
- (c)  $\mathcal{H}$  is finite-dimensional,  $C$  is a polyhedral set,  $f$  is a polyhedral function, and  $C \cap \text{dom } f \neq \emptyset$ .

*Then the following are equivalent:*

- (i)  $\bar{x}$  is a solution to the problem

$$\underset{x \in C}{\text{minimize}} \quad f(x). \quad (26.4)$$

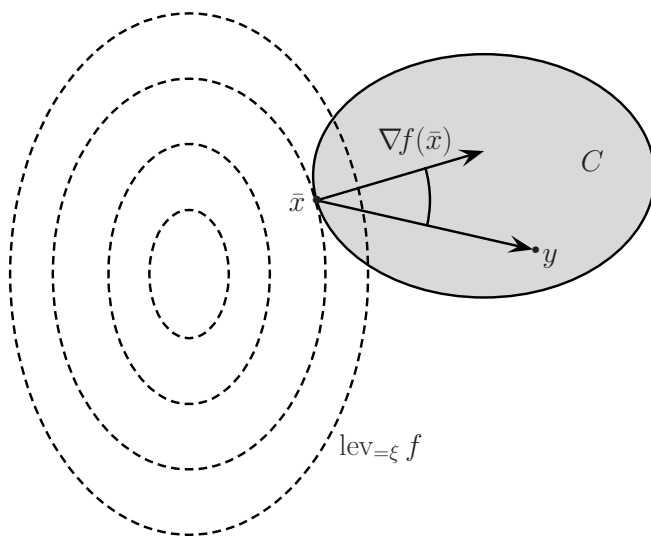
- (ii)  $\bar{x} \in \text{zer}(N_C + \partial f)$ .
- (iii)  $\bar{x} \in \text{Prox}_{\gamma f}(\text{Fix}(2P_C - \text{Id}) \circ (2\text{Prox}_{\gamma f} - \text{Id}))$ .
- (iv)  $(\exists u \in N_C \bar{x}) -u \in \partial f(\bar{x})$ .
- (v)  $(\exists u \in \partial f(\bar{x})) -u \in N_C \bar{x}$ .
- (vi)  $\bar{x} \in C$  and  $(\exists u \in \partial f(\bar{x}))(\forall y \in C) \langle \bar{x} - y \mid u \rangle \leq 0$ .

*Moreover, if  $f$  is Gâteaux differentiable at  $\bar{x}$ , each of items (i)–(vi) is also equivalent to each of the following:*

- (vi)  $-\nabla f(\bar{x}) \in N_C \bar{x}$ .
- (vii)  $\bar{x} \in C$  and  $(\forall y \in C) \langle \bar{x} - y \mid \nabla f(\bar{x}) \rangle \leq 0$ .
- (viii)  $\bar{x} = P_C(\bar{x} - \gamma \nabla f(\bar{x}))$ .

*Proof.* Apply Corollary 26.3 to the functions  $\iota_C$  and  $f$ , and use Example 16.12 and Example 12.25.  $\square$

**Remark 26.6** Condition (vii) in Proposition 26.5 is an instance of the variational inequality considered in Example 25.14.



**Fig. 26.1** Illustration of the equivalence (i) ⇔ (vii) in Proposition 26.5 when  $\mathcal{H} = \mathbb{R}^2$ :  $\bar{x} \in C$  is a minimizer of  $f$  over  $C$  if and only if, for every  $y \in C$ ,  $\langle y - \bar{x} \mid \nabla f(\bar{x}) \rangle \geq 0$ , i.e., the vectors  $y - \bar{x}$  and  $\nabla f(\bar{x})$  form an acute angle. Each dashed line represents a level line  $\text{lev}_{=\xi} f$ .

**Example 26.7** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $x \in \mathcal{H}$ , let  $p \in \mathcal{H}$ , and set  $f = (1/2)\|\cdot - x\|^2$ . Then Example 2.48 yields  $\nabla f: y \mapsto y - x$  and we deduce from Proposition 26.5 that  $p = P_C x \Leftrightarrow x - p \in N_C p$ . We thus recover Proposition 6.46.

The following two results have found many uses in the study of partial differential equations.

**Example 26.8 (Stampacchia)** Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow \mathbb{R}$  be a bilinear form such that, for some  $\alpha$  and  $\beta$  in  $\mathbb{R}_{++}$ ,

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad |F(x, y)| \leq \beta \|x\| \|y\| \quad \text{and} \quad F(x, x) \geq \alpha \|x\|^2, \quad (26.5)$$

let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , and let  $\ell \in \mathcal{B}(\mathcal{H}, \mathbb{R})$ . Then the following hold:

- (i) There exists a unique point  $\bar{x} \in \mathcal{H}$  such that

$$\bar{x} \in C \quad \text{and} \quad (\forall y \in C) \quad F(\bar{x}, y - \bar{x}) \geq \ell(y - \bar{x}). \quad (26.6)$$

- (ii) Suppose that  $F$  is symmetric, let  $\bar{x} \in \mathcal{H}$ , and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2)F(x, x) - \ell(x)$ . Then the following are equivalent:
- (a)  $\bar{x}$  solves (26.6).
  - (b)  $\text{Argmin}_C f = \{\bar{x}\}$ .

*Proof.* (i): By Fact 2.17, there exists  $u \in \mathcal{H}$  such that  $\ell = \langle \cdot | u \rangle$ . Similarly, for every  $x \in \mathcal{H}$ ,  $F(x, \cdot) \in \mathcal{B}(\mathcal{H}, \mathbb{R})$  and, therefore, there exists a vector in  $\mathcal{H}$ , which we denote by  $Lx$ , such that  $F(x, \cdot) = \langle \cdot | Lx \rangle$ . We derive from (26.5) and Example 20.29 that  $L$  is a strongly monotone, maximally monotone operator in  $\mathcal{B}(\mathcal{H})$ , and hence that  $B: x \mapsto Lx - u$  is likewise. Furthermore, (6.31) implies that the set of solutions to (26.6) is  $\text{zer}(N_C + B)$ , while Example 20.41 and Corollary 24.4(i) imply that  $N_C + B$  is maximally monotone. Thus, since  $N_C + B$  is strongly monotone, the claim follows from Corollary 23.37(ii).

(ii): It is clear that  $f \in \Gamma_0(\mathcal{H})$  and that  $\text{dom } f = \mathcal{H}$ . Moreover, by Example 2.47,  $Df(x) = F(x, \cdot) - \ell$ . Hence, the result follows from (i) and the equivalence (i) $\Leftrightarrow$ (vii) in Proposition 26.5.  $\square$

**Example 26.9 (Lax–Milgram)** Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow \mathbb{R}$  be a bilinear form such that, for some  $\alpha$  and  $\beta$  in  $\mathbb{R}_{++}$ ,

$$(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) \quad |F(x, y)| \leq \beta \|x\| \|y\| \quad \text{and} \quad F(x, x) \geq \alpha \|x\|^2, \quad (26.7)$$

and let  $\ell \in \mathcal{B}(\mathcal{H}, \mathbb{R})$ . Then the following hold:

- (i) There exists a unique point  $\bar{x} \in \mathcal{H}$  such that

$$(\forall y \in \mathcal{H}) \quad F(\bar{x}, y) = \ell(y). \quad (26.8)$$

- (ii) Suppose that  $F$  is symmetric, let  $\bar{x} \in \mathcal{H}$ , and set  $f: \mathcal{H} \rightarrow \mathbb{R}: x \mapsto (1/2)F(x, x) - \ell(x)$ . Then the following are equivalent:
- (a)  $\bar{x}$  solves (26.8).
  - (b)  $\text{Argmin } f = \{\bar{x}\}$ .

*Proof.* Set  $C = \mathcal{H}$  in Example 26.8.  $\square$

**Proposition 26.10** Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{int dom } f \neq \emptyset$ , and let  $C$  be a nonempty convex subset of  $\text{int dom } f$ . Suppose that  $f$  is Gâteaux differentiable on  $C$ , and that  $x$  and  $y$  belong to  $\text{Argmin}_C f$ . Then  $\nabla f(x) = \nabla f(y)$ .

*Proof.* By the implication (i) $\Rightarrow$ (vii) in Proposition 26.5,  $\langle x - y | \nabla f(x) \rangle \leq 0$  and  $\langle y - x | \nabla f(y) \rangle \leq 0$ . Hence,  $\langle x - y | \nabla f(x) - \nabla f(y) \rangle \leq 0$  and, by Proposition 17.10(iii),  $\langle x - y | \nabla f(x) - \nabla f(y) \rangle = 0$ . In turn, by Example 22.3(i) and Proposition 17.26(i),  $f$  is paramonotone. Thus,  $\nabla f(x) = \nabla f(y)$ .  $\square$

## 26.3 Affine Constraints

We first revisit the setting of Proposition 19.19 in the light of Theorem 26.2.

**Proposition 26.11** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $\bar{x} \in \mathcal{H}$ , and suppose that*

$$r \in \text{sri } L(\text{dom } f). \quad (26.9)$$

*Consider the problem*

$$\underset{\substack{x \in \mathcal{H} \\ Lx=r}}{\text{minimize}} \quad f(x). \quad (26.10)$$

*Then  $\bar{x}$  is a solution to (26.10) if and only if*

$$L\bar{x} = r \quad \text{and} \quad (\exists \bar{v} \in \mathcal{K}) \quad -L^*\bar{v} \in \partial f(\bar{x}), \quad (26.11)$$

*in which case  $\bar{v}$  is a Lagrange multiplier associated with  $\bar{x}$ , and  $\bar{x}$  solves the problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \langle x \mid L^*\bar{v} \rangle. \quad (26.12)$$

*Moreover, if  $f$  is Gâteaux differentiable at  $\bar{x}$ , (26.11) becomes*

$$L\bar{x} = r \quad \text{and} \quad (\exists \bar{v} \in \mathcal{K}) \quad \nabla f(\bar{x}) = -L^*\bar{v}. \quad (26.13)$$

*Proof.* Set  $g = \iota_{\{r\}}$ . Then Problem (26.10) is a special case of (26.2). Moreover, (26.9) implies that condition (a) in Theorem 26.2 is satisfied. Hence, the characterization (26.11) follows from the equivalence (i)  $\Leftrightarrow$  (iii) in Theorem 26.2, from which we deduce the characterization (26.13) via Proposition 17.26(i). Finally, it follows from Proposition 19.19(v), Theorem 15.23, and Remark 19.20 that  $\bar{v}$  is a Lagrange multiplier associated with  $\bar{x}$ , and that  $\bar{x}$  solves (26.12).  $\square$

In the next corollary, we revisit the setting of Corollary 19.21 using the tools of Proposition 26.11.

**Corollary 26.12** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(u_i)_{i \in I} \in \mathcal{H}^m$ , and suppose that*

$$(\rho_i)_{i \in I} \in \text{ri} \left\{ (\langle x \mid u_i \rangle)_{i \in I} \mid x \in \text{dom } f \right\}. \quad (26.14)$$

*Consider the problem*

$$\underset{\substack{x \in \mathcal{H} \\ \langle x \mid u_1 \rangle = \rho_1, \dots, \langle x \mid u_m \rangle = \rho_m}}{\text{minimize}} \quad f(x), \quad (26.15)$$

*and let  $\bar{x} \in \mathcal{H}$ . Then  $\bar{x}$  is a solution to (26.15) if and only if*

$$(\forall i \in I) \quad \langle \bar{x} \mid u_i \rangle = \rho_i \quad \text{and} \quad (\exists (\bar{v}_i)_{i \in I} \in \mathbb{R}^m) \quad - \sum_{i \in I} \bar{v}_i u_i \in \partial f(\bar{x}), \quad (26.16)$$

in which case  $(\bar{\nu}_i)_{i \in I}$  are Lagrange multipliers associated with  $\bar{x}$ , and  $\bar{x}$  solves the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \sum_{i \in I} \bar{\nu}_i \langle x \mid u_i \rangle. \quad (26.17)$$

Moreover, if  $f$  is Gâteaux differentiable at  $\bar{x}$ , (26.16) becomes

$$(\forall i \in I) \quad \langle \bar{x} \mid u_i \rangle = \rho_i \quad \text{and} \quad (\exists (\bar{\nu}_i)_{i \in I} \in \mathbb{R}^m) \quad \nabla f(\bar{x}) = - \sum_{i \in I} \bar{\nu}_i u_i. \quad (26.18)$$

*Proof.* Set  $\mathcal{K} = \mathbb{R}^m$ ,  $L: \mathcal{H} \rightarrow \mathcal{K}: x \mapsto (\langle x \mid u_i \rangle)_{i \in I}$ , and  $r = (\rho_i)_{i \in I}$ . Then (26.15) appears as a special case of (26.10),  $L^*: (\eta_i)_{i \in I} \mapsto \sum_{i \in I} \eta_i u_i$ , and it follows from Fact 6.14(i) that (26.14) coincides with (26.9). Altogether, the results follow from Proposition 26.11.  $\square$

Next, we consider a more geometrical formulation by revisiting Proposition 26.5 in the case when  $C$  is a closed affine subspace.

**Proposition 26.13** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $V$  be a closed linear subspace of  $\mathcal{H}$ , let  $z \in \mathcal{H}$ , and let  $\bar{x} \in \mathcal{H}$ . Suppose that*

$$V + \text{cone}(z - \text{dom } f) \text{ is a closed linear subspace,} \quad (26.19)$$

and consider the problem

$$\underset{x \in z + V}{\text{minimize}} \quad f(x). \quad (26.20)$$

Then  $\bar{x}$  is a solution to (26.20) if and only if

$$\bar{x} - z \in V \quad \text{and} \quad (\exists u \in \partial f(\bar{x})) \quad u \perp V. \quad (26.21)$$

Moreover, if  $f$  is Gâteaux differentiable at  $\bar{x}$ , (26.21) becomes

$$\bar{x} - z \in V \quad \text{and} \quad \nabla f(\bar{x}) \perp V. \quad (26.22)$$

*Proof.* Set  $C = z + V$ . We have  $C - \text{dom } f = V + (z - \text{dom } f)$ . Hence, it follows from Proposition 6.19(iii) that  $0 \in \text{sri}(V + (z - \text{dom } f)) = \text{sri}(C - \text{dom } f)$ . Thus, (26.19) implies that (a) in Proposition 26.5 is satisfied. Therefore, the characterizations (26.21) and (26.22) follow from Example 6.42 and, respectively, from the equivalences (i) $\Leftrightarrow$ (v) and (i) $\Leftrightarrow$ (vi) in Proposition 26.5.  $\square$

## 26.4 Cone Constraints

We consider the problem of minimizing a convex function over the inverse linear image of a convex cone.

**Proposition 26.14** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $K$  be a closed convex cone in  $\mathcal{K}$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , and let  $\bar{x} \in \mathcal{H}$ . Suppose that one of the following holds:*

- (a)  $K - \text{cone } L(\text{dom } f)$  is a closed linear subspace.
- (b)  $\mathcal{K}$  is finite-dimensional,  $K$  is polyhedral, and  $K \cap \text{ri } L(\text{dom } f) \neq \emptyset$ .
- (c)  $\mathcal{H}$  and  $\mathcal{K}$  are finite-dimensional,  $f$  is a polyhedral function,  $K$  is polyhedral, and  $K \cap L(\text{dom } f) \neq \emptyset$ .

Consider the problem

$$\underset{Lx \in K}{\text{minimize}} \quad f(x). \quad (26.23)$$

Then  $\bar{x}$  is a solution to (26.23) if and only if

$$L\bar{x} \in K \quad \text{and} \quad (\exists \bar{v} \in K^\ominus) \quad \begin{cases} -L^*\bar{v} \in \partial f(\bar{x}), \\ \langle \bar{x} \mid L^*\bar{v} \rangle = 0, \end{cases} \quad (26.24)$$

in which case  $\bar{v}$  is a Lagrange multiplier associated with  $\bar{x}$ , and  $\bar{x}$  solves the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \langle x \mid L^*\bar{v} \rangle. \quad (26.25)$$

Moreover, if  $f$  is Gâteaux differentiable at  $\bar{x}$ , then (26.24) becomes

$$L\bar{x} \in K \quad \text{and} \quad (\exists \bar{v} \in K^\ominus) \quad \begin{cases} \nabla f(\bar{x}) = -L^*\bar{v}, \\ \langle \bar{x} \mid L^*\bar{v} \rangle = 0. \end{cases} \quad (26.26)$$

*Proof.* Set  $g = \iota_K$ . Then Problem (26.23) is a special case of (26.2). Using Proposition 6.19(iii) for (a), we note that conditions (a)–(c) imply their counterparts in Theorem 26.2. In turn, we derive from Example 16.12 and Example 6.39 that the characterization (26.24) follows from the equivalence (i)  $\Leftrightarrow$  (iii) in Theorem 26.2, from which we deduce the characterization (26.26) via Proposition 17.26(i). Finally, it follows from Proposition 26.1 that  $-L^*\bar{v} \in \partial f(\bar{x}) \Rightarrow 0 \in \partial(f + \langle \cdot \mid L^*\bar{v} \rangle)(\bar{x}) \Rightarrow \bar{x}$  solves (26.25). Hence, if  $\bar{x}$  solves (26.23), we derive from (26.24), Proposition 19.23(v), and Remark 19.24 that  $\bar{v}$  is a Lagrange multiplier associated with  $\bar{x}$ .  $\square$

**Example 26.15** Let  $M$  and  $N$  be strictly positive integers, let  $A \in \mathbb{R}^{M \times N}$ , let  $f \in \Gamma_0(\mathbb{R}^N)$ , and let  $\bar{x} \in \mathbb{R}^N$ . Suppose that  $\mathbb{R}_+^M \cap \text{ri } A(\text{dom } f) \neq \emptyset$  and consider the problem

$$\underset{Ax \in \mathbb{R}_+^M}{\text{minimize}} \quad f(x). \quad (26.27)$$

Then  $\bar{x}$  is a solution to (26.27) if and only if

$$A\bar{x} \in \mathbb{R}_+^M \quad \text{and} \quad (\exists \bar{v} \in \mathbb{R}_-^M) \quad \begin{cases} -A^\top \bar{v} \in \partial f(\bar{x}), \\ \langle \bar{x} \mid A^\top \bar{v} \rangle = 0, \end{cases} \quad (26.28)$$

in which case  $\bar{v}$  is a Lagrange multiplier associated with  $\bar{x}$ , and  $\bar{x}$  solves the problem

$$\underset{x \in \mathbb{R}^N}{\text{minimize}} \quad f(x) + \langle x \mid A^\top \bar{v} \rangle. \quad (26.29)$$

*Proof.* Suppose that  $\mathcal{H} = \mathbb{R}^N$ ,  $K = \mathbb{R}_+^M$ , and  $L = A$ . Then  $K$  is a closed convex polyhedral cone and condition (b) in Proposition 26.14 is satisfied. Hence, the result follows from Proposition 26.14.  $\square$

**Corollary 26.16** *Let  $K$  be a closed convex cone in  $\mathcal{H}$ , let  $f \in \Gamma_0(\mathcal{H})$ , and let  $\bar{x} \in \mathcal{H}$ . Suppose that one of the following holds:*

- (a)  $K - \text{cone}(\text{dom } f)$  is a closed linear subspace.
- (b)  $\mathcal{H}$  is finite-dimensional,  $K$  is polyhedral, and  $K \cap \text{ri dom } f \neq \emptyset$ .
- (c)  $\mathcal{H}$  is finite-dimensional,  $f$  is a polyhedral function,  $K$  is polyhedral, and  $K \cap \text{dom } f \neq \emptyset$ .

Consider the problem

$$\underset{x \in K}{\text{minimize}} \quad f(x). \quad (26.30)$$

Then  $\bar{x}$  is a solution to (26.30) if and only if

$$\bar{x} \in K \quad \text{and} \quad (\exists u \in K^\oplus \cap \partial f(\bar{x})) \quad \langle \bar{x} \mid u \rangle = 0. \quad (26.31)$$

*Proof.* Apply Proposition 26.14 to  $\mathcal{K} = \mathcal{H}$  and  $L = \text{Id}$ .  $\square$

## 26.5 Convex Inequality Constraints

In this section, we turn our attention to the minimization of a convex function subject to convex inequality constraints. The following result will be required.

**Lemma 26.17** *Let  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be a convex function such that  $\text{lev}_{<0} g \neq \emptyset$ , set  $C = \text{lev}_{\leq 0} g$ , and let  $x \in C$ . Then*

$$N_C x = \begin{cases} N_{\text{dom } g} x \cup \text{cone } \partial g(x), & \text{if } g(x) = 0; \\ N_{\text{dom } g} x, & \text{if } g(x) < 0. \end{cases} \quad (26.32)$$

*Proof.* Clearly,  $C \subset \text{dom } g$ . Hence,

$$\begin{aligned} (\forall u \in \mathcal{H}) \quad u \in N_{\text{dom } g} x &\Leftrightarrow \sup \langle \text{dom } g - x \mid u \rangle \leq 0 \\ &\Rightarrow \sup \langle C - x \mid u \rangle \leq 0 \\ &\Leftrightarrow u \in N_C x. \end{aligned} \quad (26.33)$$

Thus,

$$N_{\text{dom } g} x \subset N_C x. \quad (26.34)$$

We consider two cases.

(a)  $g(x) < 0$ : Take  $u \in N_C x$  and fix  $y \in \text{dom } g$ . Then  $\sup \langle C - x \mid u \rangle \leq 0$ . Moreover, for  $\alpha \in ]0, 1[$  sufficiently small,  $g((1 - \alpha)x + \alpha y) \leq (1 - \alpha)g(x) +$

$\alpha g(y) \leq 0$  and therefore  $\alpha \langle y - x \mid u \rangle = \langle ((1 - \alpha)x + \alpha y) - x \mid u \rangle \leq 0$ . Thus  $\langle y - x \mid u \rangle \leq 0$  and it follows that  $u \in N_{\text{dom } g} x$ . In view of (26.34), we obtain  $N_C x = N_{\text{dom } g} x$ .

(b)  $g(x) = 0$ : Set

$$K = \text{cone}(\text{lev}_{<0} g - x). \quad (26.35)$$

Then  $K$  is a nonempty convex cone and, using Proposition 6.23(iii) and Corollary 9.11, we deduce that

$$N_C x = (C - x)^\ominus = (\overline{C} - x)^\ominus = (\overline{\text{lev}_{<0} g} - x)^\ominus = (\text{lev}_{<0} g - x)^\ominus = K^\ominus. \quad (26.36)$$

We claim that

$$K = \{y \in \mathcal{H} \mid g'(x; y) < 0\}. \quad (26.37)$$

Let  $y \in \mathcal{H}$ . If  $y \in K$ , then  $y = \alpha(z - x)$ , for some  $z \in \text{lev}_{<0} g$  and some  $\alpha \in \mathbb{R}_{++}$ . Hence,  $g'(x; y) = g'(x; \alpha(z - x)) = \alpha g'(x; z - x) \leq \alpha(g(z) - g(x)) = \alpha g(z) < 0$  by Proposition 17.2(iii)&(iv). Conversely, if  $g'(x; y) < 0$  then, for some  $\alpha \in ]0, 1[$  sufficiently small, we have  $(g(x + \alpha y) - g(x))/\alpha = g'(x; \alpha y) < 0$  and hence  $y = ((x + \alpha y) - x)/\alpha \in K$ . This verifies (26.37). We now set

$$h = g'(x; \cdot). \quad (26.38)$$

Then  $h$  is sublinear and  $h(0) = 0$  by Proposition 17.2(iv). Moreover,  $h^* = \iota_{\partial g(x)}$  by Proposition 17.18. We consider two cases.

(b.1)  $x \in \text{dom } \partial g$ : Then  $\text{dom } h^* = \partial g(x) \neq \emptyset$  and hence Proposition 13.39 implies that  $h^{**} = \check{h}$ . Corollary 9.11 thus yields

$$\overline{K} = \overline{\text{lev}_{<0} h} = \text{lev}_{\leq 0} \check{h} = \text{lev}_{\leq 0} h^{**} = \text{lev}_{\leq 0} \iota_{\partial g(x)}^* = \text{lev}_{\leq 0} \sigma_{\partial g(x)} = (\partial g(x))^\ominus. \quad (26.39)$$

In view of (26.36), Proposition 6.23(iii), and Proposition 6.32, we deduce that

$$N_C x = K^\ominus = \overline{K}^\ominus = (\partial g(x))^{\ominus\ominus} = \overline{\text{cone}} \partial g(x). \quad (26.40)$$

On the other hand,  $\partial g(x)$  is nonempty by assumption, and closed and convex by Proposition 16.3(iii). Moreover, since  $\text{lev}_{<0} g \neq \emptyset$  and  $g(x) = 0$  by assumption,  $x \notin \text{Argmin } g$  and Theorem 16.2 yields  $0 \notin \partial g(x)$ . Consequently, Corollary 6.52 implies that  $\overline{\text{cone}} \partial g(x) = (\text{cone } \partial g(x)) \cup (\text{rec } \partial g(x))$ . However, using successively Proposition 16.4, Theorem 21.2, Proposition 21.14, Corollary 16.29, and Proposition 13.40(i), we obtain  $\text{rec } \partial g(x) = \text{rec } \partial g^{**}(x) = N_{\overline{\text{dom}} \partial g^{**}} x = N_{\overline{\text{dom}} g^{**}} x = N_{\overline{\text{dom}} g} x = N_{\text{dom } g} x$ . Altogether,

$$N_C x = \overline{\text{cone}} \partial g(x) = (\text{cone } \partial g(x)) \cup N_{\text{dom } g} x, \quad (26.41)$$

as announced in (26.32).

(b.2)  $x \notin \text{dom } \partial g$ : If  $\check{h}$  is proper, then so is  $(\check{h})^* = h^*$  by Proposition 13.14(iv) and Theorem 13.32. Hence, since  $h^* = \iota_{\partial g(x)} = \iota_\emptyset \equiv +\infty$  is not proper, neither is  $\check{h}$ . In view of Proposition 9.6,  $\check{h}$  therefore takes on



only the values  $-\infty$  and  $+\infty$ . Using Proposition 9.8(iv), we thus see that  $\text{lev}_{\leq 0} \check{h} = \text{dom } \check{h} = \overline{\text{dom } h}$ . On the other hand,  $\text{dom } h = \text{cone}(\text{dom } g - x)$  by Proposition 17.2(v) and hence  $\overline{\text{dom } h} = \overline{\text{cone}(\text{dom } g - x)}$ . It follows from (26.37), (26.38), and Corollary 9.11 that

$$\overline{K} = \overline{\text{lev}_{< 0} h} = \text{lev}_{\leq 0} \check{h} = \overline{\text{dom } h} = \overline{\text{cone}(\text{dom } g - x)}. \quad (26.42)$$

Therefore, using (26.36) and Proposition 6.23(iii), we conclude that

$$N_C x = K^\ominus = (\overline{\text{cone}(\text{dom } g - x)})^\ominus = (\text{dom } g - x)^\ominus = N_{\text{dom } g} x. \quad (26.43)$$

The entire lemma is proven.  $\square$

The minimization of a convex function under convex inequality constraints was already examined in Corollary 19.28. We investigate it here in a new light, and provide in particular a sufficient condition for the existence of Lagrange multipliers.

**Proposition 26.18** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , suppose that  $(g_i)_{i \in I}$  are functions in  $\Gamma_0(\mathcal{H})$  such that the Slater condition*

$$\begin{cases} (\forall i \in I) & \text{lev}_{\leq 0} g_i \subset \text{int dom } g_i, \\ \text{dom } f \cap \bigcap_{i \in I} \text{lev}_{< 0} g_i \neq \emptyset, \end{cases} \quad (26.44)$$

*is satisfied, and let  $\bar{x} \in \mathcal{H}$ . Consider the problem*

$$\begin{aligned} & \underset{\substack{x \in \mathcal{H} \\ g_1(x) \leq 0, \dots, g_m(x) \leq 0}}{\text{minimize}} & f(x). \end{aligned} \quad (26.45)$$

*Then  $\bar{x}$  is a solution to (26.45) if and only if*

$$\begin{aligned} & \left( \exists (\bar{\nu}_i)_{i \in I} \in \mathbb{R}_+^m \right) \left( \exists (u_i)_{i \in I} \in \bigtimes_{i \in I} \partial g_i(\bar{x}) \right) \\ & - \sum_{i \in I} \bar{\nu}_i u_i \in \partial f(\bar{x}) \quad \text{and} \quad (\forall i \in I) \quad \begin{cases} g_i(\bar{x}) \leq 0, \\ \bar{\nu}_i g_i(\bar{x}) = 0, \end{cases} \end{aligned} \quad (26.46)$$

*in which case  $(\bar{\nu}_i)_{i \in I}$  are Lagrange multipliers associated with  $\bar{x}$ , and  $\bar{x}$  solves the problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \sum_{i \in I} \bar{\nu}_i g_i(x). \quad (26.47)$$

*Moreover, if the functions  $f$  and  $(g_i)_{i \in I}$  are Gâteaux differentiable at  $\bar{x}$ , then (26.46) becomes*

$$(\exists (\bar{\nu}_i)_{i \in I} \in \mathbb{R}_+^m) \quad \begin{cases} \nabla f(\bar{x}) = - \sum_{i \in I} \bar{\nu}_i \nabla g_i(\bar{x}), \\ (\forall i \in I) \quad \begin{cases} g_i(\bar{x}) \leq 0, \\ \bar{\nu}_i g_i(\bar{x}) = 0. \end{cases} \end{cases} \quad (26.48)$$

*Proof.* Set

$$C = \bigcap_{i \in I} C_i, \quad \text{where} \quad (\forall i \in I) \quad C_i = \text{lev}_{\leq 0} g_i. \quad (26.49)$$

Corollary 8.30(ii) states that  $(\forall i \in I) \text{ int dom } g_i = \text{cont } g_i$ . Hence, it follows from (26.44) and Corollary 8.38(i) that  $\text{int } C = \bigcap_{i \in I} \text{int } C_i = \bigcap_{i \in I} \text{lev}_{< 0} g_i$  and that  $0 \in \text{int}(C - \text{dom } f) \subset \text{sri}(C - \text{dom } f)$ . Therefore, Proposition 26.5 asserts that

$$\bar{x} \in \text{Argmin}_C f \quad \Leftrightarrow \quad (\exists u \in N_C \bar{x}) \quad -u \in \partial f(\bar{x}). \quad (26.50)$$

Now suppose that  $\bar{x} \in C$ . To determine  $N_C \bar{x}$ , set

$$I_+ = \{i \in I \mid g_i(\bar{x}) = 0\} \quad \text{and} \quad I_- = \{i \in I \mid g_i(\bar{x}) < 0\}. \quad (26.51)$$

As seen above,

$$\bigcap_{i \in I} \text{int dom } \iota_{C_i} = \bigcap_{i \in I} \text{int } C_i = \bigcap_{i \in I} \text{lev}_{< 0} g_i \neq \emptyset. \quad (26.52)$$

On the other hand, (26.44) yields  $\bar{x} \in \bigcap_{i \in I} \text{int dom } g_i$ , which implies, by Proposition 6.43(ii), that

$$(\forall i \in I) \quad N_{\text{dom } g_i} \bar{x} = \{0\}. \quad (26.53)$$

Hence, Example 16.12, (26.49), (26.52), Corollary 16.39(iv), and Lemma 26.17 yield

$$\begin{aligned} N_C \bar{x} &= \partial \iota_{\bigcap_{i \in I} C_i}(\bar{x}) \\ &= \partial \left( \sum_{i \in I} \iota_{C_i} \right)(\bar{x}) \\ &= \sum_{i \in I} \partial \iota_{C_i}(\bar{x}) \\ &= \sum_{i \in I_+} N_{C_i} \bar{x} + \sum_{i \in I_-} N_{C_i} \bar{x} \\ &= \sum_{i \in I_+} \bigcup_{\nu_i \in \mathbb{R}_{++}} \nu_i \partial g_i(\bar{x}). \end{aligned} \quad (26.54)$$

Thus,  $N_C \bar{x}$  consists of all vectors of the form  $\sum_{i \in I_+} \nu_i u_i$ , where  $(\forall i \in I_+) \nu_i \in \mathbb{R}_{++}$  and  $u_i \in \partial g_i(x)$ . Therefore, (26.50) implies (26.46) and, in turn, (26.48) via Proposition 17.26(i). This provides the announced characterizations.

Next, suppose that  $\bar{x}$  solves (26.45). Then we derive from (26.46), (26.44), (26.52), and Corollary 16.39(iv) that there exist  $(\bar{\nu}_i)_{i \in I} \in \mathbb{R}_+^m$  such that

$$0 \in \partial f(\bar{x}) + \sum_{i \in I} \bar{\nu}_i \partial g_i(\bar{x}) = \partial \left( f + \sum_{i \in I} \bar{\nu}_i g_i \right) (\bar{x}). \quad (26.55)$$

In view of Theorem 16.2, this shows that  $\bar{x}$  solves (26.47). In turn, we derive from (26.46), from Proposition 19.23(v) applied to  $\mathcal{K} = \mathbb{R}^m$ ,  $K = \mathbb{R}_+^m$ , and  $R: x \mapsto (g_i(x))_{i \in I}$ , and from Remark 19.24 that  $(\bar{\nu}_i)_{i \in I}$  are Lagrange multipliers associated with  $\bar{x}$ .  $\square$

**Remark 26.19** The characterization (26.46) is often referred to as the *Karush–Kuhn–Tucker conditions*.

## 26.6 Regularization of Minimization Problems

Let  $f \in \Gamma_0(\mathcal{H})$  and suppose that  $\text{Argmin } f \neq \emptyset$ , i.e., the minimization problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) \quad (26.56)$$

has at least one solution. In order to obtain a specific minimizer, one can introduce, for every  $\varepsilon \in ]0, 1[$ , the regularized problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \varepsilon g(x), \quad (26.57)$$

where  $g \in \Gamma_0(\mathcal{H})$ . The objective is to choose the regularization function  $g$  such that (26.57) admits a unique solution  $x_\varepsilon$  and such that the net  $(x_\varepsilon)_{\varepsilon \in ]0, 1[}$  converges to a specific point in  $\text{Argmin } f$ . For instance, when  $g = (1/2)\|\cdot\|^2$ , we obtain the classical *Tykhonov regularization* framework. In this case, it follows from Proposition 26.5 that (26.57) is equivalent to

$$\text{find } x_\varepsilon \in \mathcal{H} \quad \text{such that} \quad 0 \in \partial f(x_\varepsilon) + \varepsilon x_\varepsilon, \quad (26.58)$$

which is a special case of (23.35). In turn, if we denote by  $x_0$  the minimum norm minimizer of  $f$ , Theorem 23.44(i) asserts that  $x_\varepsilon \rightarrow x_0$  as  $\varepsilon \downarrow 0$ . The next theorem explores the asymptotic behavior of the curve  $(x_\varepsilon)_{\varepsilon \in ]0, 1[}$  for more general choices of the regularization function  $g$ .

**Theorem 26.20** *Let  $f$  and  $g$  be in  $\Gamma_0(\mathcal{H})$ . Suppose that  $\text{Argmin } f \cap \text{dom } g \neq \emptyset$  and that  $g$  is coercive and strictly convex. Then  $g$  admits a unique minimizer  $x_0$  over  $\text{Argmin } f$  and, for every  $\varepsilon \in ]0, 1[$ , the regularized problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad f(x) + \varepsilon g(x) \quad (26.59)$$

admits a unique solution  $x_\varepsilon$ . Moreover, the following hold:

- (i)  $x_\varepsilon \rightharpoonup x_0$  as  $\varepsilon \downarrow 0$ .
- (ii)  $g(x_\varepsilon) \rightarrow g(x_0)$  as  $\varepsilon \downarrow 0$ .
- (iii) Suppose that  $g$  is uniformly convex on every closed ball in  $\mathcal{H}$ . Then  $x_\varepsilon \rightarrow x_0$  as  $\varepsilon \downarrow 0$ .

*Proof.* Set  $S = \text{Argmin } f$ . Let  $\varepsilon \in ]0, 1[$  and set  $h_\varepsilon = f + \varepsilon g$ . Then  $h_\varepsilon \in \Gamma_0(\mathcal{H})$  and  $h_\varepsilon$  is strictly convex. Moreover, since  $\inf f(\mathcal{H}) > -\infty$ , existence and uniqueness of  $x_0$  and  $x_\varepsilon$  follow from Corollary 11.15(ii) and Corollary 11.8. Now fix  $z \in S \cap \text{dom } g$ , let  $(\varepsilon_n)_{n \in \mathbb{N}}$  be an arbitrary sequence in  $]0, 1[$  such that  $\varepsilon_n \downarrow 0$ , and set

$$(\forall n \in \mathbb{N}) \quad y_n = x_{\varepsilon_n}. \quad (26.60)$$

(i): We have

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad f(y_n) + \varepsilon_n g(y_n) &= \inf (f + \varepsilon_n g)(\mathcal{H}) \\ &\leq f(z) + \varepsilon_n g(z) \end{aligned} \quad (26.61)$$

$$\leq f(y_n) + \varepsilon_n g(z). \quad (26.62)$$

Therefore

$$(\forall n \in \mathbb{N}) \quad g(y_n) \leq g(z) < +\infty. \quad (26.63)$$

Accordingly,  $(y_n)_{n \in \mathbb{N}}$  lies in  $\text{lev}_{\leq g(z)} g$  and it follows from Proposition 11.11 that  $(y_n)_{n \in \mathbb{N}}$  is bounded. Furthermore,  $\inf g(\mathcal{H}) > -\infty$ . Thus, (26.61) implies that

$$\begin{aligned} (\forall n \in \mathbb{N}) \quad f(y_n) &\leq \inf f(\mathcal{H}) + \varepsilon_n (g(z) - g(y_n)) \\ &\leq \inf f(\mathcal{H}) + \varepsilon_n (g(z) - \inf g(\mathcal{H})). \end{aligned} \quad (26.64)$$

Thus, since  $\varepsilon_n \downarrow 0$ , we get

$$\overline{\lim} f(y_n) \leq \inf f(\mathcal{H}). \quad (26.65)$$

Now, let  $x$  be a weak sequential cluster point of  $(y_n)_{n \in \mathbb{N}}$ , say  $y_{k_n} \rightharpoonup x$ . Since, by Theorem 9.1,  $f$  is weakly lower semicontinuous, (26.65) yields

$$\inf f(\mathcal{H}) \leq f(x) \leq \underline{\lim} f(y_{k_n}) \leq \overline{\lim} f(y_{k_n}) \leq \inf f(\mathcal{H}). \quad (26.66)$$

Consequently,  $f(x) = \inf f(\mathcal{H})$ , i.e.,  $x \in S$ . Furthermore, it follows from (26.63) and the weak lower semicontinuity of  $g$  that  $g(x) \leq \underline{\lim} g(y_{k_n}) \leq \inf g(S)$ . In turn, since  $x \in S$ , we obtain  $g(x) = \inf g(S)$ , i.e.,  $x = x_0$ . Altogether,  $(y_n)_{n \in \mathbb{N}}$  is bounded and has  $x_0$  as its unique weak sequential cluster point. We therefore deduce from Lemma 2.38 that  $y_n \rightharpoonup x_0$ . Finally, since  $(\varepsilon_n)_{n \in \mathbb{N}}$  was chosen arbitrarily in  $]0, 1[$ , we conclude that  $x_\varepsilon \rightharpoonup x_0$  as  $\varepsilon \downarrow 0$ .

(ii): Since  $\underline{g}$  is weakly lower semicontinuous, (i) implies that  $g(x_0) \leq \underline{\lim} g(y_n) \leq \overline{\lim} g(y_n)$ . However, (26.63) yields  $\underline{\lim} g(y_n) \leq g(x_0)$ . Hence  $g(y_n) \rightarrow g(x_0)$  and, in turn,  $g(x_\varepsilon) \rightarrow g(x_0)$ .

(iii): Since  $(y_n)_{n \in \mathbb{N}}$  is bounded, there exists an increasing function  $\phi: \mathbb{R}_+ \rightarrow \mathbb{R}_+$  that vanishes only at 0 such that

$$(\forall n \in \mathbb{N}) \quad \frac{1}{4}\phi(\|y_n - x_0\|) \leq \frac{g(y_n) + g(x_0)}{2} - g\left(\frac{y_n + x_0}{2}\right). \quad (26.67)$$

Consequently, it follows from (ii) that

$$\begin{aligned} \frac{1}{4}\overline{\lim} \phi(\|y_n - x_0\|) &\leq \overline{\lim} \frac{g(y_n) + g(x_0)}{2} - \underline{\lim} g\left(\frac{y_n + x_0}{2}\right) \\ &= g(x_0) - \underline{\lim} g\left(\frac{y_n + x_0}{2}\right). \end{aligned} \quad (26.68)$$

However, since  $(y_n + x_0)/2 \rightarrow x_0$  by (ii),  $g(x_0) \leq \underline{\lim} g((y_n + x_0)/2)$ . Altogether  $\overline{\lim} \phi(\|y_n - x_0\|) \leq 0$ ; hence  $y_n \rightarrow x_0$ . We conclude that  $x_\varepsilon \rightarrow x_0$  as  $\varepsilon \downarrow 0$ .  $\square$

**Example 26.21** Let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } L$  is closed. Then the generalized inverse of  $L$  (see Definition 3.26) satisfies

$$(\forall y \in \mathcal{K}) \quad L^\dagger y = \lim_{\varepsilon \downarrow 0} (L^*L + \varepsilon \text{Id})^{-1} L^* y. \quad (26.69)$$

*Proof.* Let  $y \in \mathcal{K}$  and let  $\varepsilon \in \mathbb{R}_{++}$ . The solution to

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \frac{1}{2}\|Lx - y\|^2 + \frac{\varepsilon}{2}\|x\|^2 \quad (26.70)$$

is  $x_\varepsilon = (L^*L + \varepsilon \text{Id})^{-1} L^* y$ . However, by applying Theorem 26.20(iii) to  $f: x \mapsto (1/2)\|Lx - y\|^2$  and the strongly convex function  $g = (1/2)\|\cdot\|^2$ , we deduce that  $x_\varepsilon \rightarrow x_0$  as  $\varepsilon \downarrow 0$ , where  $x_0$  is the minimum norm minimizer of  $x \mapsto \|Lx - y\|^2/2$ . On the other hand, it follows from Definition 3.26 and Proposition 3.25 that  $x_0 = L^\dagger y$ . Altogether,  $(L^*L + \varepsilon \text{Id})^{-1} L^* y \rightarrow L^\dagger y$  as  $\varepsilon \downarrow 0$ .  $\square$

## Exercises

**Exercise 26.1** Let  $f: \mathcal{H} \rightarrow ]-\infty, +\infty]$  be proper and convex. Show that  $\text{Argmin } f \subset \partial f^*(0)$  and that  $\text{Argmin } f \neq \partial f^*(0)$  may occur. Compare to Proposition 26.1.

**Exercise 26.2** Provide two functions  $f$  and  $g$  in  $\Gamma_0(\mathcal{H})$  such that  $\text{dom } f \cap \text{dom } g \neq \emptyset$  and for which the conclusion of Corollary 26.3 fails.

**Exercise 26.3** Let  $F: \mathcal{H} \times \mathcal{H} \rightarrow \mathbb{R}$  be a symmetric bilinear form that satisfies (26.5) for some  $\alpha$  and  $\beta$  in  $\mathbb{R}_{++}$ . Derive the conclusions of the Stampacchia theorem (Example 26.8) from Theorem 3.14.

**Exercise 26.4** Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$  such that  $0 \in \text{sri}(C - D)$ . Show that

$$C \cap D = P_D(\text{Fix}(2P_C - \text{Id}) \circ (2P_D - \text{Id})). \quad (26.71)$$

**Exercise 26.5** Let  $F$  be as in Example 26.8. Show that there exists a unique operator  $L \in \mathcal{B}(\mathcal{H})$  such that  $(\forall x \in \mathcal{H})(\forall y \in \mathcal{H}) F(x, y) = \langle y | Lx \rangle$ . Moreover, show that  $\|L\| \leq \beta$ , that  $L$  is invertible, and that  $\|L^{-1}\| \leq \alpha^{-1}$ .

**Exercise 26.6** Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$  such that  $C \cap D \neq \emptyset$ . Use Proposition 26.5 to obtain the identity

$$C \cap D = \text{Fix } P_C \circ P_D. \quad (26.72)$$

**Exercise 26.7** Let  $u \in \mathcal{H} \setminus \{0\}$ , let  $\eta \in \mathbb{R}$ , set  $C = \{x \in \mathcal{H} \mid \langle x | u \rangle = \eta\}$ , and let  $g: \mathcal{H} \rightarrow \mathbb{R}$  be lower semicontinuous, strictly convex, and Gâteaux differentiable on  $\mathcal{H}$ . Suppose that  $\text{dom } \partial g^* = \text{int dom } g^*$  and that  $\text{Argmin}_C g \neq \emptyset$ . Show that there exists a unique  $\nu \in \mathbb{R}$  such that  $\langle \nabla g^*(\nu u) | u \rangle = \eta$  and that the unique element in  $\text{Argmin}_C g$  is  $\nabla g^*(\nu u)$ .

**Exercise 26.8** Let  $f \in \Gamma_0(\mathcal{H})$  and let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } L$  is closed. Suppose that  $0 \in \text{sri}(L(\text{dom } f))$ . Consider the problem

$$\underset{x \in \ker L}{\text{minimize}} \quad f(x), \quad (26.73)$$

and let  $\bar{x} \in \mathcal{H}$ . Derive from Proposition 26.11 that  $\bar{x}$  is a solution to (26.73) if and only if

$$L\bar{x} = 0 \quad \text{and} \quad (\exists \bar{v} \in \mathcal{K}) \quad -L^*\bar{v} \in \partial f(\bar{x}). \quad (26.74)$$

**Exercise 26.9** Let  $u \in \mathcal{H} \setminus \{0\}$ , let  $\eta \in \mathbb{R}$ , set  $C = \{x \in \mathcal{H} \mid \langle x | u \rangle = \eta\}$ , and let  $z \in \mathcal{H}$ . Use Corollary 26.12 to show that

$$P_C z = z + \frac{\eta - \langle z | u \rangle}{\|u\|^2} u. \quad (26.75)$$

**Exercise 26.10** Suppose that  $\mathcal{H} = \mathbb{R}^N$  and let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be convex and Gâteaux differentiable on  $\mathcal{H}$ . Set  $K = \mathbb{R}_+^N$  and let  $x \in \mathcal{H}$ . Show that  $x \in \text{Argmin}_K f$  if and only if  $x \in K$ ,  $\nabla f(x) \in K$ , and  $x \perp \nabla f(x)$ .

**Exercise 26.11** Provide an example of a proper and convex function  $g: \mathcal{H} \rightarrow ]-\infty, +\infty]$  such that  $\text{lev}_{<0} g = \emptyset$  and  $N_{C^*} x \neq N_{\text{dom } g} x \cup \text{cone } \partial g(x)$ , where  $C = \text{lev}_{\leq 0} g$  and  $x \in C$ . Compare to Lemma 26.17.

**Exercise 26.12** Let  $(y_n)_{n \in \mathbb{N}}$  be the sequence defined in the proof of Theorem 26.20. Show that  $(y_n)_{n \in \mathbb{N}}$  is a minimizing sequence of  $f$ .





# Chapter 27

## Proximal Minimization

As seen in Chapter 26, the solutions to variational problems can be characterized by fixed point equations involving proximity operators. Since proximity operators are firmly nonexpansive, they can be used to devise efficient algorithms to solve minimization problems. Such algorithms, called proximal algorithms, are investigated in this chapter.

Throughout this chapter,  $\mathcal{K}$  is a real Hilbert space.

### 27.1 The Proximal-Point Algorithm

As seen in Proposition 26.1, minimizing a function  $f \in \Gamma_0(\mathcal{H})$  amounts to finding a zero of its subdifferential operator  $\partial f$ , which is a maximally monotone operator with resolvent  $J_{\partial f} = \text{Prox}_f$ . Thus, a minimizer of  $f$  can be obtained via the proximal-point algorithm (23.31). In this vein, our first result is a refinement of Theorem 23.41 that features a more relaxed condition on the parameter sequence  $(\gamma_n)_{n \in \mathbb{N}}$ .

**Theorem 27.1 (proximal-point algorithm)** *Let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{Argmin } f \neq \emptyset$ , let  $(\gamma_n)_{n \in \mathbb{N}}$  be a sequence in  $\mathbb{R}_{++}$  such that  $\sum_{n \in \mathbb{N}} \gamma_n = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = \text{Prox}_{\gamma_n f} x_n. \quad (27.1)$$

*Then the following hold:*

- (i)  $(x_n)_{n \in \mathbb{N}}$  is a minimizing sequence of  $f$ ; more precisely,  $f(x_n) \downarrow \inf f(\mathcal{H})$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Argmin } f$ .
- (iii) Suppose that  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique point in  $\text{Argmin } f$ .

*Proof.* (i): Let  $z \in \text{Argmin } f$ . It follows from (27.1) and (16.30) that

$$(\forall n \in \mathbb{N}) \quad x_n - x_{n+1} \in \gamma_n \partial f(x_{n+1}). \quad (27.2)$$

In turn, we derive from (16.1) that

$$(\forall n \in \mathbb{N}) \quad \langle z - x_{n+1} \mid x_n - x_{n+1} \rangle / \gamma_n \leq f(z) - f(x_{n+1}) \quad (27.3)$$

and

$$(\forall n \in \mathbb{N}) \quad 0 \leq \langle x_n - x_{n+1} \mid x_n - x_{n+1} \rangle / \gamma_n \leq f(x_n) - f(x_{n+1}). \quad (27.4)$$

Hence, for every  $n \in \mathbb{N}$ , (27.3) yields

$$\begin{aligned} \|x_{n+1} - z\|^2 &= \|x_n - z\|^2 + 2 \langle x_n - z \mid x_{n+1} - x_n \rangle + \|x_{n+1} - x_n\|^2 \\ &= \|x_n - z\|^2 - \|x_{n+1} - x_n\|^2 + 2 \langle x_{n+1} - z \mid x_{n+1} - x_n \rangle \\ &\leq \|x_n - z\|^2 - 2\gamma_n (f(x_{n+1}) - \inf f(\mathcal{H})). \end{aligned} \quad (27.5)$$

This shows that  $(x_n)_{n \in \mathbb{N}}$  is Fejér monotone with respect to  $\operatorname{Argmin} f$  and that  $\sum_{n \in \mathbb{N}} \gamma_n (f(x_{n+1}) - \inf f(\mathcal{H})) < +\infty$ . Hence, since  $\sum_{n \in \mathbb{N}} \gamma_n = +\infty$ , we have  $\varliminf f(x_n) = \inf f(\mathcal{H})$  and it follows from (27.4) that  $f(x_n) \downarrow \inf f(\mathcal{H})$ .

(ii): Let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ . It follows from (i) and Proposition 11.20 that  $x \in \operatorname{Argmin} f$ . Now apply Theorem 5.5.

(iii): It follows from (ii) and Theorem 16.2 that there exists  $x \in \operatorname{Argmin} f \subset \operatorname{dom} \partial f$  such that  $x_n \rightharpoonup x$ , and from (27.5) and (27.2) that  $(x_{n+1})_{n \in \mathbb{N}}$  is a bounded sequence in  $\operatorname{dom} \partial f$ . Hence,  $\{x\} \cup \{x_{n+1}\}_{n \in \mathbb{N}}$  is a bounded subset of  $\operatorname{dom} \partial f$  and we derive from (10.2) that there exists an increasing function  $\phi: \mathbb{R}_+ \rightarrow [0, +\infty]$  vanishing only at 0 such that

$$(\forall n \in \mathbb{N} \setminus \{0\}) \quad \frac{1}{4} \phi(\|x_n - x\|) \leq \frac{f(x_n) + f(x)}{2} - f\left(\frac{x_n + x}{2}\right). \quad (27.6)$$

By (i), we have  $f(x_n) \downarrow f(x)$ . In addition, since, by Proposition 10.23,  $f$  is weakly sequentially lower semicontinuous and since  $(x_n + x)/2 \rightharpoonup x$ , we have  $f(x) \leq \varliminf f((x_n + x)/2) \leq \overline{\lim} f((x_n + x)/2) \leq \overline{\lim} (f(x_n) + f(x))/2 = f(x)$ . Altogether,  $\phi(\|x_n - x\|) \rightarrow 0$  and we conclude that  $x_n \rightarrow x$ .  $\square$

## 27.2 Douglas–Rachford Algorithm

In this section we apply the results of Theorem 25.6 on the Douglas–Rachford splitting algorithm to the minimization of the sum of two functions in  $\Gamma_0(\mathcal{H})$ . For this purpose, the following proposition will be useful.

**Proposition 27.2** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that one of the following holds:*

- (i)  $\text{Argmin}(f + g) \neq \emptyset$  (see Corollary 11.15 for sufficient conditions) and  $0 \in \text{sri}(\text{dom } f - \text{dom } g)$  (see Proposition 6.19 for sufficient conditions).
- (ii)  $\text{Argmin}(f + g) \subset \text{Argmin } f \cap \text{Argmin } g \neq \emptyset$ .
- (iii)  $f = \iota_C$  and  $g = \iota_D$ , where  $C$  and  $D$  are closed convex subsets of  $\mathcal{H}$  such that  $C \cap D \neq \emptyset$ .

Then  $\text{Argmin}(f + g) = \text{zer}(\partial f + \partial g) \neq \emptyset$ .

*Proof.* Let  $x \in \mathcal{H}$ . Using Theorem 16.2 and Proposition 16.5(ii), we have  $x \in \text{Argmin } f \cap \text{Argmin } g \Leftrightarrow 0 \in \partial f(x) \cap \partial g(x) \Rightarrow 0 \in \partial f(x) + \partial g(x) \Rightarrow 0 \in \partial(f + g)(x) \Leftrightarrow x \in \text{Argmin}(f + g)$ . Hence,

$$\text{Argmin } f \cap \text{Argmin } g \subset \text{zer}(\partial f + \partial g) \subset \text{zer } \partial(f + g) = \text{Argmin}(f + g). \quad (27.7)$$

(i): By Corollary 16.38(i),  $\partial(f + g) = \partial f + \partial g$ . In view of (27.7), we obtain  $\text{zer}(\partial f + \partial g) = \text{Argmin}(f + g)$ .

(ii): Clear from (27.7).

(iii) $\Rightarrow$ (ii):  $\text{Argmin}(\iota_C + \iota_D) = C \cap D = \text{Argmin } \iota_C \cap \text{Argmin } \iota_D$ .  $\square$

**Remark 27.3** It follows from Example 16.40 that the implication (iii) $\Rightarrow$ (i) fails.

**Corollary 27.4 (Douglas–Rachford algorithm)** *Let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$  such that (see Proposition 27.2 for sufficient conditions)*

$$\text{zer}(\partial f + \partial g) \neq \emptyset, \quad (27.8)$$

*let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = \text{Prox}_{\gamma g} x_n, \\ z_n = \text{Prox}_{\gamma f}(2y_n - x_n), \\ x_{n+1} = x_n + \lambda_n(z_n - y_n). \end{cases} \quad (27.9)$$

*Then there exists  $x \in \mathcal{H}$  such that the following hold:*

- (i)  $\text{Prox}_{\gamma g} x \in \text{Argmin}(f + g)$ .
- (ii)  $(y_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (iii)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to  $x$ .
- (iv)  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to  $\text{Prox}_{\gamma g} x$ .
- (v) Suppose that one of the following holds:
  - (a)  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ .
  - (b)  $g$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial g$ .

*Then  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to  $\text{Prox}_{\gamma g} x$ , which is the unique minimizer of  $f + g$ .*

*Proof.* Set  $A = \partial f$  and  $B = \partial g$ . Then, by Theorem 20.40,  $A$  and  $B$  are maximally monotone. In addition, we deduce from Proposition 16.5(ii) that  $\text{zer}(A+B) \subset \text{zer } \partial(f+g) = \text{Argmin}(f+g)$ . The results therefore follow from Example 23.3, Example 22.4, and Theorem 25.6(i)–(v)&(vii).  $\square$

**Corollary 27.5** *Let  $f \in \Gamma_0(\mathcal{H})$  and  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ . Suppose that  $LL^* = \mu \text{Id}$  for some  $\mu \in \mathbb{R}_{++}$ , that*

$$r \in \text{sri } L(\text{dom } f), \quad (27.10)$$

*and that the problem*

$$\underset{\substack{x \in \mathcal{H} \\ Lx=r}}{\text{minimize}} \quad f(x) \quad (27.11)$$

*has at least one solution. Let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} p_n = x_n + \mu^{-1}L^*(r - Lx_n), \\ y_n = \text{Prox}_{\gamma f} x_n, \\ q_n = y_n + \mu^{-1}L^*(r - Ly_n), \\ x_{n+1} = x_n + \lambda_n(2q_n - p_n - y_n). \end{cases} \quad (27.12)$$

*Then  $(p_n)_{n \in \mathbb{N}}$  converges weakly to a solution to (27.11).*

*Proof.* Set  $C = \{x \in \mathcal{H} \mid Lx = r\}$ ,  $A = N_C$ , and  $B = \partial f$ . It follows from Example 20.41 and Theorem 20.40 that  $A$  and  $B$  are maximally monotone. Moreover, since  $\text{ran } LL^* = \text{ran } \mu \text{Id} = \mathcal{K}$ , we derive from Fact 2.19 that  $\text{ran } L^*$  is closed. In turn, Example 6.42 and Fact 2.18(iv) yield

$$(\forall x \in \mathcal{H}) \quad N_C x = \begin{cases} (\ker L)^\perp = \text{ran } L^*, & \text{if } Lx = r; \\ \emptyset, & \text{if } Lx \neq r. \end{cases} \quad (27.13)$$

Next, we note that the set of solutions to (27.11) is  $\text{Argmin}_C f$ . However, it follows from (27.10), Proposition 26.11, and (27.13) that

$$\text{Argmin}_C f = \{x \in C \mid (\exists v \in \mathcal{K}) -L^*v \in \partial f(x)\} = \text{zer}(A+B). \quad (27.14)$$

On the other hand, we derive from Example 23.4 and Proposition 23.32 that

$$J_{\gamma A} = P_C = \text{Prox}_{\iota_{\{r\}} \circ L} = \text{Id} - \mu^{-1}L^* \circ (\text{Id} - P_{\{r\}}) \circ L = \text{Id} - \mu^{-1}L^* \circ (L - r). \quad (27.15)$$

Therefore, since  $J_{\gamma B} = \text{Prox}_{\gamma f}$  by Example 23.3 and since  $J_{\gamma A} = P_C$  is affine by Corollary 3.20(ii), we derive from (27.12) that

$$(\forall n \in \mathbb{N}) \quad \begin{cases} p_n = J_{\gamma A} x_n, \\ y_n = J_{\gamma B} x_n, \\ q_n = J_{\gamma A} y_n, \\ x_{n+1} = x_n + \lambda_n(J_{\gamma A}(2y_n - x_n) - y_n). \end{cases} \quad (27.16)$$

Thus, we recover (25.9), and the result follows from Theorem 25.6(vi).  $\square$

**Example 27.6** Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , let  $(f_i)_{i \in I}$  be functions in  $\Gamma_0(\mathcal{H})$ , let  $r \in \mathcal{H}$ , and suppose that the problem

$$\underset{\substack{x_1 \in \mathcal{H}, \dots, x_m \in \mathcal{H} \\ \sum_{i \in I} x_i = r}}{\text{minimize}} \sum_{i \in I} f_i(x_i) \quad (27.17)$$

has at least one solution and that

$$r \in \text{sri} \left\{ \sum_{i \in I} x_i \mid (\forall i \in I) \ x_i \in \text{dom } f_i \right\}. \quad (27.18)$$

Let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $(x_{i,0})_{i \in I} \in \mathcal{H}^m$ . Set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} u_n = \frac{1}{m} \left( r - \sum_{j \in I} x_{j,n} \right), \\ (\forall i \in I) \quad \begin{cases} p_{i,n} = x_{i,n} + u_n, \\ y_{i,n} = \text{Prox}_{\gamma f_i} x_{i,n}, \end{cases} \\ v_n = \frac{1}{m} \left( r - \sum_{j \in I} y_{j,n} \right), \\ (\forall i \in I) \ x_{i,n+1} = x_{i,n} + \lambda_n(y_{i,n} - x_{i,n} + 2v_n - u_n). \end{cases} \quad (27.19)$$

Then, for every  $i \in I$ ,  $(p_{i,n})_{n \in \mathbb{N}}$  converges weakly to a point  $p_i \in \mathcal{H}$ , and  $(p_i)_{i \in I}$  is a solution to (27.17).

*Proof.* Set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}$ ,  $\mathbf{f} = \bigoplus_{i \in I} f_i$ ,  $\mathcal{K} = \mathcal{H}$ , and  $\mathbf{L}: \mathcal{H} \rightarrow \mathcal{K}: \mathbf{x} \mapsto \sum_{i \in I} x_i$ , where  $\mathbf{x} = (x_i)_{i \in I}$  denotes a generic element in  $\mathcal{H}$ . Then (27.17) becomes

$$\underset{\substack{\mathbf{x} \in \mathcal{H} \\ \mathbf{L}\mathbf{x} = r}}{\text{minimize}} \mathbf{f}(\mathbf{x}). \quad (27.20)$$

Moreover, the assumptions imply that this problem admits at least one solution and that  $r \in \text{sri } \mathbf{L}(\text{dom } \mathbf{f})$ . On the other hand,  $\mathbf{L}^*: \mathcal{K} \rightarrow \mathcal{H}: x \mapsto (x, \dots, x)$  and Proposition 23.30 yields

$$\text{Prox}_{\gamma \mathbf{f}}: \mathcal{H} \rightarrow \mathcal{H}: \mathbf{x} \mapsto (\text{Prox}_{\gamma f_i} x_i)_{i \in I}. \quad (27.21)$$

Hence upon setting  $\mu = m$  and, for every  $n \in \mathbb{N}$ ,  $\mathbf{x}_n = (x_{i,n})_{i \in I}$ ,  $\mathbf{p}_n = (p_{i,n})_{i \in I}$ ,  $\mathbf{y}_n = (y_{i,n})_{i \in I}$ , and  $\mathbf{q}_n = (y_{i,n} + v_n)_{i \in I}$ , we can rewrite (27.19) as

$$(\forall n \in \mathbb{N}) \quad \begin{cases} \mathbf{p}_n = \mathbf{x}_n + \mu^{-1} \mathbf{L}^*(r - \mathbf{L}\mathbf{x}_n), \\ \mathbf{y}_n = \text{Prox}_{\gamma f} \mathbf{x}_n, \\ \mathbf{q}_n = \mathbf{y}_n + \mu^{-1} \mathbf{L}^*(r - \mathbf{L}\mathbf{y}_n), \\ \mathbf{x}_{n+1} = \mathbf{x}_n + \lambda_n(2\mathbf{q}_n - \mathbf{p}_n - \mathbf{y}_n), \end{cases} \quad (27.22)$$

which is precisely (27.12). Since Corollary 27.5 asserts that  $(\mathbf{p}_n)_{n \in \mathbb{N}}$  converges weakly to a solution to (27.11), we obtain the announced result.  $\square$

The next result focuses on the finite-dimensional setting.

**Corollary 27.7** *Suppose that  $\mathcal{H}$  is finite-dimensional, let  $f$  and  $g$  be functions in  $\Gamma_0(\mathcal{H})$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $x_0 \in \mathcal{H}$ . Suppose that  $\text{Argmin}(f + g) \neq \emptyset$  and that one of the following holds:*

- (a)  $(\text{ri dom } f) \cap (\text{ri dom } g) \neq \emptyset$ .
- (b)  $g$  is polyhedral and  $\text{dom } g \cap \text{ri dom } f \neq \emptyset$ .
- (c)  $f$  and  $g$  are polyhedral, and  $\text{dom } f \cap \text{dom } g \neq \emptyset$ .

Set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = \text{Prox}_{\gamma g} x_n, \\ z_n = \text{Prox}_{\gamma f}(2y_n - x_n), \\ x_{n+1} = x_n + \lambda_n(z_n - y_n). \end{cases} \quad (27.23)$$

Then there exists  $x \in \mathcal{H}$  such that the following hold:

- (i)  $\text{Prox}_{\gamma g} x \in \text{Argmin}(f + g)$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges to  $x$ .
- (iii)  $(y_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge to  $\text{Prox}_{\gamma g} x$ .

*Proof.* Set  $A = \partial f$  and  $B = \partial g$ . Theorem 20.40 asserts that  $A$  and  $B$  are maximally monotone. Moreover, by the equivalence (i)  $\Leftrightarrow$  (ii) in Corollary 26.3,  $\text{Argmin}(f + g) = \text{zer}(A + B)$ . The results therefore follow from Example 23.3 and Theorem 25.6(i)&(iii)–(v).  $\square$

We now describe a parallel splitting algorithm for minimizing a finite sum of functions in  $\Gamma_0(\mathcal{H})$ .

**Proposition 27.8 (parallel splitting algorithm)** *Let  $m$  be an integer such that  $m \geq 2$ , set  $I = \{1, \dots, m\}$ , and let  $(f_i)_{i \in I}$  be functions in  $\Gamma_0(\mathcal{H})$ . Suppose that the problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \sum_{i \in I} f_i(x) \quad (27.24)$$

*has at least one solution and that one of the following holds:*

- (i)  $0 \in \bigcap_{i=2}^m \text{sri}(\text{dom } f_i - \bigcap_{j=1}^{i-1} \text{dom } f_j)$ .
- (ii)  $\text{dom } f_1 \cap \bigcap_{i=2}^m \text{int dom } f_i \neq \emptyset$ .
- (iii)  $\mathcal{H}$  is finite-dimensional and  $\bigcap_{i \in I} \text{ri dom } f_i \neq \emptyset$ .

Let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(2 - \lambda_n) = +\infty$ , let  $\gamma \in \mathbb{R}_{++}$ , and let  $(x_{i,0})_{i \in I} \in \mathcal{H}^m$ . Set

$$(\forall n \in \mathbb{N}) \quad \left\{ \begin{array}{l} p_n = \frac{1}{m} \sum_{i \in I} x_{i,n}, \\ (\forall i \in I) \quad y_{i,n} = \text{Prox}_{\gamma f_i} x_{i,n}, \\ q_n = \frac{1}{m} \sum_{i \in I} y_{i,n}, \\ (\forall i \in I) \quad x_{i,n+1} = x_{i,n} + \lambda_n(2q_n - p_n - y_{i,n}). \end{array} \right. \quad (27.25)$$

Then  $(p_n)_{n \in \mathbb{N}}$  converges weakly to a solution to (27.24).

*Proof.* Set  $(\forall i \in I) \quad A_i = \partial f_i$  in Proposition 25.7. Then it follows from Example 23.3 that  $(p_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{zer} \sum_{i \in I} \partial f_i$ . However, Corollary 16.39 and Theorem 16.2 yield  $\text{zer} \sum_{i \in I} \partial f_i = \text{zer} \partial(\sum_{i \in I} f_i) = \text{Argmin} \sum_{i \in I} f_i$ .  $\square$

## 27.3 Forward–Backward Algorithm

To minimize the sum of two functions in  $\Gamma_0(\mathcal{H})$  when one of them is smooth, we can use the following version of the forward–backward algorithm (25.26). This method is sometimes called the *proximal-gradient algorithm*.

**Corollary 27.9 (forward–backward algorithm)** *Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g: \mathcal{H} \rightarrow \mathbb{R}$  be convex and differentiable with a  $1/\beta$ -Lipschitz continuous gradient for some  $\beta \in \mathbb{R}_{++}$ , let  $\gamma \in ]0, 2\beta[$ , and set  $\delta = \min\{1, \beta/\gamma\} + 1/2$ . Furthermore, let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, \delta]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(\delta - \lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Suppose that  $\text{Argmin}(f + g) \neq \emptyset$  and set*

$$(\forall n \in \mathbb{N}) \quad \left\{ \begin{array}{l} y_n = x_n - \gamma \nabla g(x_n), \\ x_{n+1} = x_n + \lambda_n(\text{Prox}_{\gamma f} y_n - x_n). \end{array} \right. \quad (27.26)$$

Then the following hold:

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Argmin}(f + g)$ .
- (ii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and let  $x \in \text{Argmin}(f + g)$ . Then  $(\nabla g(x_n))_{n \in \mathbb{N}}$  converges strongly to  $\nabla g(x)$ .
- (iii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and that one of the following holds:
  - (a)  $f$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ .
  - (b)  $g$  is uniformly convex on every nonempty bounded subset of  $\mathcal{H}$ .

Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique minimizer of  $f + g$ .

*Proof.* Set  $A = \partial f$  and  $B = \nabla g$ . Then  $A$  and  $B$  are maximally monotone by Theorem 20.40 and, since  $\text{dom } g = \mathcal{H}$ , Corollary 26.3 yields  $\text{Argmin}(f + g) =$

$\text{zer}(A + B)$ . On the other hand, by Corollary 18.16,  $B$  is  $\beta$ -cocoercive. In view of Example 23.3 and Example 22.4, the claims therefore follow from Theorem 25.8.  $\square$

A special case of the proximal-gradient algorithm (27.26) is the *projection-gradient algorithm* described next.

**Corollary 27.10 (projection-gradient algorithm)** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be convex and differentiable with a  $1/\beta$ -Lipschitz continuous gradient for some  $\beta \in \mathbb{R}_{++}$ , let  $\gamma \in ]0, 2\beta[$ , and set  $\delta = \min\{1, \beta/\gamma\} + 1/2$ . Furthermore, let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, \delta]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(\delta - \lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Suppose that  $\text{Argmin}_C f \neq \emptyset$  and set*

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n(P_C(x_n - \gamma \nabla f(x_n)) - x_n). \quad (27.27)$$

Then the following hold:

- (i)  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point in  $\text{Argmin}_C f$ .
- (ii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and let  $x \in \text{Argmin}_C f$ . Then  $(\nabla f(x_n))_{n \in \mathbb{N}}$  converges strongly to  $\nabla f(x)$ .
- (iii) Suppose that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and that  $f$  is uniformly convex on every nonempty bounded subset of  $\mathcal{H}$ . Then  $(x_n)_{n \in \mathbb{N}}$  converges strongly to the unique minimizer of  $f$  over  $C$ .

*Proof.* Apply Corollary 27.9 to  $\iota_C$  and  $f$ , and use Example 12.25 and Example 22.4.  $\square$

The next result concerns the alternating projection method.

**Example 27.11** Let  $C$  and  $D$  be nonempty closed convex subsets of  $\mathcal{H}$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $[0, 3/2]$  such that  $\sum_{n \in \mathbb{N}} \lambda_n(3 - 2\lambda_n) = +\infty$ , and let  $x_0 \in \mathcal{H}$ . Suppose that  $C$  or  $D$  is bounded and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + \lambda_n(P_C P_D x_n - x_n). \quad (27.28)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point  $x \in C$  that is at minimal distance from  $D$ .

*Proof.* Set  $f = (1/2)d_D^2$ . Then  $f$  is coercive if  $D$  is bounded, and we therefore deduce from Proposition 11.14 that  $\text{Argmin}_C f \neq \emptyset$ . Next, it follows from Corollary 12.30 and Corollary 4.10 that  $\nabla f = \text{Id} - P_D$  is nonexpansive. Hence, the result is an application of Corollary 27.10(i) with  $\beta = 1$  and  $\gamma = 1$ .  $\square$

We now consider an example of linear convergence.

**Example 27.12** Let  $f \in \Gamma_0(\mathcal{H})$  be  $\alpha$ -strongly convex for some  $\alpha \in \mathbb{R}_{++}$ , let  $g: \mathcal{H} \rightarrow \mathbb{R}$  be convex and differentiable with a  $1/\beta$ -Lipschitz continuous gradient for some  $\beta \in \mathbb{R}_{++}$ , and let  $\gamma \in ]0, 2\beta[$ . Let  $x_0 \in \mathcal{H}$  and set



$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma \nabla g(x_n), \\ x_{n+1} = \text{Prox}_{\gamma f} y_n. \end{cases} \quad (27.29)$$

Then  $(x_n)_{n \in \mathbb{N}}$  converges linearly to the unique point in  $\text{Argmin}(f + g)$ .

*Proof.* Set  $A = \partial f$  and  $B = \nabla g$ . Then Corollary 26.3 yields  $\text{Argmin}(f + g) = \text{zer}(A + B)$ . We also note that  $A$  is maximally monotone by Theorem 20.40 and  $\alpha$ -strongly monotone by Example 22.3(iv). Furthermore, we derive from Corollary 18.16 that  $B$  is  $\beta$ -cocoercive. In view of Example 23.3, the claim therefore follows from Proposition 25.9(i).  $\square$

## 27.4 Tseng's Splitting Algorithm

We study the convergence of a variant of (27.26) under more relaxed hypotheses on the functions  $f$  and  $g$  than those imposed in Corollary 27.9.

**Proposition 27.13 (Tseng's algorithm)** *Let  $D$  be a nonempty subset of  $\mathcal{H}$ , let  $f \in \Gamma_0(\mathcal{H})$  be such that  $\text{dom } \partial f \subset D$ , and let  $g \in \Gamma_0(\mathcal{H})$  be Gâteaux differentiable on  $D$ . Suppose that  $C$  is a closed convex subset of  $D$  such that  $C \cap \text{Argmin}(f + g) \neq \emptyset$ , and that  $\nabla g$  is  $1/\beta$ -Lipschitz continuous relative to  $C \cup \text{dom } \partial f$ , for some  $\beta \in \mathbb{R}_{++}$ . Let  $x_0 \in C$ , let  $\gamma \in ]0, \beta[$ , and set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma \nabla g(x_n), \\ z_n = \text{Prox}_{\gamma f} y_n, \\ r_n = z_n - \gamma \nabla g(z_n), \\ x_{n+1} = P_C(x_n - y_n + r_n). \end{cases} \quad (27.30)$$

Then the following hold:

- (i)  $(x_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (ii)  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to a point in  $C \cap \text{Argmin}(f + g)$ .
- (iii) Suppose that  $f$  or  $g$  is uniformly convex on every nonempty bounded subset of  $\text{dom } \partial f$ . Then  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique point in  $C \cap \text{Argmin}(f + g)$ .

*Proof.* Set  $A = \partial f$  and  $B = \partial g$ . Then  $A$  and  $B$  are maximally monotone by Theorem 20.40. Moreover, it follows from Proposition 17.26(i) that  $B$  is single-valued on  $D$ , where it coincides with  $\nabla g$ . Furthermore, Proposition 17.41 and Proposition 16.21 imply that  $\text{dom } \partial f \subset D \subset \text{int dom } g = \text{int dom } \partial g$  and hence, by Corollary 24.4(ii),  $A + B = \partial f + \partial g$  is maximally monotone. In turn, it follows from Corollary 16.38(ii) that  $\partial(f + g) = \partial f + \partial g$  and hence from Proposition 26.1 that  $\text{Argmin}(f + g) = \text{zer}(A + B)$ . Altogether, the results follow from Theorem 25.10, Example 23.3, and Example 22.4.  $\square$

As an example, we obtain the following variant of the projection-gradient algorithm.

**Example 27.14** Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $\beta \in \mathbb{R}_{++}$ , and  $g \in \Gamma_0(\mathcal{H})$  be differentiable with a  $1/\beta$ -Lipschitz continuous gradient relative to  $C$ . Suppose that  $\text{Argmin}_C g \neq \emptyset$ , let  $x_0 \in C$ , let  $\gamma \in ]0, \beta[$ , and set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma \nabla g(x_n), \\ z_n = P_C y_n, \\ r_n = z_n - \gamma \nabla g(z_n), \\ x_{n+1} = P_C(x_n - y_n + r_n). \end{cases} \quad (27.31)$$

Then the following hold:

- (i)  $(x_n - z_n)_{n \in \mathbb{N}}$  converges strongly to 0.
- (ii)  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge weakly to a point in  $\text{Argmin}_C g$ .
- (iii) Suppose that  $g$  is uniformly convex on every nonempty bounded subset of  $C$ . Then  $(x_n)_{n \in \mathbb{N}}$  and  $(z_n)_{n \in \mathbb{N}}$  converge strongly to the unique point in  $\text{Argmin}_C g$ .

*Proof.* This is a special case of Proposition 27.13 applied to  $D = C$  and  $f = \iota_C$ .  $\square$

## 27.5 A Primal–Dual Algorithm

In this section we revisit the duality framework investigated in Section 19.1 and provide an algorithm to solve both the primal and the dual problems in the setting of Proposition 19.4.

**Proposition 27.15 (primal–dual algorithm)** *Let  $\varphi \in \Gamma_0(\mathcal{H})$ , let  $\psi \in \Gamma_0(\mathcal{K})$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , and suppose that  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  satisfies  $L \neq 0$  and  $r \in \text{sri}(L(\text{dom } \varphi) - \text{dom } \psi)$ . Consider the (primal) problem*

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \varphi(x) + \psi(Lx - r) + \frac{1}{2} \|x - z\|^2, \quad (27.32)$$

*together with the (dual) problem*

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad {}^1(\varphi^*)(L^*v + z) + \psi^*(-v) - \langle v \mid r \rangle. \quad (27.33)$$

*Let  $\gamma \in ]0, 2\|L\|^{-2}[$ , set  $\delta = \min\{1, \gamma^{-1}\|L\|^{-2}\} + 1/2$ , let  $(\lambda_n)_{n \in \mathbb{N}}$  be a sequence in  $]0, \delta[$  such that  $\inf_{n \in \mathbb{N}} \lambda_n > 0$  and  $\sup_{n \in \mathbb{N}} \lambda_n < \delta$ , and let  $v_0 \in \mathcal{K}$ . Set*

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = \text{Prox}_{\varphi}(L^*v_n + z), \\ v_{n+1} = v_n - \lambda_n (\text{Prox}_{\gamma\psi^*}(\gamma(Lx_n - r) - v_n) + v_n), \end{cases} \quad (27.34)$$

*and let  $\bar{x}$  be the unique solution to (27.32). Then the following hold:*

- (i)  $(v_n)_{n \in \mathbb{N}}$  converges weakly to a solution  $\bar{v}$  to the dual problem (27.33) and  $\bar{x} = \text{Prox}_\varphi(L^*\bar{v} + z)$ .  
(ii)  $(x_n)_{n \in \mathbb{N}}$  converges strongly to  $\bar{x}$ .

*Proof.* Set  $\mathfrak{h}: \mathcal{H} \rightarrow ]-\infty, +\infty]: x \mapsto \varphi(x) + (1/2)\|x - z\|^2$  and  $\mathfrak{j}: \mathcal{K} \rightarrow ]-\infty, +\infty]: y \mapsto \psi(y - r)$ . Then (27.32) can be rewritten as

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \mathfrak{h}(x) + \mathfrak{j}(Lx). \quad (27.35)$$

Moreover, the assumptions imply that  $\mathfrak{h} \in \Gamma_0(\mathcal{H})$ ,  $\mathfrak{j} \in \Gamma_0(\mathcal{K})$ , and  $0 \in \text{sri}(L(\text{dom } \mathfrak{h}) - \text{dom } \mathfrak{j})$ . Hence, it follows from Theorem 15.23 that the dual of (27.35), namely

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad \mathfrak{h}^*(L^*v) + \mathfrak{j}^*(-v), \quad (27.36)$$

has a least one solution. We derive from Proposition 14.1 and Proposition 13.20(iii) that  $\mathfrak{h}^*: u \mapsto {}^1(\varphi^*)(u + z) - (1/2)\|z\|^2$  and  $\mathfrak{j}^*: v \mapsto \psi(v) + \langle v \mid r \rangle$ . Therefore, (27.36) coincides with (27.33) up to an additive constant. We have thus shown that (27.33) is the dual of (27.32) (up to an additive constant), and that it has at least one solution. Now let us define two functions  $f$  and  $g$  on  $\mathcal{K}$  by  $f: v \mapsto \psi^*(-v) - \langle v \mid r \rangle$  and  $g: v \mapsto {}^1(\varphi^*)(L^*v + z)$ . Then (27.33) amounts to minimizing  $f + g$  on  $\mathcal{K}$ . By Corollary 13.33,  $f$  and  $g$  are in  $\Gamma_0(\mathcal{K})$  and, as just observed,  $\text{Argmin}(f + g) \neq \emptyset$ . Moreover, it follows from Proposition 12.29 and (14.6) that  $g$  is differentiable on  $\mathcal{K}$  with gradient

$$\nabla g: v \mapsto L(\text{Prox}_\varphi(L^*v + z)). \quad (27.37)$$

Hence, Proposition 12.27 yields

$$\begin{aligned} (\forall v \in \mathcal{K})(\forall w \in \mathcal{K}) \quad & \|\nabla g(v) - \nabla g(w)\| \\ & \leq \|L\| \|\text{Prox}_\varphi(L^*v + z) - \text{Prox}_\varphi(L^*w + z)\| \\ & \leq \|L\| \|L^*v - L^*w\| \\ & \leq \|L\|^2 \|v - w\|. \end{aligned} \quad (27.38)$$

The reciprocal of the Lipschitz constant of  $\nabla g$  is therefore  $\|L\|^{-2}$ . On the other hand, we derive from (27.34), (27.37), and Proposition 23.29(v)&(i) that, for every  $n \in \mathbb{N}$ ,

$$\begin{aligned} v_{n+1} &= v_n - \lambda_n(\text{Prox}_{\gamma\psi^*}(\gamma(Lx_n - r) - v_n) + v_n) \\ &= v_n - \lambda_n(\text{Prox}_{\gamma\psi^*}(\gamma(\nabla g(v_n) - r) - v_n) + v_n) \\ &= v_n + \lambda_n(\text{Prox}_{\gamma\psi^*}^\vee(v_n - \gamma\nabla g(v_n) + \gamma r) - v_n) \\ &= v_n + \lambda_n(\text{Prox}_{\gamma\psi^*}^\vee - \langle \cdot \mid \gamma r \rangle)(v_n - \gamma\nabla g(v_n)) - v_n \\ &= v_n + \lambda_n(\text{Prox}_{\gamma f}(v_n - \gamma\nabla g(v_n)) - v_n). \end{aligned} \quad (27.39)$$

We thus recover the proximal-gradient iteration (27.26).

(i): In view of the above, this follows from Corollary 27.9 and Proposition 19.4.

(ii): As seen in (i),  $v_n \rightharpoonup \bar{v}$ , where  $\bar{v}$  is a solution to (27.33), and  $\bar{x} = \text{Prox}_\varphi(L^*\bar{v} + z)$ . Now set  $\rho = \sup_{n \in \mathbb{N}} \|v_n - \bar{v}\|$ . Then Lemma 2.38 yields  $\rho < +\infty$ . Hence, using Proposition 12.27 and (27.37) we obtain, for every  $n \in \mathbb{N}$ ,

$$\begin{aligned} \|x_n - \bar{x}\|^2 &= \|\text{Prox}_\varphi(L^*v_n + z) - \text{Prox}_\varphi(L^*\bar{v} + z)\|^2 \\ &\leq \langle L^*v_n - L^*\bar{v} \mid \text{Prox}_\varphi(L^*v_n + z) - \text{Prox}_\varphi(L^*\bar{v} + z) \rangle \\ &= \langle v_n - \bar{v} \mid L(\text{Prox}_\varphi(L^*v_n + z)) - L(\text{Prox}_\varphi(L^*\bar{v} + z)) \rangle \\ &= \langle v_n - \bar{v} \mid \nabla g(v_n) - \nabla g(\bar{v}) \rangle \\ &\leq \rho \|\nabla g(v_n) - \nabla g(\bar{v})\|. \end{aligned} \quad (27.40)$$

However, Corollary 27.9(ii) yields  $\|\nabla g(v_n) - \nabla g(\bar{v})\| \rightarrow 0$ . Hence, we derive from (27.40) that  $x_n \rightarrow \bar{x}$ .  $\square$

As an illustration, we revisit Example 19.7 and Example 19.9.

**Example 27.16** Let  $K$  be a closed convex cone in  $\mathcal{H}$ , let  $\psi \in \Gamma_0(K)$  be positively homogeneous, set  $D = \partial\psi(0)$ , let  $z \in \mathcal{H}$ , let  $r \in K$ , and suppose that  $0 \neq L \in \mathcal{B}(\mathcal{H}, K)$  satisfies  $r \in \text{sri}(L(K) - \text{dom } \psi)$ . Consider the (primal) problem

$$\underset{x \in K}{\text{minimize}} \quad \psi(Lx - r) + \frac{1}{2}\|x - z\|^2, \quad (27.41)$$

together with the (dual) problem

$$\underset{-v \in D}{\text{minimize}} \quad \frac{1}{2}d_{K^\ominus}^2(L^*v + z) - \langle v \mid r \rangle. \quad (27.42)$$

Let  $\gamma \in ]0, 2\|L\|^{-2}]$ , set  $\delta = \min\{1, \gamma^{-1}\|L\|^{-2}\} + 1/2$ , let  $\lambda \in ]0, \delta[$ , and let  $v_0 \in K$ . Set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = P_K(L^*v_n + z), \\ v_{n+1} = v_n - \lambda(P_D(\gamma(Lx_n - r) - v_n) + v_n), \end{cases} \quad (27.43)$$

and let  $\bar{x}$  be the unique solution to (27.41). Then the following hold:

- (i)  $(v_n)_{n \in \mathbb{N}}$  converges weakly to a solution  $\bar{v}$  to (27.42) and  $\bar{x} = P_K(L^*\bar{v} + z)$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges strongly to  $\bar{x}$ .

*Proof.* In view of Example 19.7, this is an application of Proposition 27.15 with  $\varphi = \iota_K$ , and  $\lambda_n \equiv \lambda$ .  $\square$

**Example 27.17** Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$ , let  $z \in \mathcal{H}$ , and suppose that  $0 \neq L \in \mathcal{B}(\mathcal{H}, K)$  satisfies  $0 \in \text{sri}(D - L(C))$ . Consider the best approximation (primal) problem

$$\underset{\substack{x \in C \\ Lx \in D}}{\text{minimize}} \quad \|x - z\|, \quad (27.44)$$

together with the (dual) problem

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad \frac{1}{2} \|L^*v + z\|^2 - \frac{1}{2} d_C^2(L^*v + z) + \sigma_D(-v). \quad (27.45)$$

Let  $\gamma \in ]0, 2\|L\|^{-2}[$ , set  $\delta = \min\{1, \gamma^{-1}\|L\|^{-2}\} + 1/2$ , let  $\lambda \in ]0, \delta[$ , and let  $v_0 \in \mathcal{K}$ . Set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = P_C(L^*v_n + z), \\ v_{n+1} = v_n + \gamma\lambda(P_D(Lx_n - \gamma^{-1}v_n) - Lx_n), \end{cases} \quad (27.46)$$

and let  $\bar{x}$  be the unique solution to (27.44). Then the following hold:

- (i)  $(v_n)_{n \in \mathbb{N}}$  converges weakly to a solution  $\bar{v}$  to (27.45) and  $\bar{x} = P_C(L^*\bar{v} + z)$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges strongly to  $\bar{x}$ .

*Proof.* In view of Example 19.9, this is an application of Proposition 27.15 with  $\varphi = \iota_C$ ,  $\psi = \iota_D$ ,  $r = 0$ , and  $\lambda_n \equiv \lambda$ , where we have used Theorem 14.3(ii) to derive (27.46) from (27.34).  $\square$

**Example 27.18** Let  $C$  be a closed convex subset of  $\mathcal{H}$ , let  $z \in \mathcal{H}$ , let  $r \in \mathcal{K}$ , and suppose that  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  satisfies  $\|L\| = 1$  and that  $r \in \text{sri } L(C)$ . Consider the best approximation (primal) problem

$$\underset{\substack{x \in C \\ Lx = r}}{\text{minimize}} \quad \|x - z\|, \quad (27.47)$$

together with the (dual) problem

$$\underset{v \in \mathcal{K}}{\text{minimize}} \quad \frac{1}{2} \|L^*v + z\|^2 - \frac{1}{2} d_C^2(L^*v + z) - \langle v | r \rangle. \quad (27.48)$$

Let  $v_0 \in \mathcal{K}$ , set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} x_n = P_C(L^*v_n + z), \\ v_{n+1} = v_n + r - Lx_n, \end{cases} \quad (27.49)$$

and let  $\bar{x}$  be the unique solution to (27.47). Then the following hold:

- (i)  $(v_n)_{n \in \mathbb{N}}$  converges weakly to a solution  $\bar{v}$  to (27.48) and  $\bar{x} = P_C(L^*\bar{v} + z)$ .
- (ii)  $(x_n)_{n \in \mathbb{N}}$  converges strongly to  $\bar{x}$ .

*Proof.* Apply Example 27.17 with  $D = \{r\}$ ,  $\gamma = 1$ , and  $\lambda = 1$ .  $\square$

## Exercises

**Exercise 27.1** Let  $C$  and  $D$  be closed convex subsets of  $\mathcal{H}$  such that  $C \cap D \neq \emptyset$ , and let  $x_0 \in \mathcal{H}$ . Set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = x_n + P_C(2P_Dx_n - x_n) - P_Cx_n. \quad (27.50)$$

Show that  $(x_n)_{n \in \mathbb{N}}$  converges weakly to a point  $x \in \mathcal{H}$  and that  $P_D x_n \rightarrow P_D x \in C \cap D$ .

**Exercise 27.2** Derive Corollary 27.7 from Corollary 27.4.

**Exercise 27.3** Let  $\varphi \in \Gamma_0(\mathcal{H})$ , let  $\psi \in \Gamma_0(\mathcal{H})$ , let  $z \in \mathcal{H}$ , and suppose that  $0 \in \text{sri}(\text{dom } \psi - \text{dom } \varphi)$ . Consider the problem of constructing  $\text{Prox}_{\varphi+\psi} z$ , i.e., of finding the solution to the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \varphi(x) + \psi(x) + \frac{1}{2} \|x - z\|^2. \quad (27.51)$$

Use the following results to solve (27.51) (in each case provide an explicit algorithm and a strong convergence result):

- (i) Corollary 27.4 and Theorem 25.6(vii).
- (ii) Proposition 27.15.

**Exercise 27.4** In Corollary 27.10, suppose that  $C$  is a compact set and that  $\sup_{n \in \mathbb{N}} \lambda_n \leq 1$ . Show that the sequence  $(x_n)_{n \in \mathbb{N}}$  produced by the projection-gradient algorithm (27.27) converges strongly to a minimizer of  $f$  over  $C$ .

**Exercise 27.5** Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , and let  $C$  and  $(C_i)_{i \in I}$  be nonempty closed convex subsets of  $\mathcal{H}$ , at least one of which is bounded. Consider the problem

$$\underset{x \in C}{\text{minimize}} \quad \frac{1}{2m} \sum_{i \in I} d_{C_i}^2(x). \quad (27.52)$$

- (i) Show that (27.52) has at least one solution.
- (ii) Use the projection-gradient algorithm to solve (27.52). Carefully check that all the assumptions in Corollary 27.10 are satisfied; provide an algorithm and a weak convergence result.

**Exercise 27.6** Suppose that  $(e_k)_{k \in \mathbb{N}}$  is an orthonormal basis of  $\mathcal{H}$  and let  $(\phi_k)_{k \in \mathbb{N}}$  be a family of functions in  $\Gamma_0(\mathbb{R})$  such that  $(\forall k \in \mathbb{N}) \phi_k \geq \phi_k(0) = 0$ . Moreover, let  $L$  be a nonzero operator in  $\mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $r \in \mathcal{K}$ , and assume that the problem

$$\underset{x \in \mathcal{H}}{\text{minimize}} \quad \sum_{k \in \mathbb{N}} \phi_k(\langle x \mid e_k \rangle) + \frac{1}{2} \|Lx - r\|^2 \quad (27.53)$$

admits at least one solution.

- (i) Use the Douglas–Rachford algorithm to solve (27.53). Carefully check that all the assumptions in Corollary 27.4 are satisfied; provide an algorithm and a weak convergence result.
- (ii) Use the forward–backward algorithm to solve (27.53). Carefully check that all the assumptions in Corollary 27.9 are satisfied; provide an algorithm and a weak convergence result.

**Exercise 27.7** Suppose that  $(e_k)_{k \in \mathbb{N}}$  is an orthonormal basis of  $\mathcal{H}$  and let  $(\phi_k)_{k \in \mathbb{N}}$  be a family of functions in  $\Gamma_0(\mathbb{R})$  such that  $(\forall k \in \mathbb{N}) \phi_k \geq \phi_k(0) = 0$ . Moreover, let  $L$  be a nonzero operator in  $\mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $\psi \in \Gamma_0(\mathcal{K})$ , suppose that  $r \in \text{sri } L(\text{dom } \varphi)$ , and consider the problem

$$\underset{Lx=r}{\text{minimize}} \quad \sum_{k \in \mathbb{N}} \left( \phi_k(\langle x | e_k \rangle) + |\langle x | e_k \rangle|^2 \right). \quad (27.54)$$

- (i) Show that (27.54) admits exactly one solution.
- (ii) Use the primal–dual algorithm (27.34) to solve (27.54). Carefully check that all the assumptions in Proposition 27.15 are satisfied; provide an algorithm and a strong convergence result.





# Chapter 28

## Projection Operators

Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ . The projection  $P_C x$  of a point  $x \in \mathcal{H}$  onto  $C$  is characterized by (see Theorem 3.14)

$$P_C x \in C \quad \text{and} \quad (\forall y \in C) \quad \langle y - P_C x \mid x - P_C x \rangle \leq 0 \quad (28.1)$$

or, equivalently, by (see Proposition 6.46)  $x - P_C x \in N_C(P_C x)$ . In this chapter, we investigate further the properties of projectors and provide a variety of examples.

### 28.1 Basic Properties

As a special case of proximity operator, projectors inherit their properties. Here are some examples.

**Proposition 28.1** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$  and let  $x \in \mathcal{H}$ . Then the following hold:*

- (i) *Set  $D = z + C$ , where  $z \in \mathcal{H}$ . Then  $P_D x = z + P_C(x - z)$ .*
- (ii) *Set  $D = \rho C$ , where  $\rho \in \mathbb{R} \setminus \{0\}$ . Then  $P_D x = \rho P_C(\rho^{-1}x)$ .*
- (iii) *Set  $D = -C$ . Then  $P_D x = -P_C(-x)$ .*

*Proof.* These properties are obtained by setting  $f = \iota_C$  in the following results:

- (i): Proposition 23.29(ii) (see also Proposition 3.17).
- (ii): Proposition 23.29(iv).
- (iii): Proposition 23.29(v) or set  $\rho = -1$  in (ii). □

**Proposition 28.2** *Let  $\mathcal{K}$  be a real Hilbert space, let  $C$  be a nonempty closed convex subset of  $\mathcal{K}$ , let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , let  $x \in \mathcal{H}$ , and set  $D = L^{-1}(C)$ . Then the following hold:*

- (i) Suppose that  $LL^* = \gamma \text{Id}$  for some  $\gamma \in \mathbb{R}_{++}$ . Then  $P_D x = x + \gamma^{-1} L^*(P_C(Lx) - Lx)$ .
- (ii) Suppose that  $L$  is invertible, with  $L^{-1} = L^*$ . Then  $P_D x = L^{-1} P_C(Lx)$ .

*Proof.* (i): Set  $f = \iota_C$  in Proposition 23.32.

(ii): This follows from (i).  $\square$

**Proposition 28.3** Let  $(\mathcal{H}_i)_{i \in I}$  be a totally ordered finite family of real Hilbert spaces, set  $\mathcal{H} = \bigoplus_{i \in I} \mathcal{H}_i$ , and, for every  $i \in I$ , let  $C_i$  be a nonempty closed convex subset of  $\mathcal{H}_i$  and  $x_i \in \mathcal{H}_i$ . Set  $\mathbf{x} = (x_i)_{i \in I}$  and  $\mathbf{C} = \times_{i \in I} C_i$ . Then  $P_{\mathbf{C}} \mathbf{x} = (P_{C_i} x_i)_{i \in I}$ .

*Proof.* It is clear that  $\mathbf{C}$  is a nonempty closed convex subset of  $\mathcal{H}$ . The result therefore follows from Proposition 23.30, where  $(\forall i \in I) f_i = \iota_{C_i}$ .  $\square$

**Proposition 28.4** Let  $(C_i)_{i \in I}$  be a totally ordered family of closed intervals of  $\mathbb{R}$  containing 0, suppose that  $\mathcal{H} = \ell^2(I)$ , set  $\mathbf{C} = \times_{i \in I} C_i$ , and let  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ . Then  $P_{\mathbf{C}}$  is weakly sequentially continuous and  $P_{\mathbf{C}} x = (P_{C_i} \xi_i)_{i \in I}$ .

*Proof.* Apply Proposition 23.31 with  $(\forall i \in I) \phi_i = \iota_{C_i}$ .  $\square$

**Proposition 28.5** Let  $C$  and  $D$  be nonempty closed convex subsets of  $\mathcal{H}$ , let  $x \in \mathcal{H}$ , and suppose that  $P_C x \in D$ . Then  $P_{C \cap D} x = P_C x$ .

*Proof.* On the one hand,  $P_C x \in C \cap D$ . On the other hand, for every  $y \in C \cap D \subset C$ , (28.1) yields  $\langle y - P_C x \mid x - P_C x \rangle \leq 0$ . Altogether, Theorem 3.14 yields  $P_{C \cap D} x = P_C x$ .  $\square$

**Proposition 28.6** Let  $C$  and  $D$  be nonempty closed convex subsets of  $\mathcal{H}$  such that  $C \perp D$ . Then  $C + D$  is closed and convex, and  $P_{C+D} = P_D + P_C \circ (\text{Id} - P_D)$ .

*Proof.* It is clear that  $\text{ran}(P_D + P_C \circ (\text{Id} - P_D)) \subset C + D$ . Now let  $x \in \mathcal{H}$ , let  $y \in C$ , and let  $z \in D$ . Then, by (28.1),

$$\begin{aligned}
 & \langle (z + y) - (P_D x + P_C(x - P_D x)) \mid x - (P_D x + P_C(x - P_D x)) \rangle \\
 &= \langle z - P_D x \mid x - P_D x \rangle - \langle z \mid P_C(x - P_D x) \rangle + \langle P_D x \mid P_C(x - P_D x) \rangle \\
 & \quad + \langle y - P_C(x - P_D x) \mid (x - P_D x) - P_C(x - P_D x) \rangle \\
 & \leq 0,
 \end{aligned} \tag{28.2}$$

which yields the result.  $\square$

Next, we discuss some simple asymptotic properties.

**Proposition 28.7** Let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of nonempty closed convex subsets of  $\mathcal{H}$  such that  $(\forall n \in \mathbb{N}) C_n \subset C_{n+1}$ . Set  $C = \bigcup_{n \in \mathbb{N}} C_n$  and let  $x \in \mathcal{H}$ . Then  $P_{C_n} x \rightarrow P_C x$ .

*Proof.* It follows from Proposition 3.36(i) that  $C$  is a nonempty closed convex set. By assumption, there exists a sequence  $(y_n)_{n \in \mathbb{N}}$  in  $\mathcal{H}$  such that  $y_n \rightarrow P_C x$  and  $(\forall n \in \mathbb{N}) y_n \in C_n$ . In turn,  $(\forall n \in \mathbb{N}) \|x - P_C x\| \leq \|x - P_{C_n} x\| \leq \|x - y_n\|$ . Hence,

$$\|x - P_{C_n} x\| \rightarrow \|x - P_C x\|. \quad (28.3)$$

Therefore, every weak sequential cluster point  $z$  of  $(P_{C_n} x)_{n \in \mathbb{N}}$  lies in  $C$  by Theorem 3.32 and since, by Lemma 2.35 and (28.3),  $\|x - P_C x\| \leq \|x - z\| \leq \lim \|x - P_{C_n} x\|$ , it satisfies  $\|x - z\| = \|x - P_C x\|$ . It follows that  $P_C x$  is the only weak sequential cluster point of  $(P_{C_n} x)_{n \in \mathbb{N}}$  and, in turn, that  $x - P_{C_n} x \rightarrow x - P_C x$  by Lemma 2.38. In view of (28.3) and Corollary 2.42, we conclude that  $x - P_{C_n} x \rightarrow x - P_C x$ .  $\square$

**Proposition 28.8** *Let  $(C_n)_{n \in \mathbb{N}}$  be a sequence of nonempty closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{n \in \mathbb{N}} C_n \neq \emptyset$  and  $(\forall n \in \mathbb{N}) C_{n+1} \subset C_n$ , and let  $x \in \mathcal{H}$ . Then  $P_{C_n} x \rightarrow P_C x$ .*

*Proof.* Set  $(\forall n \in \mathbb{N}) p_n = P_{C_n} x$ . Then  $(\forall n \in \mathbb{N}) \|x - p_n\| \leq \|x - p_{n+1}\| \leq \|x - P_C x\|$ . Hence,  $(p_n)_{n \in \mathbb{N}}$  is bounded and  $(\|x - p_n\|)_{n \in \mathbb{N}}$  is convergent. For every  $m$  and  $n$  in  $\mathbb{N}$  such that  $m \leq n$ , since  $(p_m + p_n)/2 \in C_m$ , Lemma 2.11(iv) yields

$$\begin{aligned} \|p_n - p_m\|^2 &= 2(\|p_n - x\|^2 + \|p_m - x\|^2) - 4\|(p_n + p_m)/2 - x\|^2 \\ &\leq 2(\|p_n - x\|^2 - \|p_m - x\|^2) \\ &\rightarrow 0 \quad \text{as } m, n \rightarrow +\infty. \end{aligned} \quad (28.4)$$

Hence,  $(p_n)_{n \in \mathbb{N}}$  is a Cauchy sequence and, in turn,  $p_n \rightarrow p$  for some  $p \in \mathcal{H}$ . For every  $n \in \mathbb{N}$ ,  $(p_k)_{k \geq n}$  lies in  $C_n$  and hence  $p \in C_n$ , since  $C_n$  is closed. Thus,  $p \in C$  and therefore  $\|x - P_C x\| \leq \|x - p\| = \lim \|x - p_n\| \leq \|x - P_C x\|$ . Since  $C$  is a Chebyshev set, we deduce that  $p = P_C x$ .  $\square$

**Remark 28.9** As seen in Proposition 3.18, the condition  $\bigcap_{n \in \mathbb{N}} C_n \neq \emptyset$  in Proposition 28.8 holds if  $C_0$  is bounded.

The proof of the following result is left as Exercise 28.5.

**Proposition 28.10** *Let  $C$  be a nonempty closed convex subset of  $\mathcal{H}$ , let  $\varepsilon \in \mathbb{R}_{++}$ , let  $x \in \mathcal{H}$ , and set  $D = C + B(0; \varepsilon)$ . Then  $D$  is a nonempty closed convex subset of  $\mathcal{H}$  and*

$$P_D x = \begin{cases} x, & \text{if } \|x - P_C x\| \leq \varepsilon; \\ P_C x + \varepsilon \frac{x - P_C x}{\|x - P_C x\|}, & \text{otherwise.} \end{cases} \quad (28.5)$$

## 28.2 Projections onto Affine Subspaces

The following result complements Corollary 3.20.

**Proposition 28.11** *Let  $C$  be a closed affine subspace of  $\mathcal{H}$ . Then the following hold:*

- (i)  $P_C$  is a weakly continuous affine operator.
- (ii)  $(\forall x \in \mathcal{H})(\forall y \in C)(\forall z \in C) \quad \langle x - P_C x \mid y - z \rangle = 0$ .
- (iii)  $(\forall x \in \mathcal{H})(\forall y \in C) \quad \|x - P_C x\|^2 = \langle x - y \mid x - P_C x \rangle$ .

*Proof.* (i): This is a consequence of Proposition 4.8, Corollary 3.20(ii), and Lemma 2.34.

(ii): This follows from Corollary 3.20(i).

(iii): Let  $x \in \mathcal{H}$  and  $y \in C$ . Then (ii) implies that  $\|x - P_C x\|^2 = \langle x - P_C x \mid (x - y) + (y - P_C x) \rangle = \langle x - P_C x \mid x - y \rangle$ .  $\square$

**Proposition 28.12** *Let  $\{e_i\}_{i \in I}$  be a countable orthonormal subset of  $\mathcal{H}$  and set  $C = \overline{\text{span}} \{e_i\}_{i \in I}$ . Then*

$$(\forall x \in \mathcal{H}) \quad P_C x = \sum_{i \in I} \langle x \mid e_i \rangle e_i. \quad (28.6)$$

*Proof.* If  $I$  is finite, then  $C = \text{span}\{e_i\}_{i \in I}$  and hence Example 3.8 yields the result. If  $I$  is countably infinite, we combine the finite case with Proposition 28.7 to obtain the conclusion.  $\square$

**Proposition 28.13** *Let  $I$  be a totally ordered finite set, let  $(\omega_i)_{i \in I}$  be real numbers in  $]0, 1]$  such that  $\sum_{i \in I} \omega_i = 1$ , and let  $\mathcal{H}$  be the real Hilbert space obtained by endowing the Cartesian product  $\times_{i \in I} \mathcal{H}$  with the usual vector space structure and with the scalar product  $(\mathbf{x}, \mathbf{y}) \mapsto \sum_{i \in I} \omega_i \langle x_i \mid y_i \rangle$ , where  $\mathbf{x} = (x_i)_{i \in I}$  and  $\mathbf{y} = (y_i)_{i \in I}$  lie in  $\mathcal{H}$ . Set  $\mathbf{D} = \{(x)_{i \in I} \mid x \in \mathcal{H}\}$ , let  $\mathbf{x} \in \mathcal{H}$ , and set  $\mathbf{p} = \sum_{i \in I} \omega_i x_i$ . Then  $P_{\mathbf{D}} \mathbf{x} = (\mathbf{p})_{i \in I}$ .*

*Proof.* Set  $\mathbf{p} = (\mathbf{p})_{i \in I}$ , let  $y \in \mathcal{H}$ , and set  $\mathbf{y} = (y)_{i \in I}$ . It is clear that  $\mathbf{p} \in \mathbf{D}$ , that  $\mathbf{y} \in \mathbf{D}$ , and that  $\mathbf{D}$  is a closed linear subspace of  $\mathcal{H}$ . Furthermore,  $\langle \mathbf{x} - \mathbf{p} \mid \mathbf{y} \rangle = \sum_{i \in I} \omega_i \langle x_i - p \mid y \rangle = \langle \sum_{i \in I} \omega_i x_i - p \mid y \rangle = 0$ . Hence, we derive from Corollary 3.22(i) that  $\mathbf{p} = P_{\mathbf{D}} \mathbf{x}$ .  $\square$

Below, we provide a couple of examples of projectors onto affine subspaces.

**Example 28.14** Let  $\mathcal{K}$  be a real Hilbert space, let  $L \in \mathcal{B}(\mathcal{H}, \mathcal{K})$  be such that  $\text{ran } L$  is closed, let  $y \in \text{ran } L$ , set  $C = \{x \in \mathcal{H} \mid Lx = y\}$ , and let  $x \in \mathcal{H}$ . Then the following hold:

- (i)  $(\forall z \in C) \quad P_C x = x - L^* L^{\dagger}(x - z)$ .
- (ii)  $P_C x = x - L^* L^{\dagger}(x - L^{\dagger} y)$ .
- (iii) Suppose that  $LL^*$  is invertible. Then  $P_C x = x - L^*(LL^*)^{-1}(Lx - y)$ .

*Proof.* (i): Take  $z \in C$ . Then  $C = z + \ker L$  and it follows from Proposition 3.17 and Proposition 3.28(iii) that  $P_C x = x - L^* L^{\dagger}(x - z)$ .

(ii): Set  $z = L^{\dagger} y$  in (i).

(iii): By (ii) and Example 3.27,  $P_C x = x - L^*(LL^*)^{-1}L(x - L^{\dagger} y) = x - L^*(LL^*)^{-1}(Lx - y)$ .  $\square$

**Example 28.15** Suppose that  $u$  is a nonzero vector in  $\mathcal{H}$ , let  $\eta \in \mathbb{R}$ , and set  $C = \{x \in \mathcal{H} \mid \langle x \mid u \rangle = \eta\}$ . Then

$$(\forall x \in \mathcal{H}) \quad P_C x = x + \frac{\eta - \langle x \mid u \rangle}{\|u\|^2} u. \quad (28.7)$$

*Proof.* Set  $\phi = \iota_{\{\eta\}}$  in Corollary 23.33, or set  $L = \langle \cdot \mid u \rangle$  and  $y = \eta$  in Example 28.14(iii), or see Example 3.21.  $\square$

## 28.3 Projections onto Special Polyhedra

In the next two examples, we provide closed-form expressions for the projectors onto a half-space and onto a hyperslab, respectively.

**Example 28.16** Let  $u \in \mathcal{H}$ , let  $\eta \in \mathbb{R}$ , and set  $C = \{x \in \mathcal{H} \mid \langle x \mid u \rangle \leq \eta\}$ . Then exactly one of the following holds:

- (i)  $u = 0$  and  $\eta \geq 0$ , in which case  $C = \mathcal{H}$  and  $P_C = \text{Id}$ .
- (ii)  $u = 0$  and  $\eta < 0$ , in which case  $C = \emptyset$ .
- (iii)  $u \neq 0$ , in which case  $C \neq \emptyset$  and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x, & \text{if } \langle x \mid u \rangle \leq \eta; \\ x + \frac{\eta - \langle x \mid u \rangle}{\|u\|^2} u, & \text{if } \langle x \mid u \rangle > \eta. \end{cases} \quad (28.8)$$

*Proof.* (i)&(ii): Clear.

(iii): Set  $\phi = \iota_{]-\infty, \eta]}$  in Corollary 23.33.  $\square$

**Example 28.17** Let  $u \in \mathcal{H}$ , let  $\eta_1$  and  $\eta_2$  be in  $\mathbb{R}$ , and set

$$C = \{x \in \mathcal{H} \mid \eta_1 \leq \langle x \mid u \rangle \leq \eta_2\}. \quad (28.9)$$

Then exactly one of the following holds:

- (i)  $u = 0$  and  $\eta_1 \leq 0 \leq \eta_2$ , in which case  $C = \mathcal{H}$  and  $P_C = \text{Id}$ .
- (ii)  $u = 0$  and  $[\eta_1 > 0 \text{ or } \eta_2 < 0]$ , in which case  $C = \emptyset$ .
- (iii)  $u \neq 0$  and  $\eta_1 > \eta_2$ , in which case  $C = \emptyset$ .
- (iv)  $u \neq 0$  and  $\eta_1 \leq \eta_2$ , in which case  $C \neq \emptyset$  and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x + \frac{\eta_1 - \langle x \mid u \rangle}{\|u\|^2} u, & \text{if } \langle x \mid u \rangle < \eta_1; \\ x, & \text{if } \eta_1 \leq \langle x \mid u \rangle \leq \eta_2; \\ x + \frac{\eta_2 - \langle x \mid u \rangle}{\|u\|^2} u, & \text{if } \langle x \mid u \rangle > \eta_2. \end{cases} \quad (28.10)$$

*Proof.* (i)–(iii): Clear.

(iv): Set  $\phi = \iota_{[\eta_1, \eta_2]}$  in Corollary 23.33.  $\square$

Next, we investigate the projector onto the intersection of two half-spaces with parallel boundaries.

**Proposition 28.18** *Let  $u_1$  and  $u_2$  be in  $\mathcal{H}$ , and let  $\eta_1$  and  $\eta_2$  be in  $\mathbb{R}$ . Suppose that  $\|u_1\|^2\|u_2\|^2 = |\langle u_1 | u_2 \rangle|^2$ , i.e.,  $\{u_1, u_2\}$  is linearly dependent, and set*

$$C = \{x \in \mathcal{H} \mid \langle x | u_1 \rangle \leq \eta_1\} \cap \{x \in \mathcal{H} \mid \langle x | u_2 \rangle \leq \eta_2\}. \quad (28.11)$$

*Then exactly one of the following cases occurs:*

- (i)  $u_1 = u_2 = 0$  and  $0 \leq \min\{\eta_1, \eta_2\}$ . Then  $C = \mathcal{H}$  and  $P_C = \text{Id}$ .
- (ii)  $u_1 = u_2 = 0$  and  $\min\{\eta_1, \eta_2\} < 0$ . Then  $C = \emptyset$ .
- (iii)  $u_1 \neq 0$ ,  $u_2 = 0$ , and  $0 \leq \eta_2$ . Then  $C = \{x \in \mathcal{H} \mid \langle x | u_1 \rangle \leq \eta_1\}$  and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x, & \text{if } \langle x | u_1 \rangle \leq \eta_1; \\ x + \frac{\eta_1 - \langle x | u_1 \rangle}{\|u_1\|^2} u_1, & \text{if } \langle x | u_1 \rangle > \eta_1. \end{cases} \quad (28.12)$$

- (iv)  $u_1 \neq 0$ ,  $u_2 = 0$ , and  $\eta_2 < 0$ . Then  $C = \emptyset$ .
- (v)  $u_1 = 0$ ,  $u_2 \neq 0$ , and  $0 \leq \eta_1$ . Then  $C = \{x \in \mathcal{H} \mid \langle x | u_2 \rangle \leq \eta_2\}$  and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x, & \text{if } \langle x | u_2 \rangle \leq \eta_2; \\ x + \frac{\eta_2 - \langle x | u_2 \rangle}{\|u_2\|^2} u_2, & \text{if } \langle x | u_2 \rangle > \eta_2. \end{cases} \quad (28.13)$$

- (vi)  $u_1 = 0$ ,  $u_2 \neq 0$ , and  $\eta_1 < 0$ . Then  $C = \emptyset$ .
- (vii)  $u_1 \neq 0$ ,  $u_2 \neq 0$ , and  $\langle u_1 | u_2 \rangle > 0$ . Then  $C = \{x \in \mathcal{H} \mid \langle x | u \rangle \leq \eta\}$ , where  $u = \|u_2\| u_1$  and  $\eta = \min\{\eta_1 \|u_2\|, \eta_2 \|u_1\|\}$ , and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x, & \text{if } \langle x | u \rangle \leq \eta; \\ x + \frac{\eta - \langle x | u \rangle}{\|u\|^2} u, & \text{if } \langle x | u \rangle > \eta. \end{cases} \quad (28.14)$$

- (viii)  $u_1 \neq 0$ ,  $u_2 \neq 0$ ,  $\langle u_1 | u_2 \rangle < 0$ , and  $\eta_1 \|u_2\| + \eta_2 \|u_1\| < 0$ . Then  $C = \emptyset$ .
- (ix)  $u_1 \neq 0$ ,  $u_2 \neq 0$ ,  $\langle u_1 | u_2 \rangle < 0$ , and  $\eta_1 \|u_2\| + \eta_2 \|u_1\| \geq 0$ . Then  $C = \{x \in \mathcal{H} \mid \gamma_1 \leq \langle x | u \rangle \leq \gamma_2\} \neq \emptyset$ , where

$$u = \|u_2\| u_1, \quad \gamma_1 = -\eta_2 \|u_1\|, \quad \text{and} \quad \gamma_2 = \eta_1 \|u_2\|, \quad (28.15)$$

and

$$(\forall x \in \mathcal{H}) \quad P_C x = \begin{cases} x - \frac{\langle x | u \rangle - \gamma_1}{\|u\|^2} u, & \text{if } \langle x | u \rangle < \gamma_1; \\ x, & \text{if } \gamma_1 \leq \langle u | x \rangle \leq \gamma_2; \\ x - \frac{\langle x | u \rangle - \gamma_2}{\|u\|^2} u, & \text{if } \langle x | u \rangle > \gamma_2. \end{cases} \quad (28.16)$$

*Proof.* Let  $x \in \mathcal{H}$ .

(i), (ii), (iv), & (vi): Clear.

(iii)&(v): This follows from Example 28.16(iii).

(vii): Since  $u_1$  and  $u_2$  are linearly dependent and  $\langle u_1 | u_2 \rangle > 0$ , we have  $\|u_2\|u_1 = \|u_1\|u_2$ . Now set  $u = \|u_2\|u_1$ . We have  $\langle x | u_1 \rangle \leq \eta_1 \Leftrightarrow \|u_2\| \langle x | u_1 \rangle \leq \eta_1 \|u_2\| \Leftrightarrow \langle x | u \rangle \leq \eta_1 \|u_2\|$  and, similarly,  $\langle x | u_2 \rangle \leq \eta_2 \Leftrightarrow \langle x | u \rangle \leq \eta_2 \|u_1\|$ . Altogether,  $x \in C \Leftrightarrow \langle x | u \rangle \leq \eta$  and the formula for  $P_C$  follows from Example 28.16(iii).

(viii)&(ix): Set  $u = \|u_2\|u_1 = -\|u_1\|u_2$ ,  $\gamma_1 = -\eta_2\|u_1\|$ , and  $\gamma_2 = \eta_1\|u_2\|$ . Then  $\langle x | u_1 \rangle \leq \eta_1 \Leftrightarrow \|u_2\| \langle x | u_1 \rangle \leq \eta_1 \|u_2\| \Leftrightarrow \langle x | u \rangle \leq \gamma_2$ , and  $\langle x | u_2 \rangle \leq \eta_2 \Leftrightarrow \|u_1\| \langle x | u_2 \rangle \leq \eta_2 \|u_1\| \Leftrightarrow -\eta_2 \|u_1\| \leq \langle x | -\|u_1\|u_2 \rangle \Leftrightarrow \gamma_1 \leq \langle x | u \rangle$ . Altogether,  $x \in C \Leftrightarrow \gamma_1 \leq \langle x | u \rangle \leq \gamma_2$ , and the results follow from Example 28.17(iii)&(iv).  $\square$

**Proposition 28.19** *Let  $u_1$  and  $u_2$  be in  $\mathcal{H}$ , and let  $\eta_1$  and  $\eta_2$  be in  $\mathbb{R}$ . Suppose that  $\|u_1\|^2\|u_2\|^2 > |\langle u_1 | u_2 \rangle|^2$ , i.e.,  $\{u_1, u_2\}$  is linearly independent, set*

$$C = \{x \in \mathcal{H} \mid \langle x | u_1 \rangle \leq \eta_1\} \cap \{x \in \mathcal{H} \mid \langle x | u_2 \rangle \leq \eta_2\}, \quad (28.17)$$

*and let  $x \in \mathcal{H}$ . Then  $C \neq \emptyset$  and*

$$P_C x = x - \nu_1 u_1 - \nu_2 u_2, \quad (28.18)$$

*where exactly one of the following holds:*

- (i)  $\langle x | u_1 \rangle \leq \eta_1$  and  $\langle x | u_2 \rangle \leq \eta_2$ . Then  $\nu_1 = \nu_2 = 0$ .
- (ii)  $\|u_2\|^2(\langle x | u_1 \rangle - \eta_1) > \langle u_1 | u_2 \rangle(\langle x | u_2 \rangle - \eta_2)$  and  $\|u_1\|^2(\langle x | u_2 \rangle - \eta_2) > \langle u_1 | u_2 \rangle(\langle x | u_1 \rangle - \eta_1)$ . Then

$$\nu_1 = \frac{\|u_2\|^2(\langle x | u_1 \rangle - \eta_1) - \langle u_1 | u_2 \rangle(\langle x | u_2 \rangle - \eta_2)}{\|u_1\|^2\|u_2\|^2 - |\langle u_1 | u_2 \rangle|^2} > 0 \quad (28.19)$$

*and*

$$\nu_2 = \frac{\|u_1\|^2(\langle x | u_2 \rangle - \eta_2) - \langle u_1 | u_2 \rangle(\langle x | u_1 \rangle - \eta_1)}{\|u_1\|^2\|u_2\|^2 - |\langle u_1 | u_2 \rangle|^2} > 0. \quad (28.20)$$

- (iii)  $\langle x | u_2 \rangle > \eta_2$  and  $\|u_2\|^2(\langle x | u_1 \rangle - \eta_1) \leq \langle u_1 | u_2 \rangle(\langle x | u_2 \rangle - \eta_2)$ . Then

$$\nu_1 = 0 \quad \text{and} \quad \nu_2 = \frac{\langle x | u_2 \rangle - \eta_2}{\|u_2\|^2} > 0. \quad (28.21)$$

(iv)  $\langle x | u_1 \rangle > \eta_1$  and  $\|u_1\|^2(\langle x | u_2 \rangle - \eta_2) \leq \langle u_1 | u_2 \rangle (\langle x | u_1 \rangle - \eta_1)$ . Then

$$\nu_1 = \frac{\langle x | u_1 \rangle - \eta_1}{\|u_1\|^2} > 0 \quad \text{and} \quad \nu_2 = 0. \quad (28.22)$$

*Proof.* Set  $L: \mathcal{H} \rightarrow \mathbb{R}^2: y \mapsto (\langle y | u_1 \rangle, \langle y | u_2 \rangle)$  and

$$G = \begin{bmatrix} \|u_1\|^2 & \langle u_1 | u_2 \rangle \\ \langle u_1 | u_2 \rangle & \|u_2\|^2 \end{bmatrix}. \quad (28.23)$$

Since, by assumption,  $\det G \neq 0$ ,  $G$  is invertible. Hence,  $\text{ran } L = \text{ran } LL^* = \text{ran } G = \mathbb{R}^2$  and, therefore,  $C \neq \emptyset$ . Proposition 26.18 applied to the objective function  $(1/2)\|\cdot - x\|^2$  asserts the existence of  $\nu_1$  and  $\nu_2$  in  $\mathbb{R}_+$  such that

$$P_C x = x - \nu_1 u_1 - \nu_2 u_2, \quad (28.24)$$

and such that the feasibility conditions

$$\begin{cases} \langle x | u_1 \rangle - \nu_1 \|u_1\|^2 - \nu_2 \langle u_2 | u_1 \rangle \leq \eta_1, \\ \langle x | u_2 \rangle - \nu_1 \langle u_1 | u_2 \rangle - \nu_2 \|u_2\|^2 \leq \eta_2, \end{cases} \quad (28.25)$$

hold, as well as the complementary slackness conditions

$$\nu_1 (\langle x | u_1 \rangle - \nu_1 \|u_1\|^2 - \nu_2 \langle u_2 | u_1 \rangle - \eta_1) = 0 \quad (28.26)$$

and

$$\nu_2 (\langle x | u_2 \rangle - \nu_1 \langle u_1 | u_2 \rangle - \nu_2 \|u_2\|^2 - \eta_2) = 0. \quad (28.27)$$

The linear independence of  $\{u_1, u_2\}$  guarantees the uniqueness of  $(\nu_1, \nu_2)$ . This leads to four conceivable cases.

(a)  $\nu_1 = \nu_2 = 0$ : Then (28.24) yields  $P_C x = x$ , i.e.,  $x \in C$ . This verifies (i).

(b)  $\nu_1 > 0$  and  $\nu_2 > 0$ : In view of (28.23), (28.26) and (28.27) force  $G \begin{bmatrix} \nu_1 & \nu_2 \end{bmatrix}^\top = Lx - \begin{bmatrix} \eta_1 & \eta_2 \end{bmatrix}^\top$ , which yields (ii).

(c)  $\nu_1 = 0$  and  $\nu_2 > 0$ : Condition (28.27) forces  $\langle x | u_2 \rangle - \nu_2 \|u_2\|^2 = \eta_2$ , which yields the formula for  $\nu_2$  as well as the equivalence  $\nu_2 > 0 \Leftrightarrow \langle x | u_2 \rangle > \eta_2$ . In turn, (28.25) reduces to  $\langle x | u_1 \rangle - \eta_1 \leq \nu_2 \langle u_1 | u_2 \rangle$ , which yields  $\|u_2\|^2(\langle x | u_1 \rangle - \eta_1) \leq \langle u_1 | u_2 \rangle (\langle x | u_2 \rangle - \eta_2)$ , and  $\langle x | u_2 \rangle - \eta_2 = \nu_2 \|u_2\|^2$ . This verifies (iii).

(d)  $\nu_1 > 0$  and  $\nu_2 = 0$ : This is analogous to (c) and yields (iv).  $\square$

The following notation will be convenient.

**Definition 28.20** Set

$$H: \mathcal{H}^2 \rightarrow 2^{\mathcal{H}}: (x, y) \mapsto \{z \in \mathcal{H} \mid \langle z - y | x - y \rangle \leq 0\} \quad (28.28)$$

and



$$\begin{aligned}
Q: \mathcal{H}^3 &\rightarrow \mathcal{H} \\
(x, y, z) &\mapsto \begin{cases} z, & \text{if } \rho = 0 \text{ and } \chi \geq 0; \\
x + \left(1 + \frac{\chi}{\nu}\right)(z - y), & \text{if } \rho > 0 \text{ and } \chi\nu \geq \rho; \\
y + \frac{\nu}{\rho}(\chi(x - y) + \mu(z - y)), & \text{if } \rho > 0 \text{ and } \chi\nu < \rho, \end{cases} \\
\text{where } &\begin{cases} \chi = \langle x - y \mid y - z \rangle, \\
\mu = \|x - y\|^2, \\
\nu = \|y - z\|^2, \\
\rho = \mu\nu - \chi^2. \end{cases} \quad (28.29)
\end{aligned}$$

**Corollary 28.21 (Haugazeau)** *Let  $(x, y, z) \in \mathcal{H}^3$  and set*

$$C = H(x, y) \cap H(y, z). \quad (28.30)$$

*Moreover, set  $\chi = \langle x - y \mid y - z \rangle$ ,  $\mu = \|x - y\|^2$ ,  $\nu = \|y - z\|^2$ , and  $\rho = \mu\nu - \chi^2$ . Then exactly one of the following holds:*

- (i)  $\rho = 0$  and  $\chi < 0$ , in which case  $C = \emptyset$ .
- (ii) [ $\rho = 0$  and  $\chi \geq 0$ ] or  $\rho > 0$ , in which case  $C \neq \emptyset$  and

$$P_C x = Q(x, y, z). \quad (28.31)$$

*Proof.* We first observe that, by Cauchy–Schwarz,  $\rho \geq 0$ . Now set  $u_1 = x - y$ ,  $u_2 = y - z$ ,  $\eta_1 = \langle y \mid u_1 \rangle$ , and  $\eta_2 = \langle z \mid u_2 \rangle$ . Then  $H(x, y) = \{c \in \mathcal{H} \mid \langle c \mid u_1 \rangle \leq \eta_1\}$ , and  $H(y, z) = \{c \in \mathcal{H} \mid \langle c \mid u_2 \rangle \leq \eta_2\}$ .

(i): We have  $\|u_1\|^2\|u_2\|^2 = \langle u_1 \mid u_2 \rangle^2$  and  $\langle u_1 \mid u_2 \rangle < 0$ . Hence,  $\|u_2\|u_1 = -\|u_1\|u_2$  and, therefore,

$$\begin{aligned}
&\eta_1\|u_2\| + \eta_2\|u_1\| \\
&= \langle y \mid x - y \rangle \|u_2\| + \langle z \mid y - z \rangle \|u_1\| \\
&= \langle y \mid u_1 \rangle \|u_2\| + (\langle z - y \mid y - z \rangle + \langle y \mid y - z \rangle) \|u_1\| \\
&= \langle y \mid u_1 \rangle \|u_2\| + (-\|u_2\|^2 + \langle y \mid u_2 \rangle) \|u_1\| \\
&= \langle y \mid \|u_2\|u_1 + \|u_1\|u_2 \rangle - \|u_1\| \|u_2\|^2 \\
&= \langle y \mid 0 \rangle - \|u_1\| \|u_2\|^2 \\
&< 0. \quad (28.32)
\end{aligned}$$

We therefore deduce from Proposition 28.18(viii) that  $C = \emptyset$ .

(ii): We verify the formula in each of the following three cases:

(a)  $\rho = 0$  and  $\chi \geq 0$ : In view of (28.29), we must show that  $P_C x = z$ . We consider four subcases.

(a.1)  $x = y = z$ : Then  $C = \mathcal{H}$  and therefore  $P_C x = x = z$ .

(a.2)  $x \neq y = z$ : Then  $C = H(x, y)$  and it follows from (28.1) that  $P_C x = y = z$ .

(a.3)  $x = y \neq z$ : Then  $C = H(y, z)$  and it follows from (28.1) that  $P_C x = P_C y = z$ .

(a.4)  $x \neq y$  and  $y \neq z$ : Then  $\{u_1, u_2\}$  is linearly dependent and  $\chi = \langle u_1 | u_2 \rangle \geq 0$ . Hence,  $\chi > 0$ . Now set  $u = \|u_2\|u_1 = \|u_1\|u_2$  and  $\eta = \min\{\eta_1\|u_2\|, \eta_2\|u_1\|\}$ . We have

$$\langle x - z | u \rangle = \langle u_1 | u \rangle + \langle u_2 | u \rangle = \|u_1\|^2 \|u_2\| + \|u_1\| \|u_2\|^2 > 0. \quad (28.33)$$

On the other hand,  $\eta_1\|u_2\| - \eta_2\|u_1\| = \langle y | u_1 \rangle \|u_2\| - \langle z | u_2 \rangle \|u_1\| = \langle y - z | u \rangle = \langle u_2 | u \rangle = \|u_1\| \|u_2\|^2 > 0$ . Thus,  $\eta = \eta_2\|u_1\| = \langle z | u_2 \rangle \|u_1\| = \langle z | u \rangle$ , and (28.33) yields  $\langle x | u \rangle > \langle z | u \rangle = \eta$ . Hence, by Proposition 28.18(vii) and (28.33),

$$\begin{aligned} P_C x &= x + \frac{\eta - \langle x | u \rangle}{\|u\|^2} u \\ &= x - \frac{\langle x - z | u \rangle}{\|u\|^2} u \\ &= x - \frac{\|u_1\|^2 \|u_2\| + \|u_1\| \|u_2\|^2}{\|u_1\|^2 \|u_2\|^2} \|u_1\| u_2 \\ &= x - \frac{\|u_1\| u_2 + \|u_2\| u_2}{\|u_2\|} \\ &= x - u_1 - u_2 \\ &= z. \end{aligned} \quad (28.34)$$

(b)  $\rho > 0$  and  $\chi\nu \geq \rho$ : In view of (28.29), we must show that  $P_C x = x + (1 + \chi/\nu)(z - y)$ . We have  $\langle u_1 | u_2 \rangle \|u_2\|^2 \geq \|u_1\|^2 \|u_2\|^2 - \langle u_1 | u_2 \rangle^2 > 0$ , i.e.,

$$\langle u_1 | u_2 \rangle (\|u_2\|^2 + \langle u_1 | u_2 \rangle) \geq \|u_1\|^2 \|u_2\|^2 > \langle u_1 | u_2 \rangle^2. \quad (28.35)$$

Since  $\chi = \langle u_1 | u_2 \rangle > 0$ , we have

$$\begin{aligned} \langle x | u_2 \rangle - \eta_2 &= \langle x - z | u_2 \rangle \\ &= \langle u_1 + u_2 | u_2 \rangle \\ &= \langle u_1 | u_2 \rangle + \|u_2\|^2 \end{aligned} \quad (28.36)$$

$$> 0. \quad (28.37)$$

On the other hand, using (28.36) and (28.35), we obtain

$$\begin{aligned} \langle u_1 | u_2 \rangle (\langle x | u_2 \rangle - \eta_2) &= \langle u_1 | u_2 \rangle (\langle u_1 | u_2 \rangle + \|u_2\|^2) \\ &\geq \|u_1\|^2 \|u_2\|^2 \end{aligned}$$

$$= \|u_2\|^2 (\langle x | u_1 \rangle - \eta_1). \quad (28.38)$$

Altogether, (28.37), (28.38), Proposition 28.19(iii), and (28.36) yield

$$\begin{aligned} P_C x &= x - \frac{\langle x | u_2 \rangle - \eta_2}{\|u_2\|^2} u_2 \\ &= x + \frac{\|u_2\|^2 + \langle u_1 | u_2 \rangle}{\|u_2\|^2} (-u_2) \\ &= x + (1 + \chi/\nu)(z - y). \end{aligned} \quad (28.39)$$

(c)  $\rho > 0$  and  $\chi\nu < \rho$ : In view of (28.29), we must show that  $P_C x = y + (\nu/\rho)(\chi(x - y) + \mu(z - y))$ . We have

$$\begin{aligned} \|u_2\|^2 (\langle x | u_1 \rangle - \eta_1) &= \|u_2\|^2 \|u_1\|^2 \\ &= \rho + \chi^2 \end{aligned} \quad (28.40)$$

$$\begin{aligned} &> \chi(\nu + \chi) \\ &= \langle u_1 | u_2 \rangle \langle u_1 + u_2 | u_2 \rangle \\ &= \langle u_1 | u_2 \rangle \langle x - z | u_2 \rangle \\ &= \langle u_1 | u_2 \rangle (\langle x | u_2 \rangle - \eta_2). \end{aligned} \quad (28.41)$$

Next, by (28.36),

$$\begin{aligned} \|u_1\|^2 (\langle x | u_2 \rangle - \eta_2) &= \|u_1\|^2 (\langle u_1 | u_2 \rangle + \|u_2\|^2) \\ &> \|u_1\|^2 \langle u_1 | u_2 \rangle \\ &= (\langle x | u_1 \rangle - \eta_1) \langle u_1 | u_2 \rangle. \end{aligned} \quad (28.42)$$

Altogether, it follows from (28.40), (28.41), (28.42), and Proposition 28.19(ii) that

$$\begin{aligned} P_C x &= x - \left(1 - \frac{\chi\nu}{\rho}\right)(x - y) - \frac{\nu\mu}{\rho}(y - z) \\ &= y + \frac{\nu}{\rho}(\chi(x - y) + \mu(z - y)), \end{aligned} \quad (28.43)$$

which concludes the proof.  $\square$

## 28.4 Projections Involving Convex Cones

Projections onto convex cones were discussed in Proposition 6.27 and in Theorem 6.29. Here are further properties.

**Proposition 28.22** *Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ , let  $x \in \mathcal{H}$ , and let  $\rho \in \mathbb{R}_+$ . Then  $P_K(\rho x) = \rho P_K x$ .*

*Proof.* See Exercise 28.6.  $\square$

**Example 28.23** Let  $I$  be a totally ordered set, suppose that  $\mathcal{H} = \ell^2(I)$ , and set  $K = \ell^2_+(I)$ . Then  $P_K$  is weakly sequentially continuous and, for every  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ ,  $P_K x = (\max\{\xi_i, 0\})_{i \in I}$ .

*Proof.* This is a consequence of Proposition 28.4 (see also Example 6.28).  $\square$

**Example 28.24** Suppose that  $u$  is a nonzero vector in  $\mathcal{H}$  and set  $K = \mathbb{R}_+ u$ . Then

$$(\forall x \in \mathcal{H}) \quad P_K x = \begin{cases} \frac{\langle x | u \rangle}{\|u\|^2} u, & \text{if } \langle x | u \rangle \geq 0; \\ 0, & \text{otherwise.} \end{cases} \quad (28.44)$$

*Proof.* Since  $K^\ominus = \{x \in \mathcal{H} \mid \langle x | u \rangle \leq 0\}$ , the result follows from Theorem 6.29(i) and Example 28.16(iii).  $\square$

**Example 28.25** Suppose that  $\mathcal{H} = \mathbb{S}^N$ , let  $K = \mathbb{S}_+^N$  be the closed convex cone of symmetric positive semidefinite matrices, and let  $X \in \mathcal{H}$ . Then there exist a diagonal matrix  $\Lambda \in \mathbb{R}^{N \times N}$  and an orthogonal matrix  $U \in \mathbb{R}^{N \times N}$  such that  $X = U\Lambda U^\top$  and  $P_K X = U\Lambda_+ U^\top$ , where  $\Lambda_+$  is the diagonal matrix obtained from  $\Lambda$  by setting the negative entries equal to 0.

*Proof.* Set  $X_+ = U\Lambda_+ U^\top$  and  $\Lambda_- = \Lambda - \Lambda_+$ . Then  $X_+ \in K$  and  $X - X_+ = U\Lambda_- U^\top \in -K = K^\ominus$  by Example 6.25. Furthermore,  $\langle X_+ | X - X_+ \rangle = \text{tr}(X_+(X - X_+)) = \text{tr}(U\Lambda_+ U^\top U\Lambda_- U^\top) = \text{tr}(U\Lambda_+ \Lambda_- U^\top) = \text{tr}(\Lambda_+ \Lambda_-) = 0$ . Altogether, the result follows from Proposition 6.27.  $\square$

In the remainder of this section, we focus on the problem of projecting onto the intersection of a cone and a hyperplane.

**Proposition 28.26** Let  $K$  be a nonempty closed convex cone in  $\mathcal{H}$ , suppose that  $u \in K$  satisfies  $\|u\| = 1$ , let  $\eta \in \mathbb{R}_{++}$ , set  $C = K \cap \{x \in \mathcal{H} \mid \langle x | u \rangle = \eta\}$ , and let  $x \in \mathcal{H}$ . Then  $C \neq \emptyset$  and

$$P_C x = P_K(\bar{\nu}u + x), \quad (28.45)$$

where  $\bar{\nu} \in \mathbb{R}$  is a solution to the equation  $\langle P_K(\bar{\nu}u + x) | u \rangle = \eta$ .

*Proof.* The problem of finding  $P_C x$  is a special case of the primal problem (27.47) in Example 27.18, in which  $\mathcal{K} = \mathbb{R}$ ,  $L = \langle \cdot | u \rangle$ , and  $r = \eta$ . Since  $\eta \in \text{int } \langle K | u \rangle$ , we derive from Example 27.18 that the dual problem (27.48), which becomes, via Theorem 6.29(iii),

$$\underset{\nu \in \mathbb{R}}{\text{minimize}} \quad \phi(\nu), \quad \text{where} \quad \phi: \nu \mapsto \frac{1}{2} d_{K^\ominus}^2(\nu u + x) - \nu \eta, \quad (28.46)$$

has at least one solution  $\bar{\nu}$ . Furthermore,  $\bar{\nu}$  is characterized by  $\phi'(\bar{\nu}) = 0$ , i.e., using Proposition 12.31, by  $\langle P_K(\bar{\nu}u + x) | u \rangle = \eta$ , and the primal solution is  $P_C x = P_K(\bar{\nu}u + x)$ .  $\square$

Here is an application of Proposition 28.26 to the probability simplex.

**Example 28.27** Let  $N$  be a strictly positive integer, suppose that  $\mathcal{H} = \mathbb{R}^N$ , let  $\beta \in \mathbb{R}_{++}$ , set  $I = \{1, \dots, N\}$ , set  $C = \{(\xi_i)_{i \in I} \in \mathbb{R}_+^N \mid \sum_{i \in I} \xi_i = \beta\}$ , let  $x = (\xi_i)_{i \in I} \in \mathcal{H}$ , and set

$$\phi: \mathbb{R} \rightarrow \mathbb{R}: t \mapsto \sum_{i \in I} \max \{\xi_i + t, 0\}. \quad (28.47)$$

Then  $\phi$  is continuous and increasing, and there exists  $s \in \mathbb{R}$  such that  $\phi(s) = \beta$ . Furthermore,  $s$  is unique,  $P_C x = (\max \{\xi_i + s, 0\})_{i \in I}$ , and  $s \in ]-\xi, \beta - \xi]$ , where  $\xi = \max_{i \in I} \xi_i$ . (Thus,  $s$  may be found easily by the bisection method.)

*Proof.* Set  $K = \mathbb{R}_+^N$ ,  $u = (1, \dots, 1)/\sqrt{N} \in \mathbb{R}^N$ , and  $\eta = \beta/\sqrt{N}$ . Then  $\|u\| = 1$  and  $C = K \cap \{x \in \mathcal{H} \mid \langle x \mid u \rangle = \eta\}$ . For every  $i \in I$ , the function  $t \mapsto \max \{\xi_i + t, 0\}$  is increasing. Using Example 28.23, we deduce that  $\phi: \mathbb{R} \rightarrow \mathbb{R}: t \mapsto \sum_{i \in I} \max \{\xi_i + t, 0\} = \sqrt{N} \langle P_K(x + \sqrt{N}tu) \mid u \rangle$  is increasing, and, by Proposition 28.26, that the equation  $\phi(\bar{v}/\sqrt{N}) = \beta$  has a solution. Thus, there exists  $s \in \mathbb{R}$  such that  $\phi(s) = \beta$ . Since  $\beta > 0$ , there exists  $j \in I$  such that  $\xi_j + s > 0$ . Hence,  $t \mapsto \max \{\xi_j + t, 0\}$  is strictly increasing on some open interval  $U$  containing  $s$ , which implies that  $\phi$  is strictly increasing on  $U$  as well. This shows the uniqueness of  $s$ . Finally,  $\phi(-\xi) = \sum_{i \in I} \max \{\xi_i - \xi, 0\} = \sum_{i \in I} 0 = 0 < \beta \leq \sum_{i \in I} \max \{\xi_i - \xi + \beta, 0\} = \phi(-\xi + \beta)$ .  $\square$

## 28.5 Projections onto Epigraphs and Lower Level Sets

**Proposition 28.28** Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be convex and continuous, set  $C = \text{epi } f$ , and let  $(z, \zeta) \in (\mathcal{H} \times \mathbb{R}) \setminus C$ . Then the inclusion  $z \in x + (f(x) - \zeta)\partial f(x)$  has a unique solution, say  $\bar{x}$ , and  $P_C(z, \zeta) = (\bar{x}, f(\bar{x}))$ .

*Proof.* Set  $(\bar{x}, \bar{\xi}) = P_C(z, \zeta)$ . Then  $(\bar{x}, \bar{\xi})$  is the unique solution to the problem

$$\begin{aligned} & \underset{\substack{(x, \xi) \in \mathcal{H} \oplus \mathbb{R} \\ f(x) + (-1)\xi \leq 0}}{\text{minimize}} & \frac{1}{2} \|x - z\|^2 + \frac{1}{2} |\xi - \zeta|^2. \end{aligned} \quad (28.48)$$

In view of Proposition 26.18,  $(\bar{x}, \bar{\xi})$  is characterized by  $f(\bar{x}) \leq \bar{\xi}$  and the existence of  $\bar{\nu} \in \mathbb{R}_+$  such that  $\bar{\nu}(f(\bar{x}) - \bar{\xi}) = 0$ ,  $z - \bar{x} \in \bar{\nu}\partial f(\bar{x})$ , and  $\zeta - \bar{\xi} = -\bar{\nu}$ . If  $\bar{\nu} = 0$ , then  $(\bar{x}, \bar{\xi}) = (z, \zeta) \in C$ , which is impossible. Hence  $\bar{\nu} > 0$ . We conclude that  $\bar{\xi} = f(\bar{x})$  and therefore that  $\bar{\nu} = f(\bar{x}) - \zeta$ .  $\square$

**Example 28.29** Let  $N$  be a strictly positive integer, suppose that  $\mathcal{H} = \mathbb{R}^N$ , set  $I = \{1, \dots, N\}$ , set  $f: \mathcal{H} \rightarrow \mathbb{R}: (\xi_i)_{i \in I} \mapsto (1/2) \sum_{i \in I} \alpha_i |\xi_i|^2$ , where  $(\alpha_i)_{i \in I} \in \mathbb{R}_{++}^N$ . Suppose that  $(z, \zeta) \in (\mathcal{H} \times \mathbb{R}) \setminus (\text{epi } f)$  and set  $z = (\zeta_i)_{i \in I}$ . By Proposition 28.28, the system of equations  $\xi_i = \zeta_i - (f(x) - \zeta)\alpha_i \xi_i$ , where  $i \in I$ , has a unique solution  $\bar{x} = (\bar{\xi}_i)_{i \in I}$ . Now set  $\eta = f(\bar{x})$ . Then

$$(\forall i \in I) \quad \bar{\xi}_i = \frac{\zeta_i}{(\eta - \zeta)\alpha_i + 1} \quad (28.49)$$

and hence

$$\eta = \frac{1}{2} \sum_{i \in I} \alpha_i \left( \frac{\zeta_i}{(\eta - \zeta)\alpha_i + 1} \right)^2. \quad (28.50)$$

This one-dimensional equation can be solved numerically for  $\eta$ , and  $\bar{x}$  can then be recovered from (28.49).

**Proposition 28.30** *Let  $f: \mathcal{H} \rightarrow \mathbb{R}$  be convex and continuous, and let  $z \in \mathcal{H}$ . Suppose that  $\zeta \in ]\inf f(\mathcal{H}), f(z)[$  and set  $C = \text{lev}_{\leq \zeta} f$ . Then, in terms of the real variable  $\nu$ , the equation  $f(\text{Prox}_{\nu} f z) = \zeta$  has at least one solution in  $\mathbb{R}_{++}$  and, if  $\bar{\nu}$  is such a solution, then  $P_C z = \text{Prox}_{\bar{\nu}} f z$ .*

*Proof.* The vector  $P_C z$  is the unique solution to the convex optimization problem

$$\underset{\substack{x \in \mathcal{H} \\ f(x) \leq \zeta}}{\text{minimize}} \quad \frac{1}{2} \|x - z\|^2. \quad (28.51)$$

In view of Proposition 26.18, there exists  $\bar{\nu} \in \mathbb{R}_+$  such that  $f(P_C z) \leq \zeta$ ,  $\bar{\nu}(f(P_C z) - \zeta) = 0$ , and  $z - P_C z \in \bar{\nu} \partial f(P_C z)$ . Since  $z \notin C$ , we must have  $\bar{\nu} > 0$ . Hence,  $f(P_C z) = \zeta$  and the result follows from (16.30).  $\square$

**Example 28.31** Let  $C$  be a nonempty bounded closed convex subset of  $\mathcal{H}$  and recall that  $C^\odot$  denotes the polar set of  $C$  (see (7.6)). Suppose that  $z \in \mathcal{H} \setminus C^\odot$ . Then, in terms of the real variable  $\nu$ , the equation  $\nu = \langle z - \nu P_C(z/\nu) \mid \nu P_C(z/\nu) \rangle$  has a unique solution  $\bar{\nu} \in \mathbb{R}_{++}$ . Moreover,  $P_{C^\odot} z = z - \bar{\nu} P_C(z/\bar{\nu})$  and  $\bar{\nu} = \langle P_{C^\odot} z \mid z - P_{C^\odot} z \rangle$ .

*Proof.* Recall that  $C^\odot = \text{lev}_{\leq 1} \sigma_C = \text{lev}_{\leq 1} \iota_C^*$  and that our assumptions imply that  $\sigma_C(0) = 0 < 1 < \sigma_C(z)$ . By Proposition 28.30, there exists  $\bar{\nu} \in \mathbb{R}_{++}$  such that

$$P_{C^\odot} z = \text{Prox}_{\bar{\nu} \iota_C^*} z \quad \text{and} \quad \iota_C^*(P_{C^\odot} z) = 1. \quad (28.52)$$

It follows from Theorem 14.3(ii) and Example 12.25 that  $\text{Prox}_{\bar{\nu} \iota_C^*} z = z - \bar{\nu} P_C(z/\bar{\nu})$ , which provides the projection formula. In turn, we derive from (28.52) and Theorem 14.3(iii) that

$$\begin{aligned} 1 &= \iota_C^*(\text{Prox}_{\bar{\nu} \iota_C^*} z) + \iota_C(P_C(z/\bar{\nu})) \\ &= \langle \text{Prox}_{\bar{\nu} \iota_C^*} z \mid P_C(z/\bar{\nu}) \rangle \\ &= \langle P_{C^\odot} z \mid z - P_{C^\odot} z \rangle / \bar{\nu}, \end{aligned} \quad (28.53)$$

which yields  $\bar{\nu} = \langle P_{C^\odot} z \mid z - P_{C^\odot} z \rangle$ .  $\square$

**Example 28.32** Suppose that  $\mathcal{H} = \mathbb{R}^N$ , let  $A \in \mathbb{R}^{N \times N}$  be symmetric and positive semidefinite, and let  $u \in \mathbb{R}^N$ . Set  $f: \mathbb{R}^N \rightarrow \mathbb{R}: x \mapsto (1/2) \langle x \mid Ax \rangle + \langle x \mid u \rangle$ , let  $z \in \mathbb{R}^N$ , suppose that  $\zeta \in ]\inf f(\mathcal{H}), f(z)[$ , and set  $C = \text{lev}_{\leq \zeta} f$ .

Then  $\bar{x} = P_C z$  is the unique solution to the system of equations  $f(x) = \zeta$  and  $x = (\text{Id} + \bar{\nu}A)^{-1}(z - \bar{\nu}u)$ , where  $\bar{\nu} \in \mathbb{R}_{++}$ .

*Proof.* This follows from Proposition 28.30.  $\square$

## Exercises

**Exercise 28.1** Suppose that  $\mathcal{H} = \mathbf{H} \times \mathbb{R}$ , where  $\mathbf{H}$  is a real Hilbert space, let  $\rho \in \mathbb{R}_{++}$  and  $\alpha \in \mathbb{R}_{++}$ , and set  $C_1 = B(0; \rho) \times \mathbb{R}$  and  $C_2 = \mathbf{H} \times [-\alpha, +\alpha]$ . Provide formulas for the projection operators  $P_{C_1}$  and  $P_{C_2}$ , and check that the projection onto the cylinder  $C_1 \cap C_2$  is given by  $P_{C_1 \cap C_2} = P_{C_1} P_{C_2} = P_{C_2} P_{C_1}$ .

**Exercise 28.2** Show that Proposition 28.2(ii) fails if  $L^* \neq L^{-1}$ .

**Exercise 28.3** Prove Proposition 28.4 using (28.1).

**Exercise 28.4** Given finite families of  $(u_i)_{i \in I}$  in  $\mathcal{H} \setminus \{0\}$  and  $(\beta_i)_{i \in I}$  in  $\mathbb{R}$ , set  $C = \{x \in \mathcal{H} \mid \langle x \mid u_i \rangle \leq \beta_i\}$ . Let  $V$  be a closed linear subspace of  $\bigcap_{i \in I} \{u_i\}^\perp$  and set  $D = C \cap V^\perp$ . Show that  $D$  is a polyhedron in  $V^\perp$  and that  $P_C = P_V + P_D \circ P_{V^\perp}$ .

**Exercise 28.5** Prove Proposition 28.10.

**Exercise 28.6** Check Proposition 28.22.

**Exercise 28.7** Prove Example 28.24 using (28.1).

**Exercise 28.8** Let  $\alpha \in \mathbb{R}_{++}$  and set  $K_\alpha = \{(x, \rho) \in \mathcal{H} \times \mathbb{R} \mid \|x\| \leq \alpha\rho\}$ . Use Exercise 6.4 to show that  $K_\alpha^\ominus = -K_{1/\alpha}$  and that

$$(\forall (x, \rho) \in \mathcal{H} \times \mathbb{R}) \quad P_{K_\alpha}(x, \rho) = \begin{cases} (x, \rho), & \text{if } \|x\| \leq \alpha\rho; \\ (0, 0), & \text{if } \alpha\|x\| \leq -\rho; \\ \left( \frac{\alpha\|x\| + \rho}{\alpha^2 + 1} \left( \alpha \frac{x}{\|x\|}, 1 \right), \right), & \text{otherwise.} \end{cases} \quad (28.54)$$

**Exercise 28.9** Let  $z \in \mathcal{H}$  and let  $\gamma \in [0, 1]$ . Suppose that  $\|z\| = 1$  and set

$$K_{z, \gamma} = \{x \in \mathcal{H} \mid \gamma\|x\| - \langle x \mid z \rangle \leq 0\}. \quad (28.55)$$

Use Exercise 28.8 to check that  $K_{z, \gamma}^\ominus = K_{-z, \sqrt{1-\gamma^2}}$  and that

$$(\forall x \in \mathcal{H}) \quad P_{K_{z, \gamma}} x = \begin{cases} x, & \text{if } x \in K_{z, \gamma}; \\ 0, & \text{if } x \in K_{z, \gamma}^\ominus; \\ \delta y, & \text{otherwise,} \end{cases} \quad (28.56)$$

where  $y = \gamma z + \sqrt{1-\gamma^2}(x - \langle x \mid z \rangle z) / \|x - \langle x \mid z \rangle z\|$  and  $\delta = \langle x \mid y \rangle$ .

**Exercise 28.10** Consider Exercise 28.9 and its notation. Let  $\varepsilon \in \mathbb{R}_{++}$ , and let  $w \in \mathcal{H}$  be such that  $\|w\| = 1$  and  $(\forall x \in K_{z, \varepsilon/(2+\varepsilon)}) \langle x | w \rangle \geq 0$ . Prove that  $\|w - z\| < \varepsilon/2$ .

**Exercise 28.11** Let  $N$  be a strictly positive integer, let  $\beta \in \mathbb{R}_{++}$ , and set  $I = \{1, \dots, N\}$ . Using Example 28.27, show that the problem

$$\begin{aligned} & \text{minimize} && \|y\| \\ & y = (\eta_i)_{i \in I} \in \mathbb{R}_+^N \\ & \sum_{i \in I} \eta_i = \beta \end{aligned} \tag{28.57}$$

has a unique solution, namely  $(\beta/N)_{i \in I}$ .

**Exercise 28.12** Suppose that  $\mathcal{H} = \mathbb{R}$ , set  $f: \mathcal{H} \rightarrow \mathcal{H}: x \mapsto x^2$  and  $C = \text{epi } f$ , and let  $(z, t) \in \mathbb{R}^2 \setminus (\text{epi } f)$ . Then  $P_C(z, t) = (\bar{x}, \bar{x}^2)$ , where  $\bar{x} \in \mathbb{R}$  is the unique solution to the cubic equation  $2x^3 + (1 - 2t)x - z = 0$ .

**Exercise 28.13** Let  $N$  be a strictly positive integer, suppose that  $\mathcal{H} = \mathbb{R}^N$ , set  $I = \{1, \dots, N\}$ , let  $t \in \mathbb{R}_{++}$ , and let  $(\alpha_i)_{i \in I} \in \mathbb{R}_+^N$ . Set  $f: \mathbb{R}^N \rightarrow \mathbb{R}: x \mapsto (1/2) \sum_{i \in I} \alpha_i |\xi_i|^2$  and let  $z \in \mathbb{R}^N$  be such that  $f(z) > t$ . Set  $C = \text{lev}_{\leq t} f$  and  $x = P_C z$ . Show that, for every  $i \in I$ ,  $\xi_i = \zeta_i / (1 + \lambda \alpha_i)$ , where

$$\frac{1}{2} \sum_{i \in I} \alpha_i \left( \frac{\zeta_i}{1 + \lambda \alpha_i} \right)^2 = t. \tag{28.58}$$



# Chapter 29

## Best Approximation Algorithms

Best approximation algorithms were already discussed in Corollary 5.28, in Example 27.17, and in Example 27.18. In this chapter we provide further approaches for computing the projection onto the intersection of finitely many closed convex sets, employing the individual projectors, namely, Dykstra's algorithm and Haugazeau's algorithm. Applications of the latter to solving monotone inclusion and minimization problems with strongly convergent algorithms are given.

### 29.1 Dykstra's Algorithm

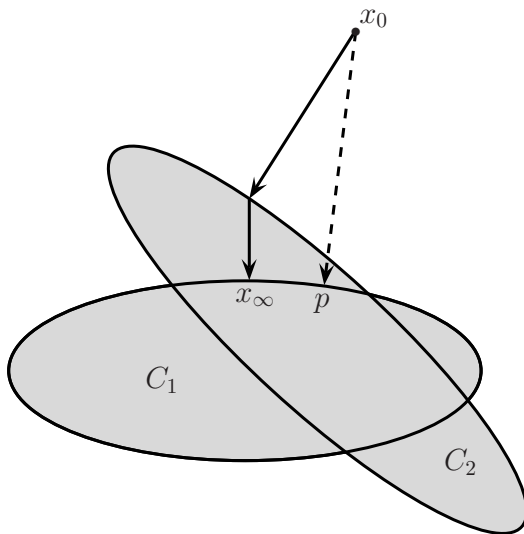
Corollary 5.23 features a weakly convergent periodic projection algorithm for finding a common point of closed convex sets. We saw in Corollary 5.28 that, when all the sets are closed affine subspaces, the iterates produced by this algorithm actually converge to the projection of the initial point  $x_0$  onto the intersection of the sets. As shown in [Figure 29.1](#), this is however not true in general. In this section, we describe a strongly convergent projection algorithm for finding the projection of a point onto the intersection of closed convex sets. The following technical result will be required.

**Lemma 29.1** *Let  $(\rho_n)_{n \in \mathbb{N}} \in \ell_+^2(\mathbb{N})$ , let  $m \in \mathbb{N}$ , and set  $(\forall n \in \mathbb{N}) \sigma_n = \sum_{k=0}^n \rho_k$ . Then  $\varliminf \sigma_n(\sigma_n - \sigma_{n-m-1}) = 0$ .*

*Proof.* Since, by Cauchy–Schwarz,

$$\sigma_n(\sigma_n - \sigma_{n-m-1}) \leq \sqrt{n+1} \sqrt{\sum_{k=0}^n \rho_k^2 (\sigma_n - \sigma_{n-m-1})}, \quad (29.1)$$

for every integer  $n > m$ , it suffices to show that  $\varliminf \sqrt{n+1}(\sigma_n - \sigma_{n-m-1}) = 0$ . We shall prove this by contradiction: assume that there exist  $\varepsilon \in \mathbb{R}_{++}$  and an



**Fig. 29.1** The method of alternating projections  $x_{n+1} = P_{C_1}P_{C_2}x_n$  converges (in one iteration) to  $x_\infty$  and fails to produce the best approximation  $p$  to  $x_0$  from  $C_1 \cap C_2$ .

integer  $n_0 > m$  such that, for every integer  $n \geq n_0$ ,  $\sigma_n - \sigma_{n-m-1} \geq \varepsilon/\sqrt{n+1}$  which, by Cauchy–Schwarz, implies that

$$\frac{\varepsilon^2}{n+1} \leq (\rho_{n-m} + \cdots + \rho_n)^2 \leq (m+1)(\rho_{n-m}^2 + \cdots + \rho_n^2). \quad (29.2)$$

Summing (29.2) over all integers  $n \geq n_0$  yields a contradiction.  $\square$

The periodic projection algorithm described next was first studied by Dykstra in the case of closed convex cones.

**Theorem 29.2 (Dykstra’s algorithm)** *Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , and let  $x_0 \in \mathcal{H}$ . Set*

$$i: \mathbb{N} \rightarrow I: n \mapsto 1 + \text{rem}(n-1, m), \quad (29.3)$$

where  $\text{rem}(\cdot, m)$  is the remainder function of the division by  $m$ . For every strictly positive integer  $n$ , set  $P_n = P_{C_n}$ , where  $C_n = C_{i(n)}$  if  $n > m$ . Moreover, set  $q_{-(m-1)} = \cdots = q_{-1} = q_0 = 0$  and

$$(\forall n \in \mathbb{N} \setminus \{0\}) \quad \begin{cases} x_n = P_n(x_{n-1} + q_{n-m}), \\ q_n = x_{n-1} + q_{n-m} - x_n. \end{cases} \quad (29.4)$$

Then  $x_n \rightarrow P_C x_0$ .

*Proof.* It follows from Theorem 3.14 that

$$(\forall n \in \mathbb{N} \setminus \{0\}) \quad x_n \in C_n \quad \text{and} \quad (\forall y \in C_n) \quad \langle x_n - y \mid q_n \rangle \geq 0. \quad (29.5)$$

Moreover, for every integer  $n \geq 1$ , (29.4) yields

$$x_{n-1} - x_n = q_n - q_{n-m} \quad (29.6)$$

and, therefore,

$$x_0 - x_n = \sum_{k=n-m+1}^n q_k. \quad (29.7)$$

Let  $z \in C$ , let  $n \in \mathbb{N}$ , and let  $x_{-(m-1)}, \dots, x_{-1}$  be arbitrary vectors in  $\mathcal{H}$ . We derive from (29.6) that

$$\begin{aligned} \|x_n - z\|^2 &= \|x_{n+1} - z\|^2 + \|x_n - x_{n+1}\|^2 + 2 \langle x_{n+1} - z \mid x_n - x_{n+1} \rangle \\ &= \|x_{n+1} - z\|^2 + \|x_n - x_{n+1}\|^2 + 2 \langle x_{n+1} - z \mid q_{n+1} - q_{n+1-m} \rangle \\ &= \|x_{n+1} - z\|^2 + \|x_n - x_{n+1}\|^2 + 2 \langle x_{n+1} - z \mid q_{n+1} \rangle \\ &\quad + 2 \langle x_{n+1-m} - x_{n+1} \mid q_{n+1-m} \rangle - 2 \langle x_{n-m+1} - z \mid q_{n-m+1} \rangle. \end{aligned} \quad (29.8)$$

Now let  $l \in \mathbb{N}$  be such that  $n \geq l$ . Using (29.8) and induction, we obtain

$$\begin{aligned} \|x_l - z\|^2 &= \|x_n - z\|^2 + \sum_{k=l+1}^n (\|x_k - x_{k-1}\|^2 + 2 \langle x_{k-m} - x_k \mid q_{k-m} \rangle) \\ &\quad + 2 \sum_{k=n-m+1}^n \langle x_k - z \mid q_k \rangle - 2 \sum_{k=l-(m-1)}^l \langle x_k - z \mid q_k \rangle. \end{aligned} \quad (29.9)$$

In particular, when  $l = 0$ , we see that

$$\begin{aligned} \|x_0 - z\|^2 &= \|x_n - z\|^2 + \sum_{k=1}^n (\|x_k - x_{k-1}\|^2 + 2 \langle x_{k-m} - x_k \mid q_{k-m} \rangle) \\ &\quad + 2 \sum_{k=n-m+1}^n \langle x_k - z \mid q_k \rangle. \end{aligned} \quad (29.10)$$

Since all the summands on the right-hand side of (29.10) lie in  $\mathbb{R}_+$ , it follows that  $(x_k)_{k \in \mathbb{N}}$  is bounded and

$$\sum_{k \in \mathbb{N}} \|x_{k-1} - x_k\|^2 < +\infty. \quad (29.11)$$

Using (29.7) and (29.5), we get

$$\begin{aligned}\langle z - x_n \mid x_0 - x_n \rangle &= \sum_{k=n-m+1}^n \langle z - x_k \mid q_k \rangle + \sum_{k=n-m+1}^n \langle x_k - x_n \mid q_k \rangle \\ &\leq \sum_{k=n-m+1}^n \langle x_k - x_n \mid q_k \rangle.\end{aligned}\quad (29.12)$$

On the other hand, for every  $k \in \{n-m+1, \dots, n-1\}$ , the identities

$$\begin{aligned}q_k &= q_k - 0 \\ &= q_k - q_{i(k)-m} \\ &= (q_k - q_{k-m}) + (q_{k-m} - q_{k-2m}) + \dots + (q_{i(k)} - q_{i(k)-m})\end{aligned}\quad (29.13)$$

and (29.6) result in

$$\sum_{k=n-m+1}^{n-1} \|q_k\| \leq \sum_{k=1}^{n-1} \|q_k - q_{k-m}\| = \sum_{k=1}^{n-1} \|x_{k-1} - x_k\|. \quad (29.14)$$

Hence, using the triangle inequality, we obtain

$$\begin{aligned}\sum_{k=n-m+1}^{n-1} |\langle x_k - x_n \mid q_k \rangle| &\leq \sum_{k=n-m+1}^{n-1} \|x_k - x_n\| \|q_k\| \\ &\leq \sum_{l=n-m+2}^n \|x_{l-1} - x_l\| \sum_{k=n-m+1}^{n-1} \|q_k\| \\ &\leq \sum_{l=n-m+2}^n \|x_{l-1} - x_l\| \sum_{k=1}^{n-1} \|x_{k-1} - x_k\|.\end{aligned}\quad (29.15)$$

In view of (29.11) and Lemma 29.1, we deduce that

$$\varliminf \sum_{k=n-m+1}^n |\langle x_k - x_n \mid q_k \rangle| = 0. \quad (29.16)$$

Let  $(x_{p_n})_{n \in \mathbb{N}}$  be a subsequence of  $(x_n)_{n \in \mathbb{N}}$  such that

$$\lim \sum_{k=p_n-m+1}^{p_n} |\langle x_k - x_{p_n} \mid q_k \rangle| = 0, \quad (29.17)$$

such that  $(x_{p_n})_{n \in \mathbb{N}}$  converges weakly, say  $x_{p_n} \rightharpoonup x \in \mathcal{H}$ , and such that  $\lim \|x_{p_n}\|$  exists. In view of the definition of  $i$ , we also assume that there exists  $j \in I$  such that  $(\forall n \in \mathbb{N}) i(p_n) = j$ . By construction,  $(x_{p_n})_{n \in \mathbb{N}}$  lies in  $C_j$ , which is weakly closed by Theorem 3.32. Hence  $x \in C_j$ . Furthermore,

since  $x_n - x_{n+1} \rightarrow 0$  by (29.11) and since, for every  $n \in \mathbb{N} \setminus \{0\}$ ,  $x_n \in C_n$ ,

$$x \in C. \quad (29.18)$$

Combining (29.12) and (29.17) yields

$$\overline{\lim} \langle z - x_{p_n} \mid x_0 - x_{p_n} \rangle \leq 0. \quad (29.19)$$

Hence, for every  $z \in C$ , it follows from Lemma 2.35 that

$$\begin{aligned} \langle z - x \mid x_0 - x \rangle &= \|x\|^2 - \langle z \mid x \rangle - \langle x \mid x_0 \rangle + \langle z \mid x_0 \rangle \\ &\leq \underline{\lim} (\|x_{p_n}\|^2 - \langle z \mid x_{p_n} \rangle - \langle x_{p_n} \mid x_0 \rangle + \langle z \mid x_0 \rangle) \\ &\leq \overline{\lim} (\|x_{p_n}\|^2 - \langle z \mid x_{p_n} \rangle - \langle x_{p_n} \mid x_0 \rangle + \langle z \mid x_0 \rangle) \\ &= \overline{\lim} \langle z - x_{p_n} \mid x_0 - x_{p_n} \rangle \\ &\leq 0. \end{aligned} \quad (29.20)$$

In view of (29.18) and (29.20), we derive from (3.6) that

$$x = P_C x_0. \quad (29.21)$$

Using (29.20), with  $z = x$ , yields  $\|x_{p_n}\| \rightarrow \|x\|$ . Since  $x_{p_n} \rightharpoonup x$ , we deduce from Corollary 2.42 that

$$x_{p_n} \rightarrow x. \quad (29.22)$$

In turn, by (29.12) and (29.17),

$$\begin{aligned} 0 &\leftarrow \langle x - x_{p_n} \mid x_0 - x_{p_n} \rangle \\ &= \sum_{k=p_n-m+1}^{p_n} \langle x - x_k \mid q_k \rangle + \sum_{k=p_n-m+1}^{p_n} \langle x_k - x_{p_n} \mid q_k \rangle \\ &\leq \sum_{k=p_n-m+1}^{p_n} \langle x_k - x_{p_n} \mid q_k \rangle \\ &\rightarrow 0. \end{aligned} \quad (29.23)$$

This implies that

$$\lim_{k=p_n-m+1}^{p_n} \langle x - x_k \mid q_k \rangle = 0. \quad (29.24)$$

We now show by contradiction that the entire sequence  $(x_n)_{n \in \mathbb{N}}$  converges to  $x$ . To this end, let us assume that there exist  $\varepsilon \in \mathbb{R}_{++}$  and a subsequence  $(x_{l_n})_{n \in \mathbb{N}}$  of  $(x_n)_{n \in \mathbb{N}}$  such that  $(\forall n \in \mathbb{N}) \|x_{l_n} - x\| \geq \varepsilon$ . After passing to a subsequence of  $(x_{l_n})_{n \in \mathbb{N}}$  and relabeling if necessary, we assume furthermore that  $(\forall n \in \mathbb{N}) l_n > p_n$ . Then (29.9), (29.5), and (29.24) yield

$$0 \leftarrow \|x_{p_n} - x\|^2$$

$$\begin{aligned}
&= \|x_{l_n} - x\|^2 + \sum_{k=p_n+1}^{l_n} (\|x_k - x_{k-1}\|^2 + 2 \langle x_{k-m} - x_k \mid q_{k-m} \rangle) \\
&\quad + 2 \sum_{k=l_n-m+1}^{l_n} \langle x_k - x \mid q_k \rangle - 2 \sum_{k=p_n-m+1}^{p_n} \langle x_k - x \mid q_k \rangle \\
&\geq \|x_{l_n} - x\|^2 + 2 \sum_{k=p_n-m+1}^{p_n} \langle x - x_k \mid q_k \rangle \\
&\geq 2 \sum_{k=p_n-m+1}^{p_n} \langle x - x_k \mid q_k \rangle \\
&\rightarrow 0.
\end{aligned} \tag{29.25}$$

It follows that  $x_{l_n} \rightarrow x$ , which is impossible.  $\square$

Theorem 29.2 allows for a different proof of Corollary 5.28; we leave the verification of the details as Exercise 29.1.

## 29.2 Haugazeau's Algorithm

In this section we consider the problem of finding the best approximation to a point  $z$  from the set of common fixed points of firmly nonexpansive operators.

**Theorem 29.3** *Let  $(T_i)_{i \in I}$  be a finite family of firmly nonexpansive operators from  $\mathcal{H}$  to  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , and let  $x_0 \in \mathcal{H}$ . Let  $i: \mathbb{N} \rightarrow I$  be such that there exists a strictly positive integer  $m$  for which*

$$(\forall i \in I)(\forall n \in \mathbb{N}) \quad i \in \{i(n), \dots, i(n+m-1)\}, \tag{29.26}$$

and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Q(x_0, x_n, T_{i(n)}x_n), \tag{29.27}$$

where  $Q$  is defined in (28.29). Then  $x_n \rightarrow P_C x_0$ .

*Proof.* Throughout, the notation (28.28) will be used. We first observe that Corollary 4.15 implies that  $C$  is a nonempty closed convex set. Also, for every nonempty closed convex subset  $D$  of  $\mathcal{H}$ , (28.1) implies that

$$(\forall x \in \mathcal{H}) \quad P_D x \in D \quad \text{and} \quad D \subset H(x, P_D x). \tag{29.28}$$

In view of Corollary 28.21, to check that the sequence  $(x_n)_{n \in \mathbb{N}}$  is well defined, we must show that  $(\forall n \in \mathbb{N}) \quad H(x_0, x_n) \cap H(x_n, T_{i(n)}x_n) \neq \emptyset$ . To this end, it is sufficient to show that  $C \subset \bigcap_{n \in \mathbb{N}} H(x_0, x_n) \cap H(x_n, T_{i(n)}x_n)$ , i.e., since (29.26) and Corollary 4.16 yield

$$C = \bigcap_{n \in \mathbb{N}} \text{Fix } T_{i(n)} \subset \bigcap_{n \in \mathbb{N}} H(x_n, T_{i(n)}x_n), \quad (29.29)$$

that  $C \subset \bigcap_{n \in \mathbb{N}} H(x_0, x_n)$ . For  $n = 0$ , it is clear that  $C \subset H(x_0, x_n) = \mathcal{H}$ . Furthermore, for every  $n \in \mathbb{N}$ , it follows from (29.29), (29.28), and Corollary 28.21 that

$$\begin{aligned} C \subset H(x_0, x_n) &\Rightarrow C \subset H(x_0, x_n) \cap H(x_n, T_{i(n)}x_n) \\ &\Rightarrow C \subset H(x_0, Q(x_0, x_n, T_{i(n)}x_n)) \\ &\Leftrightarrow C \subset H(x_0, x_{n+1}), \end{aligned} \quad (29.30)$$

which establishes the assertion by induction. Now let  $n \in \mathbb{N}$ . We observe that Corollary 28.21 yields

$$x_{n+1} \in H(x_0, x_n) \cap H(x_n, T_{i(n)}x_n). \quad (29.31)$$

Hence, since  $x_n$  is the projection of  $x_0$  onto  $H(x_0, x_n)$  and since  $x_{n+1} = Q(x_0, x_n, T_{i(n)}x_n) \in H(x_0, x_n)$ , we have  $\|x_0 - x_n\| \leq \|x_0 - x_{n+1}\|$ . On the other hand, since  $P_C x_0 \in C \subset H(x_0, x_n)$ , we have  $\|x_0 - x_n\| \leq \|x_0 - P_C x_0\|$ . Altogether,  $(\|x_0 - x_n\|)_{n \in \mathbb{N}}$  converges and

$$\lim \|x_0 - x_n\| \leq \|x_0 - P_C x_0\|. \quad (29.32)$$

For every  $n \in \mathbb{N}$ , the inclusion  $x_{n+1} \in H(x_0, x_n)$  implies that

$$\begin{aligned} \|x_0 - x_{n+1}\|^2 - \|x_0 - x_n\|^2 &= \|x_{n+1} - x_n\|^2 + 2 \langle x_{n+1} - x_n \mid x_n - x_0 \rangle \\ &\geq \|x_{n+1} - x_n\|^2, \end{aligned} \quad (29.33)$$

and it follows from the convergence of  $(\|x_0 - x_n\|)_{n \in \mathbb{N}}$  that

$$x_{n+1} - x_n \rightarrow 0. \quad (29.34)$$

In turn since, for every  $n \in \mathbb{N}$ , the inclusion  $x_{n+1} \in H(x_n, T_{i(n)}x_n)$  implies that

$$\begin{aligned} \|x_{n+1} - x_n\|^2 &= \|x_{n+1} - T_{i(n)}x_n\|^2 + 2 \langle x_{n+1} - T_{i(n)}x_n \mid T_{i(n)}x_n - x_n \rangle \\ &\quad + \|x_n - T_{i(n)}x_n\|^2 \\ &\geq \|x_{n+1} - T_{i(n)}x_n\|^2 + \|x_n - T_{i(n)}x_n\|^2, \end{aligned} \quad (29.35)$$

we obtain

$$x_n - T_{i(n)}x_n \rightarrow 0. \quad (29.36)$$

Now let  $i \in I$ , and let  $x$  be a weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$ . In view of (29.26), there exist sequences  $(k_n)_{n \in \mathbb{N}}$  and  $(p_n)_{n \in \mathbb{N}}$  in  $\mathbb{N}$  such that  $x_{k_n} \rightharpoonup x$  and

$$(\forall n \in \mathbb{N}) \quad \begin{cases} k_n \leq p_n \leq k_n + m - 1 < k_{n+1} \leq p_{n+1}, \\ i = i(p_n). \end{cases} \quad (29.37)$$

Moreover, it follows from (29.34) that

$$\begin{aligned} \|x_{p_n} - x_{k_n}\| &\leq \sum_{l=k_n}^{k_n+m-2} \|x_{l+1} - x_l\| \\ &\leq (m-1) \max_{k_n \leq l \leq k_n+m-2} \|x_{l+1} - x_l\| \\ &\rightarrow 0, \end{aligned} \quad (29.38)$$

and, in turn, that  $x_{p_n} \rightharpoonup x$ . Hence, since  $x_{p_n} - T_i x_{p_n} = x_{p_n} - T_{i(p_n)} x_{p_n} \rightarrow 0$  by (29.36), we deduce from Corollary 4.18 that  $x \in \text{Fix } T_i$  and, since  $i$  was arbitrarily chosen in  $I$ , that  $x \in C$ . Bringing into play Lemma 2.35 and (29.32), we obtain

$$\|x_0 - P_C x_0\| \leq \|x_0 - x\| \leq \underline{\lim} \|x_0 - x_{k_n}\| \leq \|x_0 - P_C x_0\|. \quad (29.39)$$

Hence, since  $C$  is a Chebyshev set,  $x = P_C x_0$  is the unique weak sequential cluster point of  $(x_n)_{n \in \mathbb{N}}$  and it follows from Lemma 2.38 that  $x_n \rightharpoonup P_C x_0$ . Thus, appealing to Lemma 2.35 and (29.32), we obtain

$$\|x_0 - P_C x_0\| \leq \underline{\lim} \|x_0 - x_n\| = \lim \|x_0 - x_n\| \leq \|x_0 - P_C x_0\| \quad (29.40)$$

and, therefore,  $\|x_0 - x_n\| \rightarrow \|x_0 - P_C x_0\|$ . Since  $x_0 - x_n \rightharpoonup x_0 - P_C x_0$ , it follows from Corollary 2.42 that  $x_0 - x_n \rightarrow x_0 - P_C x_0$ .  $\square$

Our first application yields a strongly convergent proximal-point algorithm that finds the zero of a maximally monotone operator at minimal distance from the starting point.

**Corollary 29.4** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone and such that  $0 \in \text{ran } A$ , and let  $x_0 \in \mathcal{H}$ . Set  $(\forall n \in \mathbb{N}) \ x_{n+1} = Q(x_0, x_n, J_A x_n)$ , where  $Q$  is defined in (28.29). Then  $x_n \rightarrow P_{\text{zer } A} x_0$ .*

*Proof.* Set  $T_1 = J_A$ . Then  $T_1$  is firmly nonexpansive by Corollary 23.8, and Proposition 23.38 asserts that  $\text{Fix } T_1 = \text{zer } A \neq \emptyset$ . Thus, the result is an application of Theorem 29.3 with  $I = \{1\}$ ,  $m = 1$ , and  $i: n \mapsto 1$ .  $\square$

The next result provides a strongly convergent forward-backward algorithm.

**Corollary 29.5** *Let  $A: \mathcal{H} \rightarrow 2^{\mathcal{H}}$  be maximally monotone, let  $\beta \in \mathbb{R}_{++}$ , let  $B: \mathcal{H} \rightarrow \mathcal{H}$  be  $\beta$ -cocoercive, and let  $\gamma \in ]0, 2\beta[$ . Suppose that  $\text{zer}(A+B) \neq \emptyset$ , let  $x_0 \in \mathcal{H}$ , and set*



$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma Bx_n, \\ z_n = (1/2)(x_n + J_{\gamma A}y_n), \\ x_{n+1} = Q(x_0, x_n, z_n), \end{cases} \quad (29.41)$$

where  $Q$  is defined in (28.29). Then  $x_n \rightarrow P_{\text{zer}(A+B)} x_0$ .

*Proof.* Set  $T = (1/2)(\text{Id} + J_{\gamma A} \circ (\text{Id} - \gamma B))$  and note that (29.41) yields

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Q(x_0, x_n, Tx_n). \quad (29.42)$$

By Corollary 23.10(i),  $J_{\gamma A}$  is nonexpansive. Moreover, by Proposition 4.33 and Remark 4.24(i),  $\text{Id} - \gamma B$  is nonexpansive. Thus,  $J_{\gamma A} \circ (\text{Id} - \gamma B)$  is nonexpansive and, appealing to Proposition 4.2, we deduce that  $T$  is firmly nonexpansive. Altogether, since  $\text{Fix } T = \text{Fix}(2T - \text{Id}) = \text{zer}(A + B)$  by Proposition 25.1(iv), the result follows from Theorem 29.3 with  $I = \{1\}$ ,  $m = 1$ , and  $i: n \mapsto 1$ .  $\square$

Using Remark 25.16, we can apply Corollary 29.5 to variational inequalities. Here is another application.

**Example 29.6** Let  $f \in \Gamma_0(\mathcal{H})$ , let  $g: \mathcal{H} \rightarrow \mathbb{R}$  be convex and differentiable with a  $1/\beta$ -Lipschitz continuous gradient for some  $\beta \in \mathbb{R}_{++}$ , and let  $\gamma \in ]0, 2\beta[$ . Furthermore, suppose that  $\text{Argmin}(f + g) \neq \emptyset$ , let  $x_0 \in \mathcal{H}$ , and set

$$(\forall n \in \mathbb{N}) \quad \begin{cases} y_n = x_n - \gamma \nabla g(x_n), \\ z_n = (1/2)(x_n + \text{Prox}_{\gamma f} y_n), \\ x_{n+1} = Q(x_0, x_n, z_n), \end{cases} \quad (29.43)$$

where  $Q$  is defined in (28.29). Then  $x_n \rightarrow P_{\text{Argmin}(f+g)} x_0$ .

*Proof.* This is an application of Corollary 29.5 to  $A = \partial f$  and  $B = \nabla g$ , using Corollary 18.16. Indeed,  $A$  and  $B$  are maximally monotone by Theorem 20.40 and, since  $\text{dom } g = \mathcal{H}$ , Corollary 26.3 yields  $\text{Argmin}(f + g) = \text{zer}(A + B)$ .  $\square$

**Remark 29.7** The principle employed in the proof of Corollary 29.5 extends as follows. Let  $(R_i)_{i \in I}$  be a finite family of nonexpansive operators from  $\mathcal{H}$  to  $\mathcal{H}$  such that  $\bigcap_{i \in I} \text{Fix } R_i \neq \emptyset$  and set  $(\forall i \in I) T_i = (1/2)(R_i + \text{Id})$ . Then  $\bigcap_{i \in I} \text{Fix } T_i = \bigcap_{i \in I} \text{Fix } R_i$ , and it follows from Proposition 4.2 that the operators  $(T_i)_{i \in I}$  are firmly nonexpansive. Thus, Theorem 29.3 can be used to find the best approximation to a point  $x_0 \in \mathcal{H}$  from  $\bigcap_{i \in I} \text{Fix } R_i$ .

Our last result describes an alternative to Dykstra's algorithm of Theorem 29.2 for finding the projection onto the intersection of closed convex sets using the projectors onto the individual sets periodically.

**Corollary 29.8 (Haugazeau's algorithm)** *Let  $m$  be a strictly positive integer, set  $I = \{1, \dots, m\}$ , let  $(C_i)_{i \in I}$  be a family of closed convex subsets of  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} C_i \neq \emptyset$ , and let  $x_0 \in \mathcal{H}$ . For every  $i \in I$ , denote by*

$P_i$  the projector onto  $C_i$ . Let  $\text{rem}(\cdot, m)$  be the remainder function of division by  $m$ , and set

$$(\forall n \in \mathbb{N}) \quad x_{n+1} = Q(x_0, x_n, P_{1+\text{rem}(n-1, m)}x_n), \quad (29.44)$$

where  $Q$  is defined in (28.29). Then  $x_n \rightarrow P_C x_0$ .

*Proof.* In view of Proposition 4.8, this is an application of Theorem 29.3 with  $(T_i)_{i \in I} = (P_i)_{i \in I}$  and  $i: n \mapsto 1 + \text{rem}(n - 1, m)$ .  $\square$

## Exercises

**Exercise 29.1** Use Theorem 29.2 to derive Corollary 5.28.

**Exercise 29.2** Consider the application of Dykstra's algorithm (29.4) to  $m = 2$  intersecting closed convex subsets  $C$  and  $D$  of  $\mathcal{H}$ .

(i) Show that the algorithm can be cast in the following form:

$$p_0 = q_0 = 0 \quad \text{and} \quad (\forall n \in \mathbb{N}) \quad \begin{cases} y_n = P_C(x_n + p_n), \\ p_{n+1} = x_n + p_n - y_n, \\ x_{n+1} = P_D(y_n + q_n), \\ q_{n+1} = y_n + q_n - x_{n+1}. \end{cases} \quad (29.45)$$

(ii) Suppose that  $\mathcal{H} = \mathbb{R}^2$ ,  $x_0 = (2, 2)$ ,  $C = \{(\xi_1, \xi_2) \in \mathcal{H} \mid \xi_2 \leq 0\}$ , and  $D = \{(\xi_1, \xi_2) \in \mathcal{H} \mid \xi_1 + \xi_2 \leq 0\}$ . Compute the values of  $x_n$  for  $n \in \{0, \dots, 5\}$  and draw a picture showing the progression of the iterates toward  $P_{C \cap D} x_0$ .

**Exercise 29.3** Let  $T: \mathcal{H} \rightarrow \mathcal{H}$  be nonexpansive and such that  $\text{Fix } T \neq \emptyset$ . Devise a strongly convergent algorithm to find the minimal norm fixed point of  $T$ .

**Exercise 29.4** Let  $(T_i)_{i \in I}$  be a finite family of nonexpansive operators from  $\mathcal{H}$  to  $\mathcal{H}$  such that  $C = \bigcap_{i \in I} \text{Fix } T_i \neq \emptyset$ , let  $x_0 \in \mathcal{H}$ , let  $(\omega_i)_{i \in I}$  be real numbers in  $]0, 1]$  such that  $\sum_{i \in I} \omega_i = 1$ , and suppose that  $(\alpha_i)_{i \in I}$  are real numbers in  $]0, 1[$  such that, for every  $i \in I$ ,  $T_i$  is  $\alpha_i$ -averaged. Devise a strongly convergent algorithm involving  $(\alpha_i)_{i \in I}$  to find  $P_C x_0$ .

# Bibliographical Pointers

Seminal references for convex analysis, monotone operator theory, and non-expansive operators are [40], [58], [65], [121], [150], [166], [186], [195], [196], [219], [262], [263]. Historical comments can be found in [141], [188], [189], [208], [219], [226], [264], [265], [266], [267].

Further accounts and applications can be found in [6], [9], [14], [15], [16], [28], [37], [38], [46], [48], [50], [51], [55], [67], [70], [76], [80], [82], [85], [98], [99], [100], [106], [110], [112], [117], [124], [125], [126], [128], [136], [142], [143], [144], [147], [156], [159], [160], [165], [172], [173], [182], [188], [189], [199], [203], [207], [210], [213], [224], [226], [228], [230], [233], [235], [245], [253], [258], [259], [261], [264], [265], [266], [267].

It is virtually impossible to compile a complete bibliography on the topics covered by this book, most of which are quite mature. What follows is an attempt to list work that is relevant to the individual chapters, without making any claim to exhaustiveness.

CHAPTER 1: [3], [13], [16], [37], [50], [59], [81], [102], [108], [109], [112], [140], [151], [160], [196], [226], [229], [243], [258], [261].

CHAPTER 2: [1], [3], [53], [54], [58], [96], [99], [100], [117], [134], [139], [144], [159], [169], [174], [182], [200], [201], [227], [229], [243].

CHAPTER 3: [100], [131], [134], [144], [155], [219], [238], [261].

CHAPTER 4: [18], [24], [63], [68], [69], [89], [125], [126], [212], [263].

CHAPTER 5: [17], [18], [22], [24], [57], [61], [65], [68], [69], [77], [79], [83], [84], [86], [88], [89], [113], [114], [115], [120], [130], [132], [135], [158], [176], [178], [211], [255].

CHAPTER 6: [41], [43], [44], [54], [144], [145], [152], [157], [190], [233], [261], [263].

CHAPTER 7: [39], [142], [144], [203].

CHAPTER 8: [41], [48], [52], [54], [142], [143], [175], [196], [217], [219], [261].

CHAPTER 9: [3], [46], [58], [196], [219], [233], [261].

CHAPTER 10: [55], [74], [133], [165], [206], [219], [254], [260], [261].

CHAPTER 11: [50], [103], [125], [166], [206], [256], [261].

CHAPTER 12: [5], [11], [138], [142], [190], [191], [193], [195], [196], [219], [226], [261].

CHAPTER 13: [16], [46], [75], [142], [195], [196], [217], [219], [224], [226], [261].

CHAPTER 14: [24], [31], [48], [95], [180], [194], [195], [196], [219], [226], [261].

CHAPTER 15: [7], [44], [45], [46], [71], [129], [144], [149], [196], [216], [219], [237], [261].

CHAPTER 16: [12], [24], [48], [62], [116], [187], [192], [196], [215], [219], [224], [261].

CHAPTER 17: [25], [41], [46], [48], [124], [196], [203], [219], [239], [261].

CHAPTER 18: [19], [25], [26], [30], [41], [47], [48], [49], [111], [124], [203], [219], [250], [261].

CHAPTER 19: [15], [91], [105], [112], [147], [216], [219], [224], [261].

CHAPTER 20: [21], [27], [32], [33], [58], [70], [72], [73], [122], [134], [181], [195], [196], [204], [233], [235], [237], [265], [266].

CHAPTER 21: [23], [36], [42], [58], [66], [70], [72], [97], [100], [153], [185], [186], [195], [203], [218], [219], [222], [232], [233], [234], [235], [236], [252], [257], [265], [266].

CHAPTER 22: [21], [78], [148], [220], [266].

CHAPTER 23: [6], [23], [58], [92], [94], [95], [107], [127], [154], [163], [186], [225], [251].

CHAPTER 24: [2], [10], [27], [32], [34], [60], [93], [122], [123], [164], [197], [198], [202], [221], [232], [234], [235], [237], [265], [266].

CHAPTER 25: [8], [20], [64], [89], [90], [104], [107], [118], [119], [163], [167], [168], [170], [171], [183], [184], [231], [241], [244], [248], [249], [264].

CHAPTER 26: [59], [141], [142], [143], [147], [162], [219], [242], [246], [261], [264].

CHAPTER 27: [8], [89], [91], [95], [104], [166], [177], [179], [183], [209], [219], [247], [248].

CHAPTER 28: [22], [35], [100], [137], [263].

CHAPTER 29: [29], [56], [87], [137], [161], [205], [240].

# Symbols and Notation

## Real line:

$\mathbb{R}$		The set of real numbers
$x \geq 0$	p. 4	The real number $x$ is positive
$x > 0$	p. 4	The real number $x$ is strictly positive
$x \leq 0$	p. 4	The real number $x$ is negative
$x < 0$	p. 4	The real number $x$ is strictly negative
$\mathbb{R}_+$	p. 5	The set of positive real numbers $[0, +\infty[$
$\mathbb{R}_{++}$	p. 5	The set of strictly positive real numbers $]0, +\infty[$
$\mathbb{R}_-$	p. 5	The set of negative real numbers $]-\infty, 0]$
$\mathbb{R}_{--}$	p. 5	The set of strictly negative real numbers $]-\infty, 0[$
$\mathbb{Q}$		The set of rational numbers
$\mathbb{Z}$		The set of integers
$\mathbb{N}$	p. 4	The set of positive integers $\{0, 1, \dots\}$
$\inf, \min$	p. 5	Infimum and minimum
$\sup, \max$	p. 5	Supremum and maximum
$\overline{\lim} \xi_a$	p. 5	Limit superior of a net $(\xi_a)_{a \in A}$
$\underline{\lim} \xi_a$	p. 5	Limit inferior of a net $(\xi_a)_{a \in A}$
$\alpha^+$		$\max\{\alpha, 0\}$
$\alpha \downarrow \mu$		$\alpha$ is in $] \mu, +\infty[$ and converges to $\mu$ .
$\text{rem}(n, m)$		Remainder of division of $n$ by $m$

## Sets:

$2^{\mathcal{X}}$	p. 2	Power set of a set $\mathcal{X}$
$C \times D$		Cartesian product of the sets $C$ and $D$
$C + D$	p. 1	Minkowski sum of the sets $C$ and $D$
$C - D$	p. 1	Minkowski difference of the sets $C$ and $D$
$\lambda C$	p. 1	Scaling of a set $C$ by a real number $\lambda$
$AC$	p. 1	$AC = \bigcup_{\lambda \in A} \lambda C$ , where $A \subset \mathbb{R}$

$Az$	p. 1	$Az = \{\lambda z \mid \lambda \in A\}$ , where $A \subset \mathbb{R}$
$z + C$	p. 1	Translation of a set $C$ by a vector $z$
$C - z$	p. 1	Translation of a set $C$ by a vector $-z$
$\text{span } C$	p. 1	Linear span of a set $C$
$\overline{\text{span}} C$	p. 1	Closed linear span of a set $C$
$\text{aff } C$	p. 1	Affine hull of a set $C$
$\overline{\text{aff}} C$	p. 1	Closed affine hull of a set $C$
$A(C)$	p. 2	Image of a set $C$ by an operator $A$
$\text{diam}(C)$	p. 16	Diameter of a set $C$
$\overline{C}$	p. 7	Closure of a set $C$
$\text{int } C$	p. 7	Interior of a set $C$
$\text{bdry } C$	p. 7	Boundary of a set $C$
$\text{core } C$	p. 90	Core of a set $C$
$\text{sri } C$	p. 90	Strong relative interior of a set $C$
$\text{qri } C$	p. 91	Quasirelative interior of a set $C$
$\text{ri } C$	p. 91	Relative interior of a set $C$
$\text{conv } C$	p. 44	Convex hull of a set $C$
$\overline{\text{conv}} C$	p. 44	Closed convex hull of a set $C$
$P_C$	p. 44	Projector onto a nonempty closed convex set $C$
$Q(x, y, z)$	p. 423	Projector arising in Haugazeau's algorithm
$\text{cone } C$	p. 87	Conical hull of a set $C$
$\overline{\text{cone}} C$	p. 87	Closed conical hull of a set $C$
$C^\perp$	p. 27	Orthogonal complement of a set $C$
$C^\ominus$	p. 96	Polar cone of a set $C$
$C^\oplus$	p. 96	Dual cone of a set $C$
$C^\odot$	p. 110	Polar set of a set $C$
$N_C$	p. 101	Normal cone operator of a set $C$
$T_C$	p. 100	Tangent cone operator of a set $C$
$\text{rec } C$	p. 103	Recession cone of a set $C$
$\text{bar } C$	p. 103	Barrier cone of a set $C$
$\text{spts } C$	p. 107	Set of support points of a set $C$
$\overline{\text{spts}} C$	p. 107	Closure of the set of support points of a set $C$
$\sigma_C$	p. 109	Support function of a set $C$
$1_C$		Characteristic function of a set $C$
$\iota_C$	p. 12	Indicator function of a set $C$
$d_C$	p. 16	Distance function to a set $C$
$m_C$	p. 120	Minkowski gauge of a set $C$
$H(x, y)$	p. 422	Half-space arising in Haugazeau's algorithm
$[x, y], ]x, y],$	p. 1	Line segments between $x$ and $y$
$[x, y[, ]x, y[$		

## Topology:

$\mathcal{V}(x)$	p. 7	Family of all neighborhoods of $x$
$x_a \rightarrow x$	p. 7	The net $(x_a)_{a \in A}$ converges to $x$

$\overline{C}$	p. 7	Closure of a set $C$
$\text{int } C$	p. 7	Interior of a set $C$
$\text{bdry } C$	p. 7	Boundary of a set $C$
$B(x; \rho)$	p. 16	Closed ball of center $x$ and radius $\rho$

## Functions:

$\Gamma(\mathcal{H})$	p. 129	Set of lower semicontinuous convex functions from $\mathcal{H}$ to $[-\infty, +\infty]$
$\Gamma_0(\mathcal{H})$	p. 132	Set of proper lower semicontinuous convex functions from $\mathcal{H}$ to $] -\infty, +\infty]$
$\bigoplus_{i \in I} f_i$	p. 28	Separable sum of functions
$\tau_y f$		Translation of a function $f$ by a vector $y$
$f^\vee$	p. 3	Reversal of a function $f$
$f _C$		Restriction of a function $f$ to a set $C$
$\text{dom } f$	p. 5	Domain of a function $f$
$\overline{\text{dom } f}$	p. 6	Closure of the domain of a function $f$
$\text{gra } f$	p. 5	Epigraph of a function $f$
$\text{epi } f$	p. 5	Epigraph of a function $f$
$\overline{\text{epi } f}$	p. 6	Closure of the epigraph of a function $f$
$\text{lev}_{\leq \xi} f$	p. 6	Lower level set of a function $f$
$\text{lev}_{< \xi} f$	p. 6	Strict lower level set of a function $f$
$\text{cont } f$	p. 11	Domain of continuity of a function $f$
$\bar{f}$	p. 14	Lower semicontinuous envelope of a function $f$
$\check{f}$	p. 130	Lower semicontinuous convex envelope of a function $f$
$\text{Argmin } f$	p. 156	Set of global minimizers of a function $f$
$\text{Argmin}_C f$	p. 156	$\text{Argmin}(f + \iota_C)$
$f \square g$	p. 167	Infimal convolution of the functions $f$ and $g$
$f \boxdot g$	p. 167	Exact infimal convolution of the functions $f$ and $g$
$L \triangleright f$	p. 178	Infimal postcomposition of an operator $L$ and a function $f$
$L \boxtriangleright f$	p. 178	Exact infimal postcomposition of an operator $L$ and a function $f$
$\gamma f$	p. 173	Moreau envelope of index $\gamma$ of a function $f$
$\text{Prox}_f$	p. 175	Proximity operator of a function $f$
$f^*$	p. 181	Conjugate of a function $f$
$f^{**}$	p. 181	Biconjugate of a function $f$
$\text{rec } f$	p. 152	Recession function of a function $f$
$\text{pav}(f, g)$	p. 199	Proximal average of the functions $f$ and $g$
$\partial f$	p. 223	Subdifferential of a function $f$
$\nabla f$	p. 38	Gradient operator of a function $f$
$\nabla^2 f$	p. 38	Hessian operator of a function $f$
$Df$	p. 37	Gâteaux derivative of a function $f$
$D^2 f$	p. 38	Second Gâteaux derivative of a function $f$

$1_C$		Characteristic function of a set $C$
$\iota_C$	p. 12	Indicator function of a set $C$
$d_C$	p. 16	Distance function to a set $C$
$\text{diam}(C)$	p. 16	Diameter of a set $C$
$\sigma_C$	p. 109	Support function of a set $C$
$m_C$	p. 120	Minkowski gauge of a set $C$
$F^\top$	p. 190	Transposition of the bivariate function $F$

### Set-valued operators:

$A: \mathcal{X} \rightarrow 2^{\mathcal{Y}}$	p. 2	$A$ is a set-valued operator from $\mathcal{X}$ to $\mathcal{Y}$
$\text{gra } A$	p. 2	Graph of an operator $A$
$\text{dom } A$	p. 2	Domain of an operator $A$
$\overline{\text{dom } A}$	p. 2	Closure of the domain of an operator $A$
$\text{ran } A$	p. 2	Range of an operator $A$
$\overline{\text{ran } A}$	p. 2	Closure of the range of an operator $A$
$\text{zer } A$	p. 2	Set of zeros of an operator $A$
$A^{-1}$	p. 2	Inverse of an operator $A$
$\lambda A$	p. 3	Scaling of the operator $A$ by $\lambda \in \mathbb{R}$
$A + B$	p. 3	Sum of the operators $A$ and $B$
$A \circ B$	p. 2	Composition of the operators $A$ and $B$
$\tau_y A$	p. 3	Translation of an operator $A$ by $y$
$A^\vee$	p. 3	Reversal of an operator $A$
$F_A$	p. 304	Fitzpatrick function of an operator $A$
$J_A$	p. 333	Resolvent of an operator $A$
$R_A$	p. 336	Reflected resolvent of an operator $A$
${}^\gamma A$	p. 333	Yosida approximation of an operator $A$ of index $\gamma \in \mathbb{R}_{++}$
${}^0Ax$	p. 346	The element of minimal norm in $Ax$

### Single-valued operators:

$T: \mathcal{X} \rightarrow \mathcal{Y}$	p. 2	$T$ is an operator from $\mathcal{X}$ to $\mathcal{Y}$ defined everywhere on $\mathcal{X}$
$T^\vee$	p. 3	Reversal of an operator $T$
$\text{Fix } T$	p. 20	Set of fixed points of an operator $T: \mathcal{X} \rightarrow \mathcal{Y}$
$DT$	p. 37	Gâteaux derivative of an operator $T$
$D^2T$	p. 38	Second Gâteaux derivative of an operator $T$
$T _C$		Restriction of an operator $T$ to a set $C$
$T^{-1}(C)$	p. 2	Inverse image of a set $C$ by an operator $T$
$\ker L$	p. 31	Kernel of a linear operator $L$
$\ L\ $	p. 31	Norm of a linear operator $L$
$L^*$	p. 31	Adjoint of a bounded linear operator $L$
$L^\dagger$	p. 50	Generalized inverse of a bounded linear operator $L$
$A^\top$		Transpose of a matrix $A$



# Banach spaces:

$\mathcal{H}, \mathcal{H}_i, \mathcal{K}$	p. 27	Real Hilbert spaces
$\langle \cdot   \cdot \rangle$	p. 27	Scalar product
$\  \cdot \ $	p. 27	Norm
$d$	p. 27	Distance
$\rightarrow$	p. 33	Strong convergence in a Hilbert space
$\rightharpoonup$	p. 33	Weak convergence in a Hilbert space
$\text{Id}$	p. 27	Identity operator
$\mathcal{B}(\mathcal{H}, \mathcal{K})$	p. 31	Space of bounded linear operators from $\mathcal{H}$ to $\mathcal{K}$ with domain $\mathcal{H}$
$\mathcal{B}(\mathcal{H})$	p. 31	Space of bounded linear operators from $\mathcal{H}$ to $\mathcal{H}$ with domain $\mathcal{H}$
$\bigoplus_{i \in I} \mathcal{H}_i$	p. 28	Hilbert direct sum
$\mathcal{H}^{\text{weak}}$	p. 33	A real Hilbert space endowed with the weak topology
$\mathbb{R}^N$	p. 28	The standard $N$ -dimensional Euclidean space
$\mathbb{R}_+^N$	p. 5	The positive orthant in $\mathbb{R}^N$
$\mathbb{R}_-^N$	p. 5	The negative orthant in $\mathbb{R}^N$
$(\Omega, \mathcal{F}, \mu)$	p. 28	Measure space
$(\Omega, \mathcal{F}, \mathbb{P})$	p. 29	Probability space
$EX$	p. 29	Expected value of a random variable $X$
$x'$	p. 29	Time derivative of a function $x: [0, T] \rightarrow \mathbb{H}$
$L^p((\Omega, \mathcal{F}, \mu); \mathbb{H})$	p. 28	Measurable functions $x: \Omega \rightarrow \mathbb{H}$ such that $\ x\ _{\mathbb{H}}^p$ is $\mu$ -integrable
$L^2([0, T])$	p. 29	Measurable functions $x: [0, T] \rightarrow \mathbb{R}$ such that $ x ^2$ is Lebesgue integrable
$L^2([0, T]; \mathbb{H})$	p. 29	Measurable functions $x: [0, T] \rightarrow \mathbb{H}$ such that $\ x\ _{\mathbb{H}}^2$ is Lebesgue integrable
$W^{1,2}([0, T]; \mathbb{H})$	p. 29	Functions $x \in L^2([0, T]; \mathbb{H})$ such that $x' \in L^2([0, T]; \mathbb{H})$
$L^2(\Omega)$	p. 29	Measurable functions $x: \Omega \rightarrow \mathbb{R}$ such that $ x ^p$ is Lebesgue integrable
$\ell^p(\mathbb{N})$		Space of sequences $(\xi_n)_{n \in \mathbb{N}}$ in $\mathbb{R}$ such that $\sum_{n \in \mathbb{N}}  \xi_n ^p < +\infty$ .
$\ell^2(I)$	p. 28	Hilbert space of square-summable functions from $I$ to $\mathbb{R}$
$\ell_+^2(I)$	p. 89	Set of square-summable functions from $I$ to $\mathbb{R}_+$
$\ell_-^2(I)$	p. 89	Set of square-summable functions from $I$ to $\mathbb{R}_-$
$\mathbb{S}^N$	p. 28	Space of real $N \times N$ symmetric matrices
$\mathbb{S}_+^N$	p. 426	Set of real $N \times N$ symmetric positive semidefinite matrices



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