

Convex Optimization Theory

Chapter 1

Exercises and Solutions: Extended Version

Dimitri P. Bertsekas

Massachusetts Institute of Technology

Athena Scientific, Belmont, Massachusetts

<http://www.athenasc.com>

CHAPTER 1: EXERCISES AND SOLUTIONS†

SECTION 1.1: Convex Sets and Functions

1.1

Let C be a nonempty subset of \mathfrak{R}^n , and let λ_1 and λ_2 be positive scalars. Show that if C is convex, then $(\lambda_1 + \lambda_2)C = \lambda_1 C + \lambda_2 C$ [cf. Prop. 1.1.1(c)]. Show by example that this need not be true when C is not convex.

Solution: We always have $(\lambda_1 + \lambda_2)C \subset \lambda_1 C + \lambda_2 C$, even if C is not convex. To show the reverse inclusion assuming C is convex, note that a vector x in $\lambda_1 C + \lambda_2 C$ is of the form $x = \lambda_1 x_1 + \lambda_2 x_2$, where $x_1, x_2 \in C$. By convexity of C , we have

$$\frac{\lambda_1}{\lambda_1 + \lambda_2} x_1 + \frac{\lambda_2}{\lambda_1 + \lambda_2} x_2 \in C,$$

and it follows that

$$x = \lambda_1 x_1 + \lambda_2 x_2 \in (\lambda_1 + \lambda_2)C,$$

so $\lambda_1 C + \lambda_2 C \subset (\lambda_1 + \lambda_2)C$. For a counterexample when C is not convex, let C be a set in \mathfrak{R}^n consisting of two vectors, 0 and $x \neq 0$, and let $\lambda_1 = \lambda_2 = 1$. Then C is not convex, and $(\lambda_1 + \lambda_2)C = 2C = \{0, 2x\}$, while $\lambda_1 C + \lambda_2 C = C + C = \{0, x, 2x\}$, showing that $(\lambda_1 + \lambda_2)C \neq \lambda_1 C + \lambda_2 C$.

1.2 (Properties of Cones)

Show that:

- (a) The intersection $\cap_{i \in I} C_i$ of a collection $\{C_i \mid i \in I\}$ of cones is a cone.
- (b) The Cartesian product $C_1 \times C_2$ of two cones C_1 and C_2 is a cone.

† This set of exercises will be periodically updated as new exercises are added. Many of the exercises and solutions given here were developed as part of my earlier convex optimization book [BNO03] (coauthored with Angelia Nedić and Asuman Ozdaglar), and are posted on the internet of that book's web site. The contribution of my coauthors in the development of these exercises and their solutions is gratefully acknowledged. Since some of the exercises and/or their solutions have been modified and also new exercises have been added, all errors are my sole responsibility.

- (c) The vector sum $C_1 + C_2$ of two cones C_1 and C_2 is a cone.
- (d) The image and the inverse image of a cone under a linear transformation is a cone.
- (e) A subset C is a convex cone if and only if it is closed under addition and positive scalar multiplication, i.e., $C + C \subset C$, and $\gamma C \subset C$ for all $\gamma > 0$.

Solution: (a) Let $x \in \bigcap_{i \in I} C_i$ and let α be a positive scalar. Since $x \in C_i$ for all $i \in I$ and each C_i is a cone, the vector αx belongs to C_i for all $i \in I$. Hence, $\alpha x \in \bigcap_{i \in I} C_i$, showing that $\bigcap_{i \in I} C_i$ is a cone.

(b) Let $x \in C_1 \times C_2$ and let α be a positive scalar. Then $x = (x_1, x_2)$ for some $x_1 \in C_1$ and $x_2 \in C_2$, and since C_1 and C_2 are cones, it follows that $\alpha x_1 \in C_1$ and $\alpha x_2 \in C_2$. Hence, $\alpha x = (\alpha x_1, \alpha x_2) \in C_1 \times C_2$, showing that $C_1 \times C_2$ is a cone.

(c) Let $x \in C_1 + C_2$ and let α be a positive scalar. Then, $x = x_1 + x_2$ for some $x_1 \in C_1$ and $x_2 \in C_2$, and since C_1 and C_2 are cones, $\alpha x_1 \in C_1$ and $\alpha x_2 \in C_2$. Hence, $\alpha x = \alpha x_1 + \alpha x_2 \in C_1 + C_2$, showing that $C_1 + C_2$ is a cone.

(d) First we prove that $A \cdot C$ is a cone, where A is a linear transformation and $A \cdot C$ is the image of C under A . Let $z \in A \cdot C$ and let α be a positive scalar. Then, $Ax = z$ for some $x \in C$, and since C is a cone, $\alpha x \in C$. Because $A(\alpha x) = \alpha z$, the vector αz is in $A \cdot C$, showing that $A \cdot C$ is a cone. Next we prove that the inverse image $A^{-1} \cdot C$ of C under A is a cone. Let $x \in A^{-1} \cdot C$ and let α be a positive scalar. Then $Ax \in C$, and since C is a cone, $\alpha Ax \in C$. Thus, the vector $A(\alpha x) \in C$, implying that $\alpha x \in A^{-1} \cdot C$, and showing that $A^{-1} \cdot C$ is a cone.

(e) Let C be a convex cone. Then $\gamma C \subset C$, for all $\gamma > 0$, by the definition of cone. Furthermore, by convexity of C , for all $x, y \in C$, we have $z \in C$, where

$$z = \frac{1}{2}(x + y).$$

Hence $(x + y) = 2z \in C$, since C is a cone, and it follows that $C + C \subset C$. Conversely, assume that $C + C \subset C$, and $\gamma C \subset C$. Then C is a cone. Furthermore, if $x, y \in C$ and $\alpha \in (0, 1)$, we have $\alpha x \in C$ and $(1 - \alpha)y \in C$, and $\alpha x + (1 - \alpha)y \in C$ (since $C + C \subset C$). Hence C is convex.

1.3 (Convexity under Composition)

Let C be a nonempty convex subset of \mathfrak{R}^n . Let also $f = (f_1, \dots, f_m)$, where $f_i : C \mapsto \mathfrak{R}$, $i = 1, \dots, m$, are convex functions, and let $g : \mathfrak{R}^m \mapsto \mathfrak{R}$ be a function that is convex and monotonically nondecreasing over a convex set that contains the set $\{f(x) \mid x \in C\}$, in the sense that for all u, \bar{u} in this set such that $u \leq \bar{u}$, we have $g(u) \leq g(\bar{u})$. Show that the function h defined by $h(x) = g(f(x))$ is convex over C . If in addition, $m = 1$, g is monotonically increasing and f is strictly convex, then h is strictly convex.

Solution: Let $x, y \in \mathfrak{R}^n$ and let $\alpha \in [0, 1]$. By the definitions of h and f , we have

$$\begin{aligned}
 h(\alpha x + (1 - \alpha)y) &= g\left(f(\alpha x + (1 - \alpha)y)\right) \\
 &= g\left(f_1(\alpha x + (1 - \alpha)y), \dots, f_m(\alpha x + (1 - \alpha)y)\right) \\
 &\leq g\left(\alpha f_1(x) + (1 - \alpha)f_1(y), \dots, \alpha f_m(x) + (1 - \alpha)f_m(y)\right) \\
 &= g\left(\alpha(f_1(x), \dots, f_m(x)) + (1 - \alpha)(f_1(y), \dots, f_m(y))\right) \\
 &\leq \alpha g(f_1(x), \dots, f_m(x)) + (1 - \alpha)g(f_1(y), \dots, f_m(y)) \\
 &= \alpha g(f(x)) + (1 - \alpha)g(f(y)) \\
 &= \alpha h(x) + (1 - \alpha)h(y),
 \end{aligned}$$

where the first inequality follows by convexity of each f_i and monotonicity of g , while the second inequality follows by convexity of g .

If $m = 1$, g is monotonically increasing, and f is strictly convex, then the first inequality is strict whenever $x \neq y$ and $\alpha \in (0, 1)$, showing that h is strictly convex.

1.4 (Examples of Convex Functions)

Show that the following functions from \mathfrak{R}^n to $(-\infty, \infty]$ are convex:

(a)

$$f_1(x_1, \dots, x_n) = \begin{cases} -(x_1 x_2 \cdots x_n)^{\frac{1}{n}} & \text{if } x_1 > 0, \dots, x_n > 0, \\ \infty & \text{otherwise.} \end{cases}$$

(b) $f_2(x) = \ln(e^{x_1} + \cdots + e^{x_n})$.

(c) $f_3(x) = \|x\|^p$ with $p \geq 1$.

(d) $f_4(x) = \frac{1}{f(x)}$, where f is concave and $0 < f(x) < \infty$ for all x .

(e) $f_5(x) = \alpha f(x) + \beta$, where $f : \mathfrak{R}^n \mapsto \mathfrak{R}$ is a convex function, and α and β are scalars, with $\alpha \geq 0$.

(f) $f_6(x) = e^{\beta x' A x}$, where A is a positive semidefinite symmetric $n \times n$ matrix and β is a positive scalar.

(g) $f_7(x) = f(Ax + b)$, where $f : \mathfrak{R}^m \mapsto \mathfrak{R}$ is a convex function, A is an $m \times n$ matrix, and b is a vector in \mathfrak{R}^m .

Solution: (a) Denote $X = \text{dom}(f_1)$. It can be seen that f_1 is twice continuously differentiable over X and its Hessian matrix is given by

$$\nabla^2 f_1(x) = \frac{f_1(x)}{n^2} \begin{bmatrix} \frac{1-n}{x_1^2} & \frac{1}{x_1 x_2} & \cdots & \frac{1}{x_1 x_n} \\ \frac{1}{x_2 x_1} & \frac{1-n}{x_2^2} & \cdots & \frac{1}{x_2 x_n} \\ & & \ddots & \\ \frac{1}{x_n x_1} & \frac{1}{x_1 x_2} & \cdots & \frac{1-n}{x_n^2} \end{bmatrix}$$

for all $x = (x_1, \dots, x_n) \in X$. From this, direct computation shows that for all $z = (z_1, \dots, z_n) \in \mathfrak{R}^n$ and $x = (x_1, \dots, x_n) \in X$, we have

$$z' \nabla^2 f_1(x) z = \frac{f_1(x)}{n^2} \left(\left(\sum_{i=1}^n \frac{z_i}{x_i} \right)^2 - n \sum_{i=1}^n \left(\frac{z_i}{x_i} \right)^2 \right).$$

Note that this quadratic form is nonnegative for all $z \in \mathfrak{R}^n$ and $x \in X$, since $f_1(x) < 0$, and for any real numbers $\alpha_1, \dots, \alpha_n$, we have

$$(\alpha_1 + \dots + \alpha_n)^2 \leq n(\alpha_1^2 + \dots + \alpha_n^2),$$

in view of the fact that $2\alpha_j\alpha_k \leq \alpha_j^2 + \alpha_k^2$. Hence, $\nabla^2 f_1(x)$ is positive semidefinite for all $x \in X$, and it follows from Prop. 1.1.10(a) that f_1 is convex.

(b) We show that the Hessian of f_2 is positive semidefinite at all $x \in \mathfrak{R}^n$. Let $\beta(x) = e^{x_1} + \dots + e^{x_n}$. Then a straightforward calculation yields

$$z' \nabla^2 f_2(x) z = \frac{1}{\beta(x)^2} \sum_{i=1}^n \sum_{j=1}^n e^{(x_i+x_j)} (z_i - z_j)^2 \geq 0, \quad \forall z \in \mathfrak{R}^n.$$

Hence by Prop. 1.1.10(a), f_2 is convex.

(c) The function $f_3(x) = \|x\|^p$ can be viewed as a composition $g(f(x))$ of the scalar function $g(t) = t^p$ with $p \geq 1$ and the function $f(x) = \|x\|$. In this case, g is convex and monotonically increasing over the nonnegative axis, the set of values that f can take, while f is convex over \mathfrak{R}^n (since any vector norm is convex). Using Exercise 1.3, it follows that the function $f_3(x) = \|x\|^p$ is convex over \mathfrak{R}^n .

(d) The function $f_4(x) = \frac{1}{f(x)}$ can be viewed as a composition $g(h(x))$ of the function $g(t) = -\frac{1}{t}$ for $t < 0$ and the function $h(x) = -f(x)$ for $x \in \mathfrak{R}^n$. In this case, the g is convex and monotonically increasing in the set $\{t \mid t < 0\}$, while h is convex over \mathfrak{R}^n . Using Exercise 1.3, it follows that the function $f_4(x) = \frac{1}{f(x)}$ is convex over \mathfrak{R}^n .

(e) The function $f_5(x) = \alpha f(x) + \beta$ can be viewed as a composition $g(f(x))$ of the function $g(t) = \alpha t + \beta$, where $t \in \mathfrak{R}$, and the function $f(x)$ for $x \in \mathfrak{R}^n$. In this case, g is convex and monotonically increasing over \mathfrak{R} (since $\alpha \geq 0$), while f is convex over \mathfrak{R}^n . Using Exercise 1.3, it follows that f_5 is convex over \mathfrak{R}^n .

(f) The function $f_6(x) = e^{\beta x'Ax}$ can be viewed as a composition $g(f(x))$ of the function $g(t) = e^{\beta t}$ for $t \in \mathfrak{R}$ and the function $f(x) = x'Ax$ for $x \in \mathfrak{R}^n$. In this case, g is convex and monotonically increasing over \mathfrak{R} , while f is convex over \mathfrak{R}^n (since A is positive semidefinite). Using Exercise 1.3, it follows that f_6 is convex over \mathfrak{R}^n .

(g) This part is straightforward using the definition of a convex function.

1.5 (Ascent/Descent Behavior of a Convex Function)

Let $f : \mathfrak{R} \mapsto \mathfrak{R}$ be a convex function.

- (a) (*Monotropic Property*) Use the definition of convexity to show that f is “turning upwards” in the sense that if x_1, x_2, x_3 are three scalars such that $x_1 < x_2 < x_3$, then

$$\frac{f(x_2) - f(x_1)}{x_2 - x_1} \leq \frac{f(x_3) - f(x_2)}{x_3 - x_2}.$$

- (b) Use part (a) to show that there are four possibilities as x increases to ∞ : (1) $f(x)$ decreases monotonically to $-\infty$, (2) $f(x)$ decreases monotonically to a finite value, (3) $f(x)$ reaches some value and stays at that value, (4) $f(x)$ increases monotonically to ∞ when $x \geq \bar{x}$ for some $\bar{x} \in \mathfrak{R}$.

Solution: (a) Let x_1, x_2, x_3 be three scalars such that $x_1 < x_2 < x_3$. Then we can write x_2 as a convex combination of x_1 and x_3 as follows

$$x_2 = \frac{x_3 - x_2}{x_3 - x_1}x_1 + \frac{x_2 - x_1}{x_3 - x_1}x_3,$$

so that by convexity of f , we obtain

$$f(x_2) \leq \frac{x_3 - x_2}{x_3 - x_1}f(x_1) + \frac{x_2 - x_1}{x_3 - x_1}f(x_3).$$

This relation and the fact

$$f(x_2) = \frac{x_3 - x_2}{x_3 - x_1}f(x_2) + \frac{x_2 - x_1}{x_3 - x_1}f(x_2),$$

imply that

$$\frac{x_3 - x_2}{x_3 - x_1}(f(x_2) - f(x_1)) \leq \frac{x_2 - x_1}{x_3 - x_1}(f(x_3) - f(x_2)).$$

By multiplying the preceding relation with $x_3 - x_1$ and by dividing it with $(x_3 - x_2)(x_2 - x_1)$, we obtain

$$\frac{f(x_2) - f(x_1)}{x_2 - x_1} \leq \frac{f(x_3) - f(x_2)}{x_3 - x_2}.$$

- (b) Let $\{x_k\}$ be an increasing scalar sequence, i.e., $x_1 < x_2 < x_3 < \dots$. Then according to part (a), we have for all k

$$\frac{f(x_2) - f(x_1)}{x_2 - x_1} \leq \frac{f(x_3) - f(x_2)}{x_3 - x_2} \leq \dots \leq \frac{f(x_{k+1}) - f(x_k)}{x_{k+1} - x_k}. \quad (1.1)$$

Since $(f(x_k) - f(x_{k-1})) / (x_k - x_{k-1})$ is monotonically nondecreasing, we have

$$\frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}} \rightarrow \gamma, \quad (1.2)$$

where γ is either a real number or ∞ . Furthermore,

$$\frac{f(x_{k+1}) - f(x_k)}{x_{k+1} - x_k} \leq \gamma, \quad \forall k. \quad (1.3)$$

We now show that γ is independent of the sequence $\{x_k\}$. Let $\{y_j\}$ be any increasing scalar sequence. For each j , choose x_{k_j} such that $y_j < x_{k_j}$ and $x_{k_1} < x_{k_2} < \dots < x_{k_j}$, so that we have $y_j < y_{j+1} < x_{k_{j+1}} < x_{k_{j+2}}$. By part (a), it follows that

$$\frac{f(y_{j+1}) - f(y_j)}{y_{j+1} - y_j} \leq \frac{f(x_{k_{j+2}}) - f(x_{k_{j+1}})}{x_{k_{j+2}} - x_{k_{j+1}}},$$

and letting $j \rightarrow \infty$ yields

$$\lim_{j \rightarrow \infty} \frac{f(y_{j+1}) - f(y_j)}{y_{j+1} - y_j} \leq \gamma.$$

Similarly, by exchanging the roles of $\{x_k\}$ and $\{y_j\}$, we can show that

$$\lim_{j \rightarrow \infty} \frac{f(y_{j+1}) - f(y_j)}{y_{j+1} - y_j} \geq \gamma.$$

Thus the limit in Eq. (1.2) is independent of the choice for $\{x_k\}$, and Eqs. (1.1) and (1.3) hold for any increasing scalar sequence $\{x_k\}$.

We consider separately each of the three possibilities $\gamma < 0, \gamma = 0$, and $\gamma > 0$. First, suppose that $\gamma < 0$, and let $\{x_k\}$ be any increasing sequence. By using Eq. (1.3), we obtain

$$\begin{aligned} f(x_k) &= \sum_{j=1}^{k-1} \frac{f(x_{j+1}) - f(x_j)}{x_{j+1} - x_j} (x_{j+1} - x_j) + f(x_1) \\ &\leq \sum_{j=1}^{k-1} \gamma (x_{j+1} - x_j) + f(x_1) \\ &= \gamma(x_k - x_1) + f(x_1), \end{aligned}$$

and since $\gamma < 0$ and $x_k \rightarrow \infty$, it follows that $f(x_k) \rightarrow -\infty$. To show that f decreases monotonically, pick any x and y with $x < y$, and consider the sequence $x_1 = x$, $x_2 = y$, and $x_k = y + k$ for all $k \geq 3$. By using Eq. (1.3) with $k = 1$, we have

$$\frac{f(y) - f(x)}{y - x} \leq \gamma < 0,$$

so that $f(y) - f(x) < 0$. Hence f decreases monotonically to $-\infty$, corresponding to case (1).

Suppose now that $\gamma = 0$, and let $\{x_k\}$ be any increasing sequence. Then, by Eq. (1.3), we have $f(x_{k+1}) - f(x_k) \leq 0$ for all k . If $f(x_{k+1}) - f(x_k) < 0$ for all k , then f decreases monotonically. To show this, pick any x and y with $x < y$, and consider a new sequence given by $y_1 = x$, $y_2 = y$, and $y_k = x_{K+k-3}$ for all

$k \geq 3$, where K is large enough so that $y < x_K$. By using Eqs. (1.1) and (1.3) with $\{y_k\}$, we have

$$\frac{f(y) - f(x)}{y - x} \leq \frac{f(x_{K+1}) - f(x_K)}{x_{K+1} - x_K} < 0,$$

implying that $f(y) - f(x) < 0$. Hence f decreases monotonically, and it may decrease to $-\infty$ or to a finite value, corresponding to cases (1) or (2), respectively.

If for some K we have $f(x_{K+1}) - f(x_K) = 0$, then by Eqs. (1.1) and (1.3) where $\gamma = 0$, we obtain $f(x_k) = f(x_K)$ for all $k \geq K$. To show that f stays at the value $f(x_K)$ for all $x \geq x_K$, choose any x such that $x > x_K$, and define $\{y_k\}$ as $y_1 = x_K$, $y_2 = x$, and $y_k = x_{N+k-3}$ for all $k \geq 3$, where N is large enough so that $x < x_N$. By using Eqs. (1.1) and (1.3) with $\{y_k\}$, we have

$$\frac{f(x) - f(x_K)}{x - x_K} \leq \frac{f(x_N) - f(x)}{x_N - x} \leq 0,$$

so that $f(x) \leq f(x_K)$ and $f(x_N) \leq f(x)$. Since $f(x_K) = f(x_N)$, we have $f(x) = f(x_K)$. Hence $f(x) = f(x_K)$ for all $x \geq x_K$, corresponding to case (3).

Finally, suppose that $\gamma > 0$, and let $\{x_k\}$ be any increasing sequence. Since $(f(x_k) - f(x_{k-1})) / (x_k - x_{k-1})$ is nondecreasing and tends to γ [cf. Eqs. (1.2) and (1.3)], there is a positive integer K and a positive scalar ϵ with $\epsilon < \gamma$ such that

$$\epsilon \leq \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}}, \quad \forall k \geq K. \quad (1.4)$$

Therefore, for all $k > K$

$$f(x_k) = \sum_{j=K}^{k-1} \frac{f(x_{j+1}) - f(x_j)}{x_{j+1} - x_j} (x_{j+1} - x_j) + f(x_K) \geq \epsilon(x_k - x_K) + f(x_K),$$

implying that $f(x_k) \rightarrow \infty$. To show that $f(x)$ increases monotonically to ∞ for all $x \geq x_K$, pick any $x < y$ satisfying $x_K < x < y$, and consider a sequence given by $y_1 = x_K$, $y_2 = x$, $y_3 = y$, and $y_k = x_{N+k-4}$ for $k \geq 4$, where N is large enough so that $y < x_N$. By using Eq. (1.4) with $\{y_k\}$, we have

$$\epsilon \leq \frac{f(y) - f(x)}{y - x}.$$

Thus $f(x)$ increases monotonically to ∞ for all $x \geq x_K$, corresponding to case (4) with $\bar{x} = x_K$.

1.6 (Posynomials)

A *posynomial* is a function of positive scalar variables y_1, \dots, y_n of the form

$$g(y_1, \dots, y_n) = \sum_{i=1}^m \beta_i y_1^{a_{i1}} \dots y_n^{a_{in}},$$

where a_{ij} and β_i are scalars, such that $\beta_i > 0$ for all i . Show the following:

- (a) A posynomial need not be convex.
- (b) By a logarithmic change of variables, where we set

$$f(x) = \ln(g(y_1, \dots, y_n)), \quad b_i = \ln \beta_i, \quad \forall i, \quad x_j = \ln y_j, \quad \forall j,$$

we obtain a convex function

$$f(x) = \ln \exp(Ax + b), \quad \forall x \in \mathfrak{R}^n,$$

where $\exp(z) = e^{z_1} + \dots + e^{z_m}$ for all $z \in \mathfrak{R}^m$, A is an $m \times n$ matrix with components a_{ij} , and $b \in \mathfrak{R}^m$ is a vector with components b_i .

- (c) Every function $g : \mathfrak{R}^n \mapsto \mathfrak{R}$ of the form

$$g(y) = g_1(y)^{\gamma_1} \cdots g_r(y)^{\gamma_r},$$

where g_k is a posynomial and $\gamma_k > 0$ for all k , can be transformed by a logarithmic change of variables into a convex function f given by

$$f(x) = \sum_{k=1}^r \gamma_k \ln \exp(A_k x + b_k),$$

with the matrix A_k and the vector b_k being associated with the posynomial g_k for each k .

Solution: (a) Consider the following posynomial for which we have $n = m = 1$ and $\beta = \frac{1}{2}$,

$$g(y) = y^{\frac{1}{2}}, \quad \forall y > 0.$$

This function is not convex.

- (b) Consider the following change of variables, where we set

$$f(x) = \ln(g(y_1, \dots, y_n)), \quad b_i = \ln \beta_i, \quad \forall i, \quad x_j = \ln y_j, \quad \forall j.$$

With this change of variables, $f(x)$ can be written as

$$f(x) = \ln \left(\sum_{i=1}^m e^{b_i + a_{i1}x_1 + \dots + a_{in}x_n} \right).$$

Note that $f(x)$ can also be represented as

$$f(x) = \ln \exp(Ax + b), \quad \forall x \in \mathfrak{R}^n,$$

where $\ln \exp(z) = \ln(e^{z_1} + \dots + e^{z_m})$ for all $z \in \mathfrak{R}^m$, A is an $m \times n$ matrix with entries a_{ij} , and $b \in \mathfrak{R}^m$ is a vector with components b_i . Let $f_2(z) = \ln(e^{z_1} + \dots + e^{z_m})$. This function is convex by Exercise 1.4(b). With this identification,

$f(x)$ can be viewed as the composition $f(x) = f_2(Ax + b)$, which is convex by Exercise 1.4(g).

(c) Consider a function $g : \Re^n \mapsto \Re$ of the form

$$g(y) = g_1(y)^{\gamma_1} \cdots g_r(y)^{\gamma_r},$$

where g_k is a posynomial and $\gamma_k > 0$ for all k . Using a change of variables similar to part (b), we see that we can represent the function $f(x) = \ln g(y)$ as

$$f(x) = \sum_{k=1}^r \gamma_k \ln \exp(A_k x + b_k),$$

with the matrix A_k and the vector b_k being associated with the posynomial g_k for each k . Since $f(x)$ is the weighted sum of convex functions with nonnegative coefficients [part (b)], it follows that $f(x)$ is convex.

1.7 (Arithmetic-Geometric Mean Inequality)

Show that if $\alpha_1, \dots, \alpha_n$ are positive scalars with $\sum_{i=1}^n \alpha_i = 1$, then for every set of positive scalars x_1, \dots, x_n , we have

$$x_1^{\alpha_1} x_2^{\alpha_2} \cdots x_n^{\alpha_n} \leq \alpha_1 x_1 + \alpha_2 x_2 + \cdots + \alpha_n x_n,$$

with equality if and only if $x_1 = x_2 = \cdots = x_n$. *Hint:* Show that $(-\ln x)$ is a strictly convex function on $(0, \infty)$.

Solution: Consider the function $f(x) = -\ln(x)$. Since $\nabla^2 f(x) = 1/x^2 > 0$ for all $x > 0$, the function $-\ln(x)$ is strictly convex over $(0, \infty)$. Therefore, for all positive scalars x_1, \dots, x_n and $\alpha_1, \dots, \alpha_n$ with $\sum_{i=1}^n \alpha_i = 1$, we have

$$-\ln(\alpha_1 x_1 + \cdots + \alpha_n x_n) \leq -\alpha_1 \ln(x_1) - \cdots - \alpha_n \ln(x_n),$$

which is equivalent to

$$e^{\ln(\alpha_1 x_1 + \cdots + \alpha_n x_n)} \geq e^{\alpha_1 \ln(x_1) + \cdots + \alpha_n \ln(x_n)} = e^{\alpha_1 \ln(x_1)} \cdots e^{\alpha_n \ln(x_n)},$$

or

$$\alpha_1 x_1 + \cdots + \alpha_n x_n \geq x_1^{\alpha_1} \cdots x_n^{\alpha_n},$$

as desired. Since $-\ln(x)$ is strictly convex, the above inequality is satisfied with equality if and only if the scalars x_1, \dots, x_n are all equal.

1.8 (Young and Holder Inequalities)

Use the result of Exercise 1.7 to verify Young's inequality

$$xy \leq \frac{x^p}{p} + \frac{y^q}{q}, \quad \forall x \geq 0, \forall y \geq 0,$$

where $p > 0$, $q > 0$, and

$$1/p + 1/q = 1.$$

Then, use Young's inequality to verify Holder's inequality

$$\sum_{i=1}^n |x_i y_i| \leq \left(\sum_{i=1}^n |x_i|^p \right)^{1/p} \left(\sum_{i=1}^n |y_i|^q \right)^{1/q}.$$

Solution: According to Exercise 1.7, we have

$$u^{1/p} v^{1/q} \leq \frac{u}{p} + \frac{v}{q}, \quad \forall u > 0, \forall v > 0,$$

where $1/p + 1/q = 1$, $p > 0$, and $q > 0$. The above relation also holds if $u = 0$ or $v = 0$. By setting $u = x^p$ and $v = y^q$, we obtain Young's inequality

$$xy \leq \frac{x^p}{p} + \frac{y^q}{q}, \quad \forall x \geq 0, \forall y \geq 0.$$

To show Holder's inequality, note that it holds if $x_1 = \cdots = x_n = 0$ or $y_1 = \cdots = y_n = 0$. If x_1, \dots, x_n and y_1, \dots, y_n are such that $(x_1, \dots, x_n) \neq 0$ and $(y_1, \dots, y_n) \neq 0$, then by using

$$x = \frac{|x_i|}{\left(\sum_{j=1}^n |x_j|^p \right)^{1/p}} \quad \text{and} \quad y = \frac{|y_i|}{\left(\sum_{j=1}^n |y_j|^q \right)^{1/q}}$$

in Young's inequality, we have for all $i = 1, \dots, n$,

$$\frac{|x_i|}{\left(\sum_{j=1}^n |x_j|^p \right)^{1/p}} \frac{|y_i|}{\left(\sum_{j=1}^n |y_j|^q \right)^{1/q}} \leq \frac{|x_i|^p}{p \left(\sum_{j=1}^n |x_j|^p \right)} + \frac{|y_i|^q}{q \left(\sum_{j=1}^n |y_j|^q \right)}.$$

By adding these inequalities over $i = 1, \dots, n$, we obtain

$$\frac{\sum_{i=1}^n |x_i| \cdot |y_i|}{\left(\sum_{j=1}^n |x_j|^p \right)^{1/p} \left(\sum_{j=1}^n |y_j|^q \right)^{1/q}} \leq \frac{1}{p} + \frac{1}{q} = 1,$$

which implies Holder's inequality.

1.9 (Characterization of Differentiable Convex Functions)

Let $f : \mathfrak{R}^n \mapsto \mathfrak{R}$ be a differentiable function. Show that f is convex over a nonempty convex set C if and only if

$$(\nabla f(x) - \nabla f(y))'(x - y) \geq 0, \quad \forall x, y \in C.$$

Note: The condition above says that the function f , restricted to the line segment connecting x and y , has monotonically nondecreasing gradient.

Solution: If f is convex, then by Prop. 1.1.7(a), we have

$$f(y) \geq f(x) + \nabla f(x)'(y - x), \quad \forall x, y \in C.$$

By exchanging the roles of x and y in this relation, we obtain

$$f(x) \geq f(y) + \nabla f(y)'(x - y), \quad \forall x, y \in C,$$

and by adding the preceding two inequalities, it follows that

$$(\nabla f(y) - \nabla f(x))'(x - y) \geq 0. \tag{1.5}$$

Conversely, let Eq. (1.5) hold, and let x and y be two points in C . Define the function $h : \mathfrak{R} \mapsto \mathfrak{R}$ by

$$h(t) = f(x + t(y - x)).$$

Consider some $t, t' \in [0, 1]$ such that $t < t'$. By convexity of C , we have that $x + t(y - x)$ and $x + t'(y - x)$ belong to C . Using the chain rule and Eq. (1.5), we have

$$\begin{aligned} & \left(\frac{dh(t')}{dt} - \frac{dh(t)}{dt} \right) (t' - t) \\ &= \left(\nabla f(x + t'(y - x)) - \nabla f(x + t(y - x)) \right)' (y - x) (t' - t) \\ &\geq 0. \end{aligned}$$

Thus, dh/dt is nondecreasing on $[0, 1]$ and for any $t \in (0, 1)$, we have

$$\frac{h(t) - h(0)}{t} = \frac{1}{t} \int_0^t \frac{dh(\tau)}{d\tau} d\tau \leq h(t) \leq \frac{1}{1-t} \int_t^1 \frac{dh(\tau)}{d\tau} d\tau = \frac{h(1) - h(t)}{1-t}.$$

Equivalently,

$$th(1) + (1-t)h(0) \geq h(t),$$

and from the definition of h , we obtain

$$tf(y) + (1-t)f(x) \geq f(ty + (1-t)x).$$

Since this inequality has been proved for arbitrary $t \in [0, 1]$ and $x, y \in C$, we conclude that f is convex.

1.10 (Strong Convexity)

Let $f : \mathbb{R}^n \mapsto \mathbb{R}$ be a function that is continuous over a closed convex set $C \subset \text{dom}(f)$, and let $\sigma > 0$. We say that f is *strongly convex over C with coefficient σ* if for all $x, y \in C$ and all $\alpha \in [0, 1]$, we have

$$f(\alpha x + (1 - \alpha)y) + \frac{\sigma}{2}\alpha(1 - \alpha)\|x - y\|^2 \leq \alpha f(x) + (1 - \alpha)f(y).$$

- (a) Show that if f is strongly convex over C with coefficient σ , then f is strictly convex over C . Furthermore, there exists a unique $x^* \in C$ that minimizes f over C , and we have

$$f(x) \geq f(x^*) + \frac{\sigma}{2}\|x - x^*\|^2, \quad \forall x \in C.$$

- (b) Assume that $\text{int}(C)$, the interior of C , is nonempty, and that f is continuously differentiable over $\text{int}(C)$. Show that the following are equivalent:

- (i) f is strongly convex with coefficient σ over C .
(ii) We have

$$(\nabla f(x) - \nabla f(y))'(x - y) \geq \sigma\|x - y\|^2, \quad \forall x, y \in \text{int}(C).$$

- (iii) We have

$$f(y) \geq f(x) + \nabla f(x)'(y - x) + \frac{\sigma}{2}\|x - y\|^2, \quad \forall x, y \in \text{int}(C).$$

Furthermore, if f is twice continuously differentiable over $\text{int}(C)$, the above three properties are equivalent to:

- (iv) The matrix $\nabla^2 f(x) - \sigma I$ is positive semidefinite for every $x \in \text{int}(C)$, where I is the identity matrix.

Solution: (a) The strict convexity of f over C is evident from the definition of strong convexity and the hypothesis. Strict convexity also implies that there can be at most one minimum of f over C .

To show existence of a vector x^* that minimizes f over C , we show that every level set $\{x \in C \mid f(x) \leq \gamma\}$ is bounded and hence compact (since C is closed and f is continuous over C), and then use Weierstrass' Theorem. Assume to arrive at a contradiction that a level set $L = \{x \in C \mid f(x) \leq \gamma\}$ is unbounded, and let $\{x_k\} \subset L$ be an unbounded sequence. We assume with no loss of generality that $\|x_k - x_0\| \geq 1$ for all k . Let $\alpha_k = 1/\|x_k - x_0\|$, and note that $\alpha_k \rightarrow 0$. Define

$$y_k = \alpha_k x_k + (1 - \alpha_k)x_0 = \frac{x_k - x_0}{\|x_k - x_0\|} + x_0,$$

and note that $\|y_k - x_0\| = 1$ for all $k \geq 1$. By strong convexity of f , we have

$$\begin{aligned} f(y_k) &\leq \alpha_k f(x_k) + (1 - \alpha_k) f(x_0) - \frac{1}{2} \sigma \alpha_k (1 - \alpha_k) \|x_k - x_0\|^2 \\ &= \alpha_k f(x_k) + (1 - \alpha_k) f(x_0) - \frac{1}{2} \sigma (1 - \alpha_k) \|x_k - x_0\| \\ &\leq \gamma - \frac{1}{2} \sigma (1 - \alpha_k) \|x_k - x_0\|. \end{aligned}$$

Hence $f(y_k) \rightarrow -\infty$, which contradicts the boundedness of $\{y_k\}$ and the continuity of f .

To show the inequality $f(x) \geq f(x^*) + (\sigma/2)\|x - x^*\|^2$, we write for any $x \in C$ and $\alpha \in (0, 1)$,

$$\begin{aligned} \alpha f(x) + (1 - \alpha) f(x^*) &\geq f(\alpha x + (1 - \alpha)x^*) + \frac{1}{2} \sigma \alpha (1 - \alpha) \|x - x^*\|^2 \\ &\geq f(x^*) + \frac{1}{2} \sigma \alpha (1 - \alpha) \|x - x^*\|^2. \end{aligned}$$

It follows that $f(x) \geq f(x^*) + (\sigma/2)(1 - \alpha)\|x - x^*\|^2$, and by taking the limit as $\alpha \rightarrow 0$, we obtain the desired inequality.

(b) We first show that (i) implies (ii). We have, using the definition of strong convexity,

$$f(y) + \alpha \nabla f(y)'(x - y) \leq f(y + \alpha(x - y)) \leq \alpha f(x) + (1 - \alpha) f(y) - \frac{\sigma}{2} \alpha (1 - \alpha) \|x - y\|^2,$$

for all $x, y \in \text{int}(C)$ and $\alpha \in (0, 1)$, from which

$$f(y) + \nabla f(y)'(x - y) \leq f(x) - \frac{\sigma}{2} (1 - \alpha) \|x - y\|^2.$$

Similarly,

$$f(x) + \nabla f(x)'(y - x) \leq f(y) - \frac{\sigma}{2} (1 - \alpha) \|x - y\|^2, \quad (1.6)$$

and adding these two inequalities:

$$(\nabla f(y) - \nabla f(x))'(x - y) \leq -\sigma(1 - \alpha) \|x - y\|^2,$$

or

$$(\nabla f(y) - \nabla f(x))'(y - x) \geq \sigma(1 - \alpha) \|x - y\|^2.$$

Taking the limit as $\alpha \rightarrow 0$, we obtain

$$(\nabla f(y) - \nabla f(x))'(y - x) \geq \sigma \|x - y\|^2.$$

Next we show that (ii) implies (i). For any $\alpha \in (0, 1)$ and $x_1, x_2 \in \text{int}(C)$ with $x_1 \neq x_2$, let

$$x_\alpha = \alpha x_1 + (1 - \alpha)x_2.$$

We have

$$f(x_\alpha) = f(x_1) + \int_0^1 \nabla f(x_1 + t(x_\alpha - x_1))'(x_\alpha - x_1) dt,$$

$$f(x_\alpha) = f(x_2) + \int_0^1 \nabla f(x_2 + t(x_\alpha - x_2))'(x_\alpha - x_2) dt.$$

Multiplying these relations with α and $1 - \alpha$, respectively, adding, and collecting terms using the relations $x_\alpha - x_1 = (1 - \alpha)(x_2 - x_1)$, $x_\alpha - x_2 = \alpha(x_1 - x_2)$, and

$$(x_1 + t(x_\alpha - x_1)) - (x_2 + t(x_\alpha - x_2)) = (1 - t)(x_1 - x_2),$$

we obtain

$$\begin{aligned} & \alpha f(x_1) + (1 - \alpha)f(x_2) - f(x_\alpha) \\ &= \alpha(1 - \alpha) \int_0^1 \left(\nabla f(x_1 + t(x_\alpha - x_1)) - \nabla f(x_2 + t(x_\alpha - x_2)) \right)'(x_1 - x_2) dt \\ &\geq \sigma \alpha(1 - \alpha) \|x_1 - x_2\|^2 \int_0^1 (1 - t) dt \\ &= \frac{1}{2} \sigma \alpha(1 - \alpha) \|x_1 - x_2\|^2, \end{aligned}$$

verifying the strong convexity inequality for x_1, x_2 in the interior of C [and using the continuity of f , for x_1, x_2 in the boundary of C as well].

Next we show that (iii) is equivalent to (i) and (ii). Indeed, by taking the limit in Eqs. (1.6) as $\alpha \rightarrow 0$, we see that (i) implies (iii). Conversely if (iii) holds, we have

$$f(y) \geq f(x) + \nabla f(x)'(y - x) + \frac{\sigma}{2} \|x - y\|^2, \quad \forall x, y \in \text{int}(C),$$

and

$$f(x) \geq f(y) + \nabla f(y)'(x - y) + \frac{\sigma}{2} \|x - y\|^2, \quad \forall x, y \in \text{int}(C).$$

By adding these two relations, we obtain (ii).

Assume now that f is twice continuously differentiable over $\text{int}(C)$. First we show that (iv) implies (ii). Let $x, y \in \text{int}(C)$ and consider the function $g : \mathfrak{R} \mapsto \mathfrak{R}$ defined by

$$g(t) = \nabla f(tx + (1 - t)y)'(x - y).$$

Using the Mean Value Theorem, we have

$$(\nabla f(x) - \nabla f(y))'(x - y) = g(1) - g(0) = \frac{dg(t)}{dt}$$

for some $t \in [0, 1]$. On the other hand,

$$\frac{dg(t)}{dt} = (x - y)' \nabla^2 f(tx + (1 - t)y)(x - y) \geq \sigma \|x - y\|^2,$$

where the last inequality holds because $\nabla^2 f(tx + (1-t)y) - \sigma I$ is positive semidefinite. Combining the last two relations, we obtain the desired inequality.

We finally show that (i) implies (iv). For any $\alpha \in (0, 1)$ and $x_1, x_2 \in \text{int}(C)$ with $x_1 \neq x_2$, let

$$x_\alpha = \alpha x_1 + (1 - \alpha)x_2.$$

Using the 2nd order Mean Value Theorem, we have

$$f(x_1) = f(x_\alpha) + \nabla f(x_\alpha)'(x_1 - x_\alpha) + \frac{1}{2}(x_1 - x_\alpha)'\nabla^2 f(\tilde{x}_\alpha)(x_1 - x_\alpha),$$

$$f(x_2) = f(x_\alpha) + \nabla f(x_\alpha)'(x_2 - x_\alpha) + \frac{1}{2}(x_2 - x_\alpha)'\nabla^2 f(\hat{x}_\alpha)(x_2 - x_\alpha),$$

where \tilde{x}_α and \hat{x}_α are vectors that lie in the intervals connecting x_α with x_1 and x_2 , respectively. Multiplying these relations with α and $1 - \alpha$, respectively, adding, canceling the terms involving $\nabla f(x_\alpha)$, and using the relations $x_\alpha - x_1 = (1 - \alpha)(x_2 - x_1)$ and $x_\alpha - x_2 = \alpha(x_1 - x_2)$ and the definition of strong convexity, we obtain

$$\begin{aligned} f(x_\alpha) + \frac{1}{2}\sigma\alpha(1 - \alpha)\|x_1 - x_2\|^2 &\leq \alpha f(x_1) + (1 - \alpha)f(x_2) \\ &= f(x_\alpha) + \frac{1}{2}\alpha(1 - \alpha)^2(x_1 - x_2)'\nabla^2 f(\tilde{x}_\alpha)(x_1 - x_2) \\ &\quad + \frac{1}{2}\alpha^2(1 - \alpha)(x_1 - x_2)'\nabla^2 f(\hat{x}_\alpha)(x_1 - x_2) \end{aligned}$$

and finally,

$$\sigma\|x_1 - x_2\|^2 \leq (x_1 - x_2)'((1 - \alpha)\nabla^2 f(\tilde{x}_\alpha) + \alpha\nabla^2 f(\hat{x}_\alpha))(x_1 - x_2).$$

Dividing by $\|x_1 - x_2\|^2$ and letting x_2 approach x_1 , we obtain

$$\sigma \leq d'\nabla^2 f(x_1)d,$$

where $d = (x_1 - x_2)/\|x_1 - x_2\|$. Since x_1 and x_2 were chosen arbitrarily within $\text{int}(C)$, it follows that the matrix $\nabla^2 f(x) - \sigma I$ is positive semidefinite for every $x \in \text{int}(C)$. Since by the convexity of C , every point in the boundary of C can be approached from the interior and $\nabla^2 f$ is continuous, $\nabla^2 f(x) - \sigma I$ is also positive semidefinite for every x in the boundary of C .

SECTION 1.2: Convex and Affine Hulls

1.11 (Characterization of Convex Hulls and Affine Hulls)

Let X be a nonempty subset of \mathbb{R}^n .

- (a) Show that the convex hull of X coincides with the set of all convex combinations of its elements, i.e.,

$$\text{conv}(X) = \left\{ \sum_{i \in I} \alpha_i x_i \mid I: \text{ a finite set, } \sum_{i \in I} \alpha_i = 1, \alpha_i \geq 0, x_i \in X, \forall i \in I \right\}.$$

- (b) Show that the affine hull of X coincides with the set of all linear combinations of its elements with coefficients adding to 1, i.e.,

$$\text{aff}(X) = \left\{ \sum_{i \in I} \alpha_i x_i \mid I: \text{ a finite set, } \sum_{i \in I} \alpha_i = 1, x_i \in X, \forall i \in I \right\}.$$

Show also that if m is the dimension of $\text{aff}(X)$, there exist vectors $\bar{x}, \bar{x}_1, \dots, \bar{x}_m$ from X such that $\bar{x}_1 - \bar{x}, \dots, \bar{x}_m - \bar{x}$ form a basis for the subspace that is parallel to $\text{aff}(X)$.

Solution: (a) The elements of X belong to $\text{conv}(X)$, so all their convex combinations belong to $\text{conv}(X)$ since $\text{conv}(X)$ is a convex set. On the other hand, consider any two convex combinations of elements of X , $x = \lambda_1 x_1 + \dots + \lambda_m x_m$ and $y = \mu_1 y_1 + \dots + \mu_r y_r$, where $x_i \in X$ and $y_j \in X$. The vector

$$(1 - \alpha)x + \alpha y = (1 - \alpha)(\lambda_1 x_1 + \dots + \lambda_m x_m) + \alpha(\mu_1 y_1 + \dots + \mu_r y_r),$$

where $0 \leq \alpha \leq 1$, is another convex combination of elements of X .

Thus, the set of convex combinations of elements of X is itself a convex set, which contains X , and is contained in $\text{conv}(X)$. Hence it must coincide with $\text{conv}(X)$, which by definition is the intersection of all convex sets containing X .

- (b) The set

$$A = \left\{ \sum_{i \in I} \alpha_i x_i \mid I \text{ is a finite set, } \sum_{i \in I} \alpha_i = 1, x_i \in X, \forall i \in I \right\}$$

contains every line that passes through any pair of its points, so it is affine. Since it also contains X , it must contain $\text{aff}(X)$.

To show the reverse inclusion, we note that

$$A = \bar{x} + S,$$

where \bar{x} is some vector in X and S is a subspace that must have the form

$$S = \left\{ \sum_{i \in I} \alpha_i (x_i - \bar{x}) \mid I \text{ is a finite set, } \sum_{i \in I} \alpha_i = 1, x_i \in X, \forall i \in I \right\}.$$

By taking one of the vectors x_i to be \bar{x} , we see that

$$S = \left\{ \sum_{i \in I} \beta_i (x_i - \bar{x}) \mid I \text{ is a finite set, } \beta_i \in \mathfrak{R}, x_i \in X, \forall i \in I \right\}.$$

It follows that S has a basis of the form $\bar{x}_1 - \bar{x}, \dots, \bar{x}_m - \bar{x}$, where $\bar{x}_1, \dots, \bar{x}_m \in X$, and m is the dimension of S . The subspace that is parallel to an affine set that contains X must contain the basis $\bar{x}_1 - \bar{x}, \dots, \bar{x}_m - \bar{x}$. Hence any affine set that contains X , including $\text{aff}(X)$, must contain $A = \bar{x} + S$. The proof of $\text{aff}(X) = A$ is complete.

1.12

Let $\{C_i \mid i \in I\}$ be an arbitrary collection of convex sets in \mathfrak{R}^n , and let C be the convex hull of the union of the collection. Show that

$$C = \bigcup_{\bar{I} \subset I, \bar{I}: \text{finite set}} \left\{ \sum_{i \in \bar{I}} \alpha_i C_i \mid \sum_{i \in \bar{I}} \alpha_i = 1, \alpha_i \geq 0, \forall i \in \bar{I} \right\},$$

i.e., the convex hull of the union of the sets C_i is equal to the set of all convex combinations of vectors that come from different sets C_i .

Solution: By Exercise 1.11, C is the set of all convex combinations $x = \alpha_1 y_1 + \cdots + \alpha_m y_m$, where m is a positive integer, and the vectors y_1, \dots, y_m belong to the union of the sets C_i . Actually, we can get C just by taking those combinations in which the vectors are taken from different sets C_i . Indeed, if two of the vectors, y_1 and y_2 belong to the same C_i , then the term $\alpha_1 y_1 + \alpha_2 y_2$ can be replaced by αy , where $\alpha = \alpha_1 + \alpha_2$ and

$$y = (\alpha_1/\alpha)y_1 + (\alpha_2/\alpha)y_2 \in C_i.$$

Thus, C is the union of the vector sums of the form

$$\alpha_1 C_{i_1} + \cdots + \alpha_m C_{i_m},$$

with

$$\alpha_i \geq 0, \forall i = 1, \dots, m, \quad \sum_{i=1}^m \alpha_i = 1,$$

and the indices i_1, \dots, i_m are all different, proving our claim.

1.13 (Generated Cones and Convex Hulls I)

Show that:

- (a) For a nonempty convex subset C of \mathfrak{R}^n , we have

$$\text{cone}(C) = \bigcup_{x \in C} \{\gamma x \mid \gamma \geq 0\}.$$

- (b) A cone C is convex if and only if $C + C \subset C$.

- (c) For any two convex cones C_1 and C_2 containing the origin, we have

$$C_1 + C_2 = \text{conv}(C_1 \cup C_2), \quad C_1 \cap C_2 = \bigcup_{\alpha \in [0,1]} (\alpha C_1 \cap (1-\alpha)C_2).$$

Solution: (a) Let $y \in \text{cone}(C)$. If $y = 0$, then $y \in \bigcup_{x \in C} \{\gamma x \mid \gamma \geq 0\}$. If $y \neq 0$, then by definition of $\text{cone}(C)$, we have

$$y = \sum_{i=1}^m \lambda_i x_i,$$

for some positive integer m , nonnegative scalars λ_i , and vectors $x_i \in C$. Since $y \neq 0$, we cannot have all λ_i equal to zero, implying that $\sum_{i=1}^m \lambda_i > 0$. Because $x_i \in C$ for all i and C is convex, the vector

$$x = \sum_{i=1}^m \frac{\lambda_i}{\sum_{j=1}^m \lambda_j} x_i$$

belongs to C . For this vector, we have

$$y = \left(\sum_{i=1}^m \lambda_i \right) x,$$

with $\sum_{i=1}^m \lambda_i > 0$, implying that $y \in \cup_{x \in C} \{\gamma x \mid \gamma \geq 0\}$ and showing that

$$\text{cone}(C) \subset \cup_{x \in C} \{\gamma x \mid \gamma \geq 0\}.$$

The reverse inclusion follows from the definition of $\text{cone}(C)$.

(b) Let C be a cone such that $C + C \subset C$, and let $x, y \in C$ and $\alpha \in [0, 1]$. Then since C is a cone, $\alpha x \in C$ and $(1 - \alpha)y \in C$, so that $\alpha x + (1 - \alpha)y \in C + C \subset C$, showing that C is convex. Conversely, let C be a convex cone and let $x, y \in C$. Then, since C is a cone, $2x \in C$ and $2y \in C$, so that by the convexity of C , $x + y = \frac{1}{2}(2x + 2y) \in C$, showing that $C + C \subset C$.

(c) First we prove that $C_1 + C_2 \subset \text{conv}(C_1 \cup C_2)$. Choose any $x \in C_1 + C_2$. Since $C_1 + C_2$ is a cone [see Exercise 1.2(c)], the vector $2x$ is in $C_1 + C_2$, so that $2x = x_1 + x_2$ for some $x_1 \in C_1$ and $x_2 \in C_2$. Therefore,

$$x = \frac{1}{2}x_1 + \frac{1}{2}x_2,$$

showing that $x \in \text{conv}(C_1 \cup C_2)$.

Next, we show that $\text{conv}(C_1 \cup C_2) \subset C_1 + C_2$. Since $0 \in C_1$ and $0 \in C_2$, it follows that

$$C_i = C_i + 0 \subset C_1 + C_2, \quad i = 1, 2,$$

implying that

$$C_1 \cup C_2 \subset C_1 + C_2.$$

By taking the convex hull of both sides in the above inclusion and by using the convexity of $C_1 + C_2$, we obtain

$$\text{conv}(C_1 \cup C_2) \subset \text{conv}(C_1 + C_2) = C_1 + C_2.$$

We finally show that

$$C_1 \cap C_2 = \bigcup_{\alpha \in [0,1]} (\alpha C_1 \cap (1 - \alpha)C_2).$$

We claim that for all α with $0 < \alpha < 1$, we have

$$\alpha C_1 \cap (1 - \alpha)C_2 = C_1 \cap C_2.$$

Indeed, if $x \in C_1 \cap C_2$, it follows that $x \in C_1$ and $x \in C_2$. Since C_1 and C_2 are cones and $0 < \alpha < 1$, we have $x \in \alpha C_1$ and $x \in (1 - \alpha)C_2$. Conversely, if $x \in \alpha C_1 \cap (1 - \alpha)C_2$, we have

$$\frac{x}{\alpha} \in C_1,$$

and

$$\frac{x}{(1 - \alpha)} \in C_2.$$

Since C_1 and C_2 are cones, it follows that $x \in C_1$ and $x \in C_2$, so that $x \in C_1 \cap C_2$.

If $\alpha = 0$ or $\alpha = 1$, we obtain

$$\alpha C_1 \cap (1 - \alpha)C_2 = \{0\} \subset C_1 \cap C_2,$$

since C_1 and C_2 contain the origin. Thus, the result follows.

1.14 (Generated Cones and Convex Hulls II)

Let X be a nonempty set. Show that:

- (a) X , $\text{conv}(X)$, and $\text{cl}(X)$ have the same affine hull.
- (b) $\text{cone}(X) = \text{cone}(\text{conv}(X))$.
- (c) $\text{aff}(\text{conv}(X)) \subset \text{aff}(\text{cone}(X))$. Give an example where the inclusion is strict.
- (d) If the origin belongs to $\text{conv}(X)$, then $\text{aff}(\text{conv}(X)) = \text{aff}(\text{cone}(X))$.
- (e) If A is a matrix, $A \text{conv}(X) = \text{conv}(AX)$.

Solution: (a) We first show that X and $\text{cl}(X)$ have the same affine hull. Since $X \subset \text{cl}(X)$, there holds

$$\text{aff}(X) \subset \text{aff}(\text{cl}(X)).$$

Conversely, because $X \subset \text{aff}(X)$ and $\text{aff}(X)$ is closed, we have $\text{cl}(X) \subset \text{aff}(X)$, implying that

$$\text{aff}(\text{cl}(X)) \subset \text{aff}(X).$$

We now show that X and $\text{conv}(X)$ have the same affine hull. By using a translation argument if necessary, we assume without loss of generality that X contains the origin, so that both $\text{aff}(X)$ and $\text{aff}(\text{conv}(X))$ are subspaces. Since $X \subset \text{conv}(X)$, evidently $\text{aff}(X) \subset \text{aff}(\text{conv}(X))$. To show the reverse inclusion, let the dimension of $\text{aff}(\text{conv}(X))$ be m , and let x_1, \dots, x_m be linearly independent vectors in $\text{conv}(X)$ that span $\text{aff}(\text{conv}(X))$. Then every $x \in \text{aff}(\text{conv}(X))$ is a linear combination of the vectors x_1, \dots, x_m , i.e., there exist scalars β_1, \dots, β_m such that

$$x = \sum_{i=1}^m \beta_i x_i.$$

By the definition of convex hull, each x_i is a convex combination of vectors in X , so that x is a linear combination of vectors in X , implying that $x \in \text{aff}(X)$. Hence, $\text{aff}(\text{conv}(X)) \subset \text{aff}(X)$.

(b) Since $X \subset \text{conv}(X)$, clearly $\text{cone}(X) \subset \text{cone}(\text{conv}(X))$. Conversely, let $x \in \text{cone}(\text{conv}(X))$. Then x is a nonnegative combination of some vectors in $\text{conv}(X)$, i.e., for some positive integer p , vectors $x_1, \dots, x_p \in \text{conv}(X)$, and nonnegative scalars $\alpha_1, \dots, \alpha_p$, we have

$$x = \sum_{i=1}^p \alpha_i x_i.$$

Each x_i is a convex combination of some vectors in X , so that x is a nonnegative combination of some vectors in X , implying that $x \in \text{cone}(X)$. Hence $\text{cone}(\text{conv}(X)) \subset \text{cone}(X)$.

(c) Since $\text{conv}(X)$ is the set of all convex combinations of vectors in X , and $\text{cone}(X)$ is the set of all nonnegative combinations of vectors in X , it follows that $\text{conv}(X) \subset \text{cone}(X)$. Therefore

$$\text{aff}(\text{conv}(X)) \subset \text{aff}(\text{cone}(X)).$$

For an example showing that the above inclusion can be strict, consider the set $X = \{(1, 1)\}$ in \mathbb{R}^2 . Then $\text{conv}(X) = X$, so that

$$\text{aff}(\text{conv}(X)) = X = \{(1, 1)\},$$

and the dimension of $\text{conv}(X)$ is zero. On the other hand, $\text{cone}(X) = \{(\alpha, \alpha) \mid \alpha \geq 0\}$, so that

$$\text{aff}(\text{cone}(X)) = \{(x_1, x_2) \mid x_1 = x_2\},$$

and the dimension of $\text{cone}(X)$ is one.

(d) In view of parts (a) and (c), it suffices to show that

$$\text{aff}(\text{cone}(X)) \subset \text{aff}(X).$$

It is always true that $0 \in \text{cone}(X)$, so $\text{aff}(\text{cone}(X))$ is a subspace. Let the dimension of $\text{aff}(\text{cone}(X))$ be m , and let x_1, \dots, x_m be linearly independent vectors in $\text{cone}(X)$ that span $\text{aff}(\text{cone}(X))$ [cf. Exercise 1.11(b)]. Since every vector in $\text{aff}(\text{cone}(X))$ is a linear combination of x_1, \dots, x_m , and since each x_i is a nonnegative combination of some vectors in X , it follows that every vector in $\text{aff}(\text{cone}(X))$ is a linear combination of some vectors in X . In view of the assumption that $0 \in \text{conv}(X)$, the affine hull of $\text{conv}(X)$ is a subspace, which implies by part (a) that the affine hull of X is a subspace. Hence, $\text{aff}(X)$ is the set of linear combinations of vectors from X . It follows that every vector in $\text{aff}(\text{cone}(X))$ belongs to $\text{aff}(X)$, showing that $\text{aff}(\text{cone}(X)) \subset \text{aff}(X)$.

(e) If $y \in \text{conv}(AX)$, then for some $x_1, x_2 \in X$,

$$y = \alpha Ax_1 + (1 - \alpha)Ax_2 = A(\alpha x_1 + (1 - \alpha)x_2) \in A \text{conv}(X).$$

Hence $\text{conv}(AX) \subset A \text{conv}(X)$.

Conversely, if $y \in A \text{conv}(X)$, then for some $x_1, x_2 \in X$,

$$y = A(\alpha x_1 + (1 - \alpha)x_2) = \text{conv}(AX).$$

Hence $A \text{conv}(X) \subset \text{conv}(AX)$.

1.15

Let $\{f_i \mid i \in I\}$ be an arbitrary collection of proper convex functions $f_i : \mathbb{R}^n \mapsto (-\infty, \infty]$. Define

$$f(x) = \inf \left\{ w \mid (x, w) \in \text{conv}(\cup_{i \in I} \text{epi}(f_i)) \right\}, \quad x \in \mathbb{R}^n.$$

Show that $f(x)$ is given by

$$f(x) = \inf \left\{ \sum_{i \in \bar{I}} \alpha_i f_i(x_i) \mid \sum_{i \in \bar{I}} \alpha_i x_i = x, x_i \in \mathbb{R}^n, \sum_{i \in \bar{I}} \alpha_i = 1, \alpha_i \geq 0, \forall i \in \bar{I}, \right. \\ \left. \bar{I} \subset I, \bar{I}: \text{finite} \right\}.$$

Solution: By definition, $f(x)$ is the infimum of the values of w such that $(x, w) \in C$, where C is the convex hull of the union of nonempty convex sets $\text{epi}(f_i)$. By Exercise 1.12, $(x, w) \in C$ if and only if (x, w) can be expressed as a convex combination of the form

$$(x, w) = \sum_{i \in \bar{I}} \alpha_i (x_i, w_i) = \left(\sum_{i \in \bar{I}} \alpha_i x_i, \sum_{i \in \bar{I}} \alpha_i w_i \right),$$

where $\bar{I} \subset I$ is a finite set and $(x_i, w_i) \in \text{epi}(f_i)$ for all $i \in \bar{I}$. Thus, $f(x)$ can be expressed as

$$f(x) = \inf \left\{ \sum_{i \in \bar{I}} \alpha_i w_i \mid (x, w) = \sum_{i \in \bar{I}} \alpha_i (x_i, w_i), \right. \\ \left. (x_i, w_i) \in \text{epi}(f_i), \alpha_i \geq 0, \forall i \in \bar{I}, \sum_{i \in \bar{I}} \alpha_i = 1 \right\}.$$

Since the set $\{(x_i, f_i(x_i)) \mid x_i \in \mathfrak{R}^n\}$ is contained in $\text{epi}(f_i)$, we obtain

$$f(x) \leq \inf \left\{ \sum_{i \in \bar{I}} \alpha_i f_i(x_i) \mid x = \sum_{i \in \bar{I}} \alpha_i x_i, x_i \in \mathfrak{R}^n, \alpha_i \geq 0, \forall i \in \bar{I}, \sum_{i \in \bar{I}} \alpha_i = 1 \right\}.$$

On the other hand, by the definition of $\text{epi}(f_i)$, for each $(x_i, w_i) \in \text{epi}(f_i)$ we have $w_i \geq f_i(x_i)$, implying that

$$f(x) \geq \inf \left\{ \sum_{i \in \bar{I}} \alpha_i f_i(x_i) \mid x = \sum_{i \in \bar{I}} \alpha_i x_i, x_i \in \mathfrak{R}^n, \alpha_i \geq 0, \forall i \in \bar{I}, \sum_{i \in \bar{I}} \alpha_i = 1 \right\}.$$

By combining the last two relations, we obtain

$$f(x) = \inf \left\{ \sum_{i \in \bar{I}} \alpha_i f_i(x_i) \mid x = \sum_{i \in \bar{I}} \alpha_i x_i, x_i \in \mathfrak{R}^n, \alpha_i \geq 0, \forall i \in \bar{I}, \sum_{i \in \bar{I}} \alpha_i = 1 \right\},$$

where the infimum is taken over all representations of x as a convex combination of elements x_i such that only finitely many coefficients α_i are nonzero.

1.16 (Minimization of Linear Functions)

Show that minimization of a linear function over a set is equivalent to minimization over its convex hull, i.e.,

$$\inf_{x \in \text{conv}(X)} c'x = \inf_{x \in X} c'x,$$

if $X \subset \mathfrak{R}^n$ and $c \in \mathfrak{R}^n$. Furthermore, the infimum in the left-hand side above is attained if and only if the infimum in the right-hand side is attained.

Solution: Since $X \subset \text{conv}(X)$, we have

$$\inf_{x \in \text{conv}(X)} c'x \leq \inf_{x \in X} c'x. \quad (1.7)$$

Also, any $\bar{x} \in \text{conv}(X)$ can be written as $\bar{x} = \sum_{i=1}^m \alpha_i x_i$, for some $x_1, \dots, x_m \in X$ and some scalars $\alpha_1, \dots, \alpha_m \geq 0$ with $\sum_{i=1}^m \alpha_i = 1$. Hence, since $c'x_i \geq \inf_{x \in X} c'x$, we have

$$c'\bar{x} = \sum_{i=1}^m \alpha_i c'x_i \geq \left(\sum_{i=1}^m \alpha_i \right) \inf_{x \in X} c'x = \inf_{x \in X} c'x, \quad \forall \bar{x} \in \text{conv}(X).$$

Taking the infimum of the left-hand side over $\bar{x} \in \text{conv}(X)$,

$$\inf_{\bar{x} \in \text{conv}(X)} c'\bar{x} \geq \inf_{x \in X} c'x. \quad (1.8)$$

Combining Eqs. (1.7) and (1.8), we obtain

$$\inf_{x \in \text{conv}(X)} c'x = \inf_{x \in X} c'x.$$

Since $X \subset \text{conv}(X)$ and $\inf_{x \in \text{conv}(X)} c'x = \inf_{x \in X} c'x$, every point that attains the infimum of $c'x$ over X , attains the infimum of $c'x$ over $\text{conv}(X)$. For the converse, assume that the infimum of $c'x$ over $\text{conv}(X)$ is attained at some $\bar{x} \in \text{conv}(X)$. Then, $\bar{x} = \sum_{i=1}^m \alpha_i x_i$, for some $x_1, \dots, x_m \in X$ and some scalars $\alpha_1, \dots, \alpha_m \geq 0$ with $\sum_{i=1}^m \alpha_i = 1$, and we have

$$\inf_{x \in X} c'x = \left(\sum_{i=1}^m \alpha_i \right) \inf_{x \in X} c'x \leq \sum_{i=1}^m \alpha_i c'x_i = c'\bar{x} = \inf_{x \in \text{conv}(X)} c'x = \inf_{x \in X} c'x.$$

Since the left-hand and right-hand sides are equal, it follows that equality holds throughout above, which can happen only if $c'x_i = \inf_{x \in X} c'x$ for all i with $\alpha_i > 0$. Thus the infimum of $c'x$ over X is attained.

1.17 (Extension of Caratheodory's Theorem)

Let X_1 and X_2 be nonempty subsets of \mathfrak{R}^n , and let $X = \text{conv}(X_1) + \text{cone}(X_2)$. Show that every vector x in X can be represented in the form

$$x = \sum_{i=1}^k \alpha_i x_i + \sum_{i=k+1}^m \alpha_i y_i,$$

where m is a positive integer with $m \leq n+1$, the vectors x_1, \dots, x_k belong to X_1 , the vectors y_{k+1}, \dots, y_m belong to X_2 , and the scalars $\alpha_1, \dots, \alpha_m$ are nonnegative with $\alpha_1 + \dots + \alpha_k = 1$. Furthermore, the vectors $x_2 - x_1, \dots, x_k - x_1, y_{k+1}, \dots, y_m$ are linearly independent.

Solution: The proof will be an application of Caratheodory's Theorem [Prop. 1.2.1(a)] to the subset of \mathfrak{R}^{n+1} given by

$$Y = \{(x, 1) \mid x \in X_1\} \cup \{(y, 0) \mid y \in X_2\}.$$

If $x \in X$, then

$$x = \sum_{i=1}^k \gamma_i x_i + \sum_{i=k+1}^m \gamma_i y_i,$$

where the vectors x_1, \dots, x_k belong to X_1 , the vectors y_{k+1}, \dots, y_m belong to X_2 , and the scalars $\gamma_1, \dots, \gamma_m$ are nonnegative with $\gamma_1 + \dots + \gamma_k = 1$. Equivalently, $(x, 1) \in \text{cone}(Y)$. By Caratheodory's Theorem part (a), we have that

$$(x, 1) = \sum_{i=1}^k \alpha_i (x_i, 1) + \sum_{i=k+1}^m \alpha_i (y_i, 0),$$

for some positive scalars $\alpha_1, \dots, \alpha_m$ and vectors

$$(x_1, 1), \dots, (x_k, 1), (y_{k+1}, 0), \dots, (y_m, 0),$$

which are linearly independent (implying that $m \leq n + 1$) or equivalently,

$$x = \sum_{i=1}^k \alpha_i x_i + \sum_{i=k+1}^m \alpha_i y_i, \quad 1 = \sum_{i=1}^k \alpha_i.$$

Finally, to show that the vectors $x_2 - x_1, \dots, x_k - x_1, y_{k+1}, \dots, y_m$ are linearly independent, assume to arrive at a contradiction, that there exist $\lambda_2, \dots, \lambda_m$, not all 0, such that

$$\sum_{i=2}^k \lambda_i (x_i - x_1) + \sum_{i=k+1}^m \lambda_i y_i = 0.$$

Equivalently, defining $\lambda_1 = -(\lambda_2 + \dots + \lambda_m)$, we have

$$\sum_{i=1}^k \lambda_i (x_i, 1) + \sum_{i=k+1}^m \lambda_i (y_i, 0) = 0,$$

which contradicts the linear independence of the vectors

$$(x_1, 1), \dots, (x_k, 1), (y_{k+1}, 0), \dots, (y_m, 0).$$

1.18

Let X be a nonempty bounded subset of \mathfrak{R}^n . Show that

$$\text{cl}(\text{conv}(X)) = \text{conv}(\text{cl}(X)).$$

In particular, if X is compact, then $\text{conv}(X)$ is compact (cf. Prop. 1.2.2).

Solution: The set $\text{cl}(X)$ is compact since X is bounded by assumption. Hence, by Prop. 1.2.2, its convex hull, $\text{conv}(\text{cl}(X))$, is compact, and it follows that

$$\text{cl}(\text{conv}(X)) \subset \text{cl}(\text{conv}(\text{cl}(X))) = \text{conv}(\text{cl}(X)).$$

It is also true that

$$\text{conv}(\text{cl}(X)) \subset \text{conv}(\text{cl}(\text{conv}(X))) = \text{cl}(\text{conv}(X)),$$

since by Prop. 1.1.1(d), the closure of a convex set is convex. Hence, the result follows.

1.19 (Convex Hulls and Generated Cones of Cartesian Products)

Given nonempty sets $X_i \subset \mathfrak{R}^{n_i}$, $i = 1, \dots, m$, let $X = X_1 \times \dots \times X_m$ be their Cartesian product. Show that:

- (a) The convex hull (closure, affine hull) of X is equal to the Cartesian product of the convex hulls (closures, affine hulls, respectively) of the X_i .
- (b) If all the sets X_1, \dots, X_m contain the origin, then

$$\text{cone}(X) = \text{cone}(X_1) \times \dots \times \text{cone}(X_m).$$

Furthermore, the result fails if one of the sets does not contain the origin.

Solution: (a) We first show that the convex hull of X is equal to the Cartesian product of the convex hulls of the sets X_i , $i = 1, \dots, m$. Let y be a vector that belongs to $\text{conv}(X)$. Then, by definition, for some k , we have

$$y = \sum_{i=1}^k \alpha_i y_i, \quad \text{with } \alpha_i \geq 0, \quad i = 1, \dots, m, \quad \sum_{i=1}^k \alpha_i = 1,$$

where $y_i \in X$ for all i . Since $y_i \in X$, we have that $y_i = (x_1^i, \dots, x_m^i)$ for all i , with $x_1^i \in X_1, \dots, x_m^i \in X_m$. It follows that

$$y = \sum_{i=1}^k \alpha_i (x_1^i, \dots, x_m^i) = \left(\sum_{i=1}^k \alpha_i x_1^i, \dots, \sum_{i=1}^k \alpha_i x_m^i \right),$$

thereby implying that $y \in \text{conv}(X_1) \times \dots \times \text{conv}(X_m)$.

To prove the reverse inclusion, assume that y is a vector in $\text{conv}(X_1) \times \dots \times \text{conv}(X_m)$. Then, we can represent y as $y = (y_1, \dots, y_m)$ with $y_i \in \text{conv}(X_i)$, i.e., for all $i = 1, \dots, m$, we have

$$y_i = \sum_{j=1}^{k_i} \alpha_j^i x_j^i, \quad x_j^i \in X_i, \quad \forall j, \quad \alpha_j^i \geq 0, \quad \forall j, \quad \sum_{j=1}^{k_i} \alpha_j^i = 1.$$

First, consider the vectors

$$(x_1^1, x_{r_1}^2, \dots, x_{r_{m-1}}^m), (x_2^1, x_{r_1}^2, \dots, x_{r_{m-1}}^m), \dots, (x_{k_1}^1, x_{r_1}^2, \dots, x_{r_{m-1}}^m),$$

for all possible values of r_1, \dots, r_{m-1} , i.e., we fix all components except the first one, and vary the first component over all possible x_j^1 's used in the convex combination that yields y_1 . Since all these vectors belong to X , their convex combination given by

$$\left(\left(\sum_{j=1}^{k_1} \alpha_j^1 x_j^1 \right), x_{r_1}^2, \dots, x_{r_{m-1}}^m \right)$$

belongs to the convex hull of X for all possible values of r_1, \dots, r_{m-1} . Now, consider the vectors

$$\left(\left(\sum_{j=1}^{k_1} \alpha_j^1 x_j^1 \right), x_1^2, \dots, x_{r_{m-1}}^m \right), \dots, \left(\left(\sum_{j=1}^{k_1} \alpha_j^1 x_j^1 \right), x_{k_2}^2, \dots, x_{r_{m-1}}^m \right),$$

i.e., fix all components except the second one, and vary the second component over all possible x_j^2 's used in the convex combination that yields y_2 . Since all these vectors belong to $\text{conv}(X)$, their convex combination given by

$$\left(\left(\sum_{j=1}^{k_1} \alpha_j^1 x_j^1 \right), \left(\sum_{j=1}^{k_2} \alpha_j^2 x_j^2 \right), \dots, x_{r_{m-1}}^m \right)$$

belongs to the convex hull of X for all possible values of r_2, \dots, r_{m-1} . Proceeding in this way, we see that the vector given by

$$\left(\left(\sum_{j=1}^{k_1} \alpha_j^1 x_j^1 \right), \left(\sum_{j=1}^{k_2} \alpha_j^2 x_j^2 \right), \dots, \left(\sum_{j=1}^{k_m} \alpha_j^m x_j^m \right) \right)$$

belongs to $\text{conv}(X)$, thus proving our claim.

Next, we show the corresponding result for the closure of X . Assume that $y = (x_1, \dots, x_m) \in \text{cl}(X)$. This implies that there exists some sequence $\{y^k\} \subset X$ such that $y^k \rightarrow y$. Since $y^k \in X$, we have that $y^k = (x_1^k, \dots, x_m^k)$ with $x_i^k \in X_i$ for each i and k . Since $y^k \rightarrow y$, it follows that $x_i \in \text{cl}(X_i)$ for each i , and hence $y \in \text{cl}(X_1) \times \dots \times \text{cl}(X_m)$. Conversely, suppose that $y = (x_1, \dots, x_m) \in \text{cl}(X_1) \times \dots \times \text{cl}(X_m)$. This implies that there exist sequences $\{x_i^k\} \subset X_i$ such that $x_i^k \rightarrow x_i$ for each $i = 1, \dots, m$. Since $x_i^k \in X_i$ for each i and k , we have that $y^k = (x_1^k, \dots, x_m^k) \in X$ and $\{y^k\}$ converges to $y = (x_1, \dots, x_m)$, implying that $y \in \text{cl}(X)$.

Finally, we show the corresponding result for the affine hull of X . Let's assume, by using a translation argument if necessary, that all the X_i 's contain the origin, so that $\text{aff}(X_1), \dots, \text{aff}(X_m)$ as well as $\text{aff}(X)$ are all subspaces.

Assume that $y \in \text{aff}(X)$. Let the dimension of $\text{aff}(X)$ be r , and let y^1, \dots, y^r be linearly independent vectors in X that span $\text{aff}(X)$. Thus, we can represent y as

$$y = \sum_{i=1}^r \beta^i y^i,$$

where β^1, \dots, β^r are scalars. Since $y^i \in X$, we have that $y^i = (x_1^i, \dots, x_m^i)$ with $x_j^i \in X_j$. Thus,

$$y = \sum_{i=1}^r \beta^i (x_1^i, \dots, x_m^i) = \left(\sum_{i=1}^r \beta^i x_1^i, \dots, \sum_{i=1}^r \beta^i x_m^i \right),$$

implying that $y \in \text{aff}(X_1) \times \cdots \times \text{aff}(X_m)$. Now, assume that $y \in \text{aff}(X_1) \times \cdots \times \text{aff}(X_m)$. Let the dimension of $\text{aff}(X_i)$ be r_i , and let $x_i^1, \dots, x_i^{r_i}$ be linearly independent vectors in X_i that span $\text{aff}(X_i)$. Thus, we can represent y as

$$y = \left(\sum_{j=1}^{r_1} \beta_1^j x_1^j, \dots, \sum_{j=1}^{r_m} \beta_m^j x_m^j \right).$$

Since each X_i contains the origin, we have that the vectors

$$\left(\sum_{j=1}^{r_1} \beta_1^j x_1^j, 0, \dots, 0 \right), \left(0, \sum_{j=1}^{r_2} \beta_2^j x_2^j, 0, \dots, 0 \right), \dots, \left(0, \dots, \sum_{j=1}^{r_m} \beta_m^j x_m^j \right),$$

belong to $\text{aff}(X)$, and so does their sum, which is the vector y . Thus, $y \in \text{aff}(X)$, concluding the proof.

(b) Assume that $y \in \text{cone}(X)$. We can represent y as

$$y = \sum_{i=1}^r \alpha^i y^i,$$

for some r , where $\alpha^1, \dots, \alpha^r$ are nonnegative scalars and $y_i \in X$ for all i . Since $y^i \in X$, we have that $y^i = (x_1^i, \dots, x_m^i)$ with $x_j^i \in X_j$. Thus,

$$y = \sum_{i=1}^r \alpha^i (x_1^i, \dots, x_m^i) = \left(\sum_{i=1}^r \alpha^i x_1^i, \dots, \sum_{i=1}^r \alpha^i x_m^i \right),$$

implying that $y \in \text{cone}(X_1) \times \cdots \times \text{cone}(X_m)$.

Conversely, assume that $y \in \text{cone}(X_1) \times \cdots \times \text{cone}(X_m)$. Then, we can represent y as

$$y = \left(\sum_{j=1}^{r_1} \alpha_1^j x_1^j, \dots, \sum_{j=1}^{r_m} \alpha_m^j x_m^j \right),$$

where $x_i^j \in X_i$ and $\alpha_i^j \geq 0$ for each i and j . Since each X_i contains the origin, we have that the vectors

$$\left(\sum_{j=1}^{r_1} \alpha_1^j x_1^j, 0, \dots, 0 \right), \left(0, \sum_{j=1}^{r_2} \alpha_2^j x_2^j, 0, \dots, 0 \right), \dots, \left(0, \dots, \sum_{j=1}^{r_m} \alpha_m^j x_m^j \right),$$

belong to the $\text{cone}(X)$, and so does their sum, which is the vector y . Thus, $y \in \text{cone}(X)$, concluding the proof.

Finally, consider the example where

$$X_1 = \{0, 1\} \subset \mathfrak{R}, \quad X_2 = \{1\} \subset \mathfrak{R}.$$

For this example, $\text{cone}(X_1) \times \text{cone}(X_2)$ is given by the nonnegative quadrant, whereas $\text{cone}(X)$ is given by the two halflines $\alpha(0, 1)$ and $\alpha(1, 1)$ for $\alpha \geq 0$ and the region that lies between them.

SECTION 1.3: Relative Interior and Closure

1.20 (Characterization of Twice Continuously Differentiable Convex Functions)

Let C be a nonempty convex subset of \mathfrak{R}^n and let $f : \mathfrak{R}^n \mapsto \mathfrak{R}$ be twice continuously differentiable over \mathfrak{R}^n . Let S be the subspace that is parallel to the affine hull of C . Show that f is convex over C if and only if $y' \nabla^2 f(x) y \geq 0$ for all $x \in C$ and $y \in S$. [In particular, when C has nonempty interior, f is convex over C if and only if $\nabla^2 f(x)$ is positive semidefinite for all $x \in C$.]

Solution: Suppose that $f : \mathfrak{R}^n \mapsto \mathfrak{R}$ is convex over C . We first show that for all $x \in \text{ri}(C)$ and $y \in S$, we have $y' \nabla^2 f(x) y \geq 0$. Assume to arrive at a contradiction, that there exists some $\bar{x} \in \text{ri}(C)$ such that for some $y \in S$, we have

$$y' \nabla^2 f(\bar{x}) y < 0.$$

Without loss of generality, we may assume that $\|y\| = 1$. Using the continuity of $\nabla^2 f$, we see that there is an open ball $B(\bar{x}, \epsilon)$ centered at \bar{x} with radius ϵ such that $B(\bar{x}, \epsilon) \cap \text{aff}(C) \subset C$ [since $\bar{x} \in \text{ri}(C)$], and

$$y' \nabla^2 f(x) y < 0, \quad \forall x \in B(\bar{x}, \epsilon). \quad (1.9)$$

For all positive scalars α with $\alpha < \epsilon$, we have

$$f(\bar{x} + \alpha y) = f(\bar{x}) + \alpha \nabla f(\bar{x})' y + \frac{1}{2} y' \nabla^2 f(\bar{x} + \bar{\alpha} y) y,$$

for some $\bar{\alpha} \in [0, \alpha]$. Furthermore, $\|(\bar{x} + \bar{\alpha} y) - \bar{x}\| \leq \epsilon$ [since $\|y\| = 1$ and $\bar{\alpha} < \epsilon$]. Hence, from Eq. (1.9), it follows that

$$f(\bar{x} + \alpha y) < f(\bar{x}) + \alpha \nabla f(\bar{x})' y, \quad \forall \alpha \in [0, \epsilon].$$

On the other hand, by the choice of ϵ and the assumption that $y \in S$, the vectors $\bar{x} + \alpha y$ are in C for all α with $\alpha \in [0, \epsilon]$, which is a contradiction in view of the convexity of f over C . Hence, we have $y' \nabla^2 f(x) y \geq 0$ for all $y \in S$ and all $x \in \text{ri}(C)$.

Next, let \bar{x} be a point in C that is not in the relative interior of C . Then, by the Line Segment Principle, there is a sequence $\{x_k\} \subset \text{ri}(C)$ such that $x_k \rightarrow \bar{x}$. As seen above, $y' \nabla^2 f(x_k) y \geq 0$ for all $y \in S$ and all k , which together with the continuity of $\nabla^2 f$ implies that

$$y' \nabla^2 f(\bar{x}) y = \lim_{k \rightarrow \infty} y' \nabla^2 f(x_k) y \geq 0, \quad \forall y \in S.$$

It follows that $y' \nabla^2 f(x) y \geq 0$ for all $x \in C$ and $y \in S$.

Conversely, assume that $y' \nabla^2 f(x) y \geq 0$ for all $x \in C$ and $y \in S$. For all $x, z \in C$ we have

$$f(z) = f(x) + (z - x)' \nabla f(x) + \frac{1}{2} (z - x)' \nabla^2 f(x + \alpha(z - x)) (z - x)$$

for some $\alpha \in [0, 1]$. Since $x, z \in C$, we have that $(z - x) \in S$, and using the convexity of C and our assumption, it follows that

$$f(z) \geq f(x) + (z - x)' \nabla f(x), \quad \forall x, z \in C.$$

From Prop. 1.1.7(a), we conclude that f is convex over C .

1.21

Construct an example of a point in a nonconvex set X that has the prolongation property of Prop. 1.3.3 but is not a relative interior point of X .

Solution: Take two intersecting lines in the plane, and consider the point of intersection.

For another example, take the union of two circular disks in the plane, which have a single common point, and consider the common point.

1.22

Let C be a nonempty convex subset of \mathfrak{R}^n . Show that

$$\text{ri}(C) = \text{int}(C + S^\perp) \cap C,$$

where S is the subspace that is parallel to the affine hull of C .

Solution: For any vector $a \in \mathfrak{R}^n$, we have $\text{ri}(C + a) = \text{ri}(C) + a$ (cf. Prop. 1.3.7). Therefore, we can assume without loss of generality that $0 \in C$, and $\text{aff}(C)$ coincides with S .

Let $x \in \text{ri}(C)$. Then there exists some open ball $B(x, \epsilon)$ centered at x with radius $\epsilon > 0$ such that

$$B(x, \epsilon) \cap S \subset C. \tag{1.10}$$

We now show that $B(x, \epsilon) \subset C + S^\perp$. Let z be a vector in $B(x, \epsilon)$. Then, we can express z as $z = x + \alpha y$ for some vector $y \in \mathfrak{R}^n$ with $\|y\| = 1$, and some $\alpha \in [0, \epsilon)$. Since S and S^\perp are orthogonal subspaces, y can be uniquely decomposed as $y = y_S + y_{S^\perp}$, where $y_S \in S$ and $y_{S^\perp} \in S^\perp$. Since $\|y\| = 1$, this implies that $\|y_S\| \leq 1$ (Pythagorean Theorem), and using Eq. (1.10), we obtain

$$x + \alpha y_S \in B(x, \epsilon) \cap S \subset C,$$

from which it follows that the vector $z = x + \alpha y$ belongs to $C + S^\perp$, implying that $B(x, \epsilon) \subset C + S^\perp$. This shows that $x \in \text{int}(C + S^\perp) \cap C$.

Conversely, let $x \in \text{int}(C + S^\perp) \cap C$. We have that $x \in C$ and there exists some open ball $B(x, \epsilon)$ centered at x with radius $\epsilon > 0$ such that $B(x, \epsilon) \subset C + S^\perp$. Since C is a subset of S , it can be seen that $(C + S^\perp) \cap S = C$. Therefore,

$$B(x, \epsilon) \cap S \subset C,$$

implying that $x \in \text{ri}(C)$.

1.23

Let x_0, \dots, x_m be vectors in \mathfrak{R}^n such that $x_1 - x_0, \dots, x_m - x_0$ are linearly independent. The convex hull of x_0, \dots, x_m is called an m -dimensional simplex, and x_0, \dots, x_m are called the *vertices* of the simplex.

- (a) Show that the dimension of a convex set is the maximum of the dimensions of all the simplices contained in the set.
- (b) Use part (a) to show that a nonempty convex set has a nonempty relative interior.

Solution: (a) Let C be the given convex set. The convex hull of any subset of C is contained in C . Therefore, the maximum dimension of the various simplices contained in C is the largest m for which C contains $m + 1$ vectors x_0, \dots, x_m such that $x_1 - x_0, \dots, x_m - x_0$ are linearly independent.

Let $K = \{x_0, \dots, x_m\}$ be such a set with m maximal, and let $\text{aff}(K)$ denote the affine hull of set K . Then, we have $\dim(\text{aff}(K)) = m$, and since $K \subset C$, it follows that $\text{aff}(K) \subset \text{aff}(C)$.

We claim that $C \subset \text{aff}(K)$. To see this, assume that there exists some $x \in C$, which does not belong to $\text{aff}(K)$. This implies that the set $\{x, x_0, \dots, x_m\}$ is a set of $m + 2$ vectors in C such that $x - x_0, x_1 - x_0, \dots, x_m - x_0$ are linearly independent, contradicting the maximality of m . Hence, we have $C \subset \text{aff}(K)$, and it follows that

$$\text{aff}(K) = \text{aff}(C),$$

thereby implying that $\dim(C) = m$.

(b) We first consider the case where C is n -dimensional with $n > 0$ and show that the interior of C is not empty. By part (a), an n -dimensional convex set contains an n -dimensional simplex. We claim that such a simplex S has a nonempty interior. Indeed, applying an affine transformation if necessary, we can assume that the vertices of S are the vectors $(0, 0, \dots, 0), (1, 0, \dots, 0), \dots, (0, 0, \dots, 1)$, i.e.,

$$S = \left\{ (x_1, \dots, x_n) \mid x_i \geq 0, \forall i = 1, \dots, n, \sum_{i=1}^n x_i \leq 1 \right\}.$$

The interior of the simplex S ,

$$\text{int}(S) = \left\{ (x_1, \dots, x_n) \mid x_i > 0, \forall i = 1, \dots, n, \sum_{i=1}^n x_i < 1 \right\},$$

is nonempty, which in turn implies that $\text{int}(C)$ is nonempty.

For the case where $\dim(C) < n$, consider the n -dimensional set $C + S^\perp$, where S^\perp is the orthogonal complement of the subspace parallel to $\text{aff}(C)$. Since $C + S^\perp$ is a convex set, it follows from the above argument that $\text{int}(C + S^\perp)$ is nonempty. Let $x \in \text{int}(C + S^\perp)$. We can represent x as $x = x_C + x_{S^\perp}$, where $x_C \in C$ and $x_{S^\perp} \in S^\perp$. It can be seen that $x_C \in \text{int}(C + S^\perp)$. Since

$$\text{ri}(C) = \text{int}(C + S^\perp) \cap C,$$

[cf. Exercise 1.22(a)], it follows that $x_C \in \text{ri}(C)$, so $\text{ri}(C)$ is nonempty.

1.24 (Characterizations of Relative Interior)

Let C be a nonempty convex set.

- (a) Show the following refinement of the Prolongation Lemma (Prop. 1.3.3):
 $x \in \text{ri}(C)$ if and only if for every $\bar{x} \in \text{aff}(C)$, there exists a $\gamma > 0$ such that
 $x + \gamma(x - \bar{x}) \in C$.
- (b) Show that $\text{cone}(C) = \text{aff}(C)$ if and only if $0 \in \text{ri}(C)$.

Solution: (a) Let $x \in \text{ri}(C)$. We will show that for every $\bar{x} \in \text{aff}(C)$, there exists a $\gamma > 1$ such that $x + (\gamma - 1)(x - \bar{x}) \in C$. This is true if $\bar{x} = x$, so assume that $\bar{x} \neq x$. Since $x \in \text{ri}(C)$, there exists $\epsilon > 0$ such that

$$\{z \mid \|z - x\| < \epsilon\} \cap \text{aff}(C) \subset C.$$

Choose a point $\bar{x}_\epsilon \in C$ in the intersection of the ray $\{x + \alpha(\bar{x} - x) \mid \alpha \geq 0\}$ and the set $\{z \mid \|z - x\| < \epsilon\} \cap \text{aff}(C)$. Then, for some positive scalar α_ϵ ,

$$x - \bar{x}_\epsilon = \alpha_\epsilon(x - \bar{x}).$$

Since $x \in \text{ri}(C)$ and $\bar{x}_\epsilon \in C$, by Prop. 1.3.1(c), there is $\gamma_\epsilon > 1$ such that

$$x + (\gamma_\epsilon - 1)(x - \bar{x}_\epsilon) \in C,$$

which in view of the preceding relation implies that

$$x + (\gamma_\epsilon - 1)\alpha_\epsilon(x - \bar{x}) \in C.$$

The result follows by letting $\gamma = 1 + (\gamma_\epsilon - 1)\alpha_\epsilon$ and noting that $\gamma > 1$, since $(\gamma_\epsilon - 1)\alpha_\epsilon > 0$.

The converse assertion follows from the fact $C \subset \text{aff}(C)$ and Prop. 1.3.1(c).

(b) Assume that $0 \in \text{ri}(C)$. Then, the inclusion $\text{cone}(C) \subset \text{aff}(C)$ is evident. For the reverse inclusion, note that if $\bar{x} \in \text{aff}(C)$, then $-\bar{x} \in \text{aff}(C)$, so applying part (a) with $x = 0$, we have that $\gamma\bar{x} \in C$ for some $\gamma > 0$. Hence $\bar{x} \in \text{cone}(C)$ and $\text{aff}(C) \subset \text{cone}(C)$.

Conversely, assume that $\text{aff}(C) = \text{cone}(C)$. We will show that $0 \in \text{ri}(C)$. Indeed if this is not so, by applying part (a) with $x = 0$, it follows that there exists $\bar{x} \in \text{aff}(C)$ such that $\gamma(-\bar{x}) \notin C$ for all $\gamma > 0$. Hence $-\bar{x} \notin \text{cone}(C)$, a contradiction.

1.25

Let $f : \mathbb{R}^n \mapsto (-\infty, \infty]$ be a convex function, let γ be a scalar, and let C be a nonempty convex subset of \mathbb{R}^n .

- (a) Show that if $f(x) < \gamma$ for some x , then $f(x) < \gamma$ for some $x \in \text{ri}(\text{dom}(f))$.
- (b) Show that if $C \subset \text{ri}(\text{dom}(f))$ and $f(x) < \gamma$ for some $x \in \text{cl}(C)$, then $f(x) < \gamma$ for some $x \in \text{ri}(C)$.

- (c) Show that if $C \subset \text{dom}(f)$ and $f(x) \geq \gamma$ for all $x \in C$, then $f(x) \geq \gamma$ for all $x \in \text{cl}(C)$.

Solution: (a) Assume the contrary, i.e., that $f(x) \geq \gamma$ for all $x \in \text{ri}(\text{dom}(f))$. Let \bar{x} be such that $f(\bar{x}) < \gamma$ and let \tilde{x} be any vector in $\text{ri}(\text{dom}(f))$. By the Line Segment Principle, all the points on the line segment connecting \bar{x} and \tilde{x} , except possibly \bar{x} , belong to $\text{ri}(\text{dom}(f))$ and therefore,

$$f(\alpha\tilde{x} + (1 - \alpha)\bar{x}) \geq \gamma, \quad \forall \alpha \in (0, 1].$$

Thus, we have

$$\alpha f(\tilde{x}) + (1 - \alpha)f(\bar{x}) \geq f(\alpha\tilde{x} + (1 - \alpha)\bar{x}) \geq \gamma, \quad \forall \alpha \in (0, 1].$$

By letting $\alpha \rightarrow 0$, it follows that $f(\bar{x}) \geq \gamma$, a contradiction.

- (b) Define

$$g(x) = \begin{cases} f(x) & \text{if } x \in \text{cl}(C), \\ \infty & \text{otherwise.} \end{cases}$$

Then

$$\text{ri}(C) \subset \text{dom}(g) \subset \text{cl}(C),$$

so that $\text{ri}(\text{dom}(g)) = \text{ri}(C)$, by Prop. 1.3.5. By hypothesis, there is an \bar{x} with $g(\bar{x}) < \gamma$, so by part (a), there exists an $\tilde{x} \in \text{ri}(\text{dom}(g))$ with $g(\tilde{x}) < \gamma$. This vector belongs to $\text{ri}(C)$ and satisfies $f(\tilde{x}) < \alpha$.

- (c) Assume the contrary, i.e., that $f(x) < \gamma$ for some $x \in \text{cl}(C)$. Then, by part (b), we have $f(x) < \gamma$ for some $x \in \text{ri}(C)$, which contradicts the hypothesis.

1.26

Let C_1 and C_2 be two nonempty convex sets such that $C_1 \subset C_2$.

- Give an example showing that $\text{ri}(C_1)$ need not be a subset of $\text{ri}(C_2)$.
- Assuming that the sets C_1 and C_2 have the same affine hull, show that $\text{ri}(C_1) \subset \text{ri}(C_2)$.
- Assuming that the set $\text{ri}(C_1) \cap \text{ri}(C_2)$ is nonempty, show that $\text{ri}(C_1) \subset \text{ri}(C_2)$.
- Assuming that the set $C_1 \cap \text{ri}(C_2)$ is nonempty, show that the set $\text{ri}(C_1) \cap \text{ri}(C_2)$ is nonempty.

Solution: (a) Let C_1 be the segment $\{(x_1, x_2) \mid 0 \leq x_1 \leq 1, x_2 = 0\}$ and let C_2 be the box $\{(x_1, x_2) \mid 0 \leq x_1 \leq 1, 0 \leq x_2 \leq 1\}$. We have

$$\text{ri}(C_1) = \{(x_1, x_2) \mid 0 < x_1 < 1, x_2 = 0\},$$

$$\text{ri}(C_2) = \{(x_1, x_2) \mid 0 < x_1 < 1, 0 < x_2 < 1\}.$$

Thus $C_1 \subset C_2$, while $\text{ri}(C_1) \cap \text{ri}(C_2) = \emptyset$.

(b) Let $x \in \text{ri}(C_1)$, and consider a open ball B centered at x such that $B \cap \text{aff}(C_1) \subset C_1$. Since $\text{aff}(C_1) = \text{aff}(C_2)$ and $C_1 \subset C_2$, it follows that $B \cap \text{aff}(C_2) \subset C_2$, so $x \in \text{ri}(C_2)$. Hence $\text{ri}(C_1) \subset \text{ri}(C_2)$.

(c) Because $C_1 \subset C_2$, we have

$$\text{ri}(C_1) = \text{ri}(C_1 \cap C_2).$$

Since $\text{ri}(C_1) \cap \text{ri}(C_2) \neq \emptyset$, there holds

$$\text{ri}(C_1 \cap C_2) = \text{ri}(C_1) \cap \text{ri}(C_2)$$

(Prop. 1.3.8). Combining the preceding two relations, we obtain $\text{ri}(C_1) \subset \text{ri}(C_2)$.

(d) Let x_2 be in the intersection of C_1 and $\text{ri}(C_2)$, and let x_1 be in the relative interior of C_1 [$\text{ri}(C_1)$ is nonempty by Prop. 1.3.2]. If $x_1 = x_2$, then we are done, so assume that $x_1 \neq x_2$. By the Line Segment Principle, all the points on the line segment connecting x_1 and x_2 , except possibly x_2 , belong to the relative interior of C_1 . Since $C_1 \subset C_2$, the vector x_1 is in C_2 , so that by the Line Segment Principle, all the points on the line segment connecting x_1 and x_2 , except possibly x_1 , belong to the relative interior of C_2 . Hence, all the points on the line segment connecting x_1 and x_2 , except possibly x_1 and x_2 , belong to the intersection $\text{ri}(C_1) \cap \text{ri}(C_2)$, showing that $\text{ri}(C_1) \cap \text{ri}(C_2)$ is nonempty.

1.27

Let C be a nonempty set.

- (a) If C is convex and compact, and the origin is not in the relative boundary of C , then $\text{cone}(C)$ is closed.
- (b) Give examples showing that the assertion of part (a) fails if C is unbounded or the origin is in the relative boundary of C .
- (c) If C is compact and the origin is not in the relative boundary of $\text{conv}(C)$, then $\text{cone}(C)$ is closed. *Hint:* Use part (a) and Exercise 1.14(b).

Solution: (a) If $0 \in C$, then $0 \in \text{ri}(C)$ since 0 is not on the relative boundary of C . By Exercise 1.24(b), it follows that $\text{cone}(C)$ coincides with $\text{aff}(C)$, which is a closed set. If $0 \notin C$, let y be in the closure of $\text{cone}(C)$ and let $\{y_k\} \subset \text{cone}(C)$ be a sequence converging to y . By Exercise 1.13, for every y_k , there exists a nonnegative scalar α_k and a vector $x_k \in C$ such that $y_k = \alpha_k x_k$. Since $\{y_k\} \rightarrow y$, the sequence $\{y_k\}$ is bounded, implying that

$$\alpha_k \|x_k\| \leq \sup_{m \geq 0} \|y_m\| < \infty, \quad \forall k.$$

We have $\inf_{m \geq 0} \|x_m\| > 0$, since $\{x_k\} \subset C$ and C is a compact set not containing the origin, so that

$$0 \leq \alpha_k \leq \frac{\sup_{m \geq 0} \|y_m\|}{\inf_{m \geq 0} \|x_m\|} < \infty, \quad \forall k.$$

Thus, the sequence $\{(\alpha_k, x_k)\}$ is bounded and has a limit point (α, x) such that $\alpha \geq 0$ and $x \in C$. By taking a subsequence of $\{(\alpha_k, x_k)\}$ that converges to (α, x) , and by using the facts $y_k = \alpha_k x_k$ for all k and $\{y_k\} \rightarrow y$, we see that $y = \alpha x$ with $\alpha \geq 0$ and $x \in C$. Hence, $y \in \text{cone}(C)$, showing that $\text{cone}(C)$ is closed.

(b) To see that the assertion in part (a) fails when C is unbounded, let C be the line $\{(x_1, x_2) \mid x_1 = 1, x_2 \in \mathfrak{R}\}$ in \mathfrak{R}^2 not passing through the origin. Then, $\text{cone}(C)$ is the nonclosed set $\{(x_1, x_2) \mid x_1 > 0, x_2 \in \mathfrak{R}\} \cup \{(0, 0)\}$.

To see that the assertion in part (a) fails when C contains the origin on its relative boundary, let C be the closed ball $\{(x_1, x_2) \mid (x_1 - 1)^2 + x_2^2 \leq 1\}$ in \mathfrak{R}^2 . Then, $\text{cone}(C)$ is the nonclosed set $\{(x_1, x_2) \mid x_1 > 0, x_2 \in \mathfrak{R}\} \cup \{(0, 0)\}$ (see Fig. 1.3.2).

(c) Since C is compact, the convex hull of C is compact (cf. Prop. 1.2.2). Because $\text{conv}(C)$ does not contain the origin on its relative boundary, by part (a), the cone generated by $\text{conv}(C)$ is closed. By Exercise 1.14(b), $\text{cone}(\text{conv}(C))$ coincides with $\text{cone}(C)$ implying that $\text{cone}(C)$ is closed.

1.28 (Closure and Relative Interior of Cones)

- (a) Let C be a nonempty convex cone. Show that $\text{cl}(C)$ and $\text{ri}(C)$ is also a convex cone.
- (b) Let $C = \text{cone}(\{x_1, \dots, x_m\})$. Show that

$$\text{ri}(C) = \left\{ \sum_{i=1}^m \alpha_i x_i \mid \alpha_i > 0, i = 1, \dots, m \right\}.$$

Solution: (a) Let $x \in \text{cl}(C)$ and let α be a positive scalar. Then, there exists a sequence $\{x_k\} \subset C$ such that $x_k \rightarrow x$, and since C is a cone, $\alpha x_k \in C$ for all k . Furthermore, $\alpha x_k \rightarrow \alpha x$, implying that $\alpha x \in \text{cl}(C)$. Hence, $\text{cl}(C)$ is a cone, and it also convex since the closure of a convex set is convex.

By Prop. 1.3.2(a), the relative interior of a convex set is convex. To show that $\text{ri}(C)$ is a cone, let $x \in \text{ri}(C)$. Then, $x \in C$ and since C is a cone, $\alpha x \in C$ for all $\alpha > 0$. By the Line Segment Principle, all the points on the line segment connecting x and αx , except possibly αx , belong to $\text{ri}(C)$. Since this is true for every $\alpha > 0$, it follows that $\alpha x \in \text{ri}(C)$ for all $\alpha > 0$, showing that $\text{ri}(C)$ is a cone.

(b) Consider the linear transformation A that maps $(\alpha_1, \dots, \alpha_m) \in \mathfrak{R}^m$ into $\sum_{i=1}^m \alpha_i x_i \in \mathfrak{R}^n$. Note that C is the image of the nonempty convex set

$$\{(\alpha_1, \dots, \alpha_m) \mid \alpha_1 \geq 0, \dots, \alpha_m \geq 0\}$$

under A . Therefore, by using Prop. 1.3.6, we have

$$\begin{aligned}
\text{ri}(C) &= \text{ri}\left(A \cdot \{(\alpha_1, \dots, \alpha_m) \mid \alpha_1 \geq 0, \dots, \alpha_m \geq 0\}\right) \\
&= A \cdot \text{ri}\left(\{(\alpha_1, \dots, \alpha_m) \mid \alpha_1 \geq 0, \dots, \alpha_m \geq 0\}\right) \\
&= A \cdot \{(\alpha_1, \dots, \alpha_m) \mid \alpha_1 > 0, \dots, \alpha_m > 0\} \\
&= \left\{ \sum_{i=1}^m \alpha_i x_i \mid \alpha_1 > 0, \dots, \alpha_m > 0 \right\}.
\end{aligned}$$

1.29 (Closure and Relative Interior of Level Sets)

Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a proper convex function, and let $\gamma > \inf_{x \in \mathfrak{R}^n} f(x)$.

(a) Show that

$$\text{ri}(\{x \mid f(x) \leq \gamma\}) = \text{ri}(\{x \mid f(x) < \gamma\}) = \{x \in \text{ri}(\text{dom}(f)) \mid f(x) < \gamma\},$$

$$\text{cl}(\{x \mid f(x) \leq \gamma\}) = \text{cl}(\{x \mid f(x) < \gamma\}) = \{x \mid (\text{cl } f)(x) \leq \gamma\}.$$

(b) The sets $\{x \mid f(x) \leq \gamma\}$ and $\{x \mid f(x) < \gamma\}$ have the same dimension as $\text{dom}(f)$.

(c) If f is real-valued, $\{x \mid f(x) \leq \gamma\}$ has nonempty interior. Furthermore, for all $\gamma, \bar{\gamma}$ with $\inf_{x \in \mathfrak{R}^n} f(x) < \gamma < \bar{\gamma}$, the interior of $\{x \mid f(x) \leq \gamma\}$ is contained in the interior of $\{x \mid f(x) \leq \bar{\gamma}\}$.

Solution: We have for every $\gamma \in \mathfrak{R}$

$$\{(x, \gamma) \mid f(x) \leq \gamma\} = \text{epi}(f) \cap M, \quad (1.11)$$

where M is the set

$$M = \{(x, \gamma) \mid x \in \mathfrak{R}^n\}.$$

By Prop. 1.3.10,

$$\text{ri}(\text{epi}(f)) = \{(x, w) \mid x \in \text{ri}(\text{dom}(f)), f(x) < w\}, \quad (1.12)$$

so for $\gamma > \inf_{x \in \mathfrak{R}^n} f(x)$, using also Exercise 1.25(a), we have $\text{ri}(\text{epi}(f)) \cap M \neq \emptyset$. It follows from Eq. (1.11) and Prop. 1.3.8 that

$$\text{ri}(\{(x, \gamma) \mid f(x) \leq \gamma\}) = \text{ri}(\text{epi}(f)) \cap M,$$

$$\text{cl}(\{(x, \gamma) \mid f(x) \leq \gamma\}) = \text{cl}(\text{epi}(f)) \cap M.$$

The last two equations together with Eq. (1.12) show that for every $\gamma > \inf_{x \in \mathbb{R}^n} f(x)$, we have

$$\begin{aligned} \text{ri}\left(\{x \mid f(x) \leq \gamma\}\right) &= \{x \in \text{ri}(\text{dom}(f)) \mid f(x) < \gamma\}, \\ \text{cl}\left(\{x \mid f(x) \leq \gamma\}\right) &= \{x \mid (\text{cl } f)(x) \leq \gamma\}. \end{aligned}$$

Next, we show that

$$\text{cl}\left(\{x \mid f(x) \leq \gamma\}\right) = \text{cl}\left(\{x \mid f(x) < \gamma\}\right). \quad (1.13)$$

Clearly,

$$\text{cl}\left(\{x \mid f(x) \leq \gamma\}\right) \supset \text{cl}\left(\{x \mid f(x) < \gamma\}\right).$$

To show the reverse inclusion, let $\bar{x} \in \text{cl}\left(\{x \mid f(x) \leq \gamma\}\right)$, or equivalently, $(\text{cl } f)(\bar{x}) \leq \gamma$. Also, choose \tilde{x} such that $\tilde{x} \in \text{ri}(\text{dom}(f))$ and $f(\tilde{x}) < \gamma$ [such a vector exists by Exercise 1.25(a), in view of the assumption $\gamma > \inf_{x \in \mathbb{R}^n} f(x)$]. Then, by Prop. 1.3.15, along the line segment connecting \tilde{x} and \bar{x} , there is a sequence $\{x_k\} \subset \text{ri}(\text{dom}(f))$ that converges to \bar{x} and satisfies $f(x_k) < \gamma$ for all k . It follows that $\bar{x} \in \text{cl}\left(\{x \mid f(x) < \gamma\}\right)$, showing that

$$\text{cl}\left(\{x \mid f(x) \leq \gamma\}\right) \subset \text{cl}\left(\{x \mid f(x) < \gamma\}\right),$$

and thereby proving Eq. (1.13).

Next note that since the sets $\{x \mid f(x) \leq \gamma\}$ and $\{x \mid f(x) < \gamma\}$ have the same closure, by Prop. 1.3.5(c), they have the same relative interior, i.e.,

$$\text{ri}\left(\{x \mid f(x) \leq \gamma\}\right) = \text{ri}\left(\{x \mid f(x) < \gamma\}\right).$$

Finally, since the sets $\{x \mid f(x) \leq \gamma\}$ and $\{x \mid f(x) < \gamma\}$ have the same closure and relative interior, they also have the same affine hull, and hence the same dimension.

1.30 (Relative Interior Intersection Lemma)

Let C_1 and C_2 be convex sets. Show that

$$C_1 \cap \text{ri}(C_2) \neq \emptyset \quad \text{if and only if} \quad \text{ri}(C_1 \cap \text{aff}(C_2)) \cap \text{ri}(C_2) \neq \emptyset.$$

Hint: Choose $\bar{x} \in \text{ri}(C_1 \cap \text{aff}(C_2))$ and $x \in C_1 \cap \text{ri}(C_2)$ [which belongs to $C_1 \cap \text{aff}(C_2)$], consider the line segment connecting x and \bar{x} , and use the Line Segment Principle to conclude that points close to x belong to $\text{ri}(C_1 \cap \text{aff}(C_2)) \cap \text{ri}(C_2)$.

Solution: Let $x \in C_1 \cap \text{ri}(C_2)$ and $\bar{x} \in \text{ri}(C_1 \cap \text{aff}(C_2))$. Let L be the line segment connecting x and \bar{x} . Then L belongs to $C_1 \cap \text{aff}(C_2)$ since both of its endpoints belong to $C_1 \cap \text{aff}(C_2)$. Hence, by the Line Segment Principle, all points of L except possibly x , belong to $\text{ri}(C_1 \cap \text{aff}(C_2))$. On the other hand, by the definition of relative interior, all points of L that are sufficiently close to x belong to $\text{ri}(C_2)$, and these points, except possibly for x belong to $\text{ri}(C_1 \cap \text{aff}(C_2)) \cap \text{ri}(C_2)$.

1.31 (Closedness of Finitely Generated Cones)

Let a_1, \dots, a_r be vectors of \mathbb{R}^n . Then the generated cone

$$C = \text{cone}(\{a_1, \dots, a_r\}) = \left\{ x \mid x = \sum_{j=1}^r \mu_j a_j, \mu_j \geq 0, j = 1, \dots, r \right\}$$

is closed. *Note and Hint:* One way to show this is by noting that C can be written as AX where A is the matrix with columns a_1, \dots, a_r and X is the polyhedral set of all (μ_1, \dots, μ_r) with $\mu_j \geq 0$ for all j . The result then follows from Prop. 1.4.13. The purpose of this exercise is to explore an alternative and more elementary method of proof. To this end, use induction on the number of vectors r . When $r = 1$, C is either $\{0\}$ (if $a_1 = 0$) or a halfline, and is therefore closed. Suppose, for some $r \geq 1$, all cones of the form

$$C_r = \left\{ x \mid x = \sum_{j=1}^r \mu_j a_j, \mu_j \geq 0 \right\},$$

are closed. Then, show that a cone of the form

$$C_{r+1} = \left\{ x \mid x = \sum_{j=1}^{r+1} \mu_j a_j, \mu_j \geq 0 \right\}$$

is also closed.

Solution: Without loss of generality, assume that $\|a_j\| = 1$ for all j . There are two cases: (i) The vectors $-a_1, \dots, -a_{r+1}$ belong to C_{r+1} , in which case C_{r+1} is the subspace spanned by a_1, \dots, a_{r+1} and is therefore closed, and (ii) The negative of one of the vectors, say $-a_{r+1}$, does not belong to C_{r+1} . In this case, consider the cone C_r , which is closed by the induction hypothesis. Let

$$m = \min_{x \in C_r, \|x\|=1} a'_{r+1} x.$$

Since, the set $\{x \in C_r \mid \|x\| = 1\}$ is nonempty and compact, the minimum above is attained at some x^* by Weierstrass' theorem. We have, using the Schwartz inequality,

$$m = a'_{r+1} x^* \geq -\|a_{r+1}\| \cdot \|x^*\| = -1,$$

with equality if and only if $x^* = -a_{r+1}$. It follows that

$$m > -1,$$

since otherwise we would have $x^* = -a_{r+1}$, which violates the hypothesis $(-a_{r+1}) \notin C_r$. Let $\{x_k\}$ be a convergent sequence in C_{r+1} . We will prove that its limit belongs to C_{r+1} , thereby showing that C_{r+1} is closed. Indeed, for all k , we have

$x_k = \xi_k a_{r+1} + y_k$, where $\xi_k \geq 0$ and $y_k \in C_r$. Using the fact $\|a_{r+1}\| = 1$, we obtain

$$\begin{aligned}\|x_k\|^2 &= \xi_k^2 + \|y_k\|^2 + 2\xi_k a'_{r+1} y_k \\ &\geq \xi_k^2 + \|y_k\|^2 + 2m\xi_k \|y_k\| \\ &= (\xi_k - \|y_k\|)^2 + 2(1+m)\xi_k \|y_k\|.\end{aligned}$$

Since $\{x_k\}$ converges, $\xi_k \geq 0$, and $1+m > 0$, it follows that the sequences $\{\xi_k\}$ and $\{y_k\}$ are bounded and hence, they have limit points denoted by ξ and y , respectively. The limit of $\{x_k\}$ is

$$\lim_{k \rightarrow \infty} (\xi_k a_{r+1} + y_k) = \xi a_{r+1} + y,$$

which belongs to C_{r+1} , since $\xi \geq 0$ and $y \in C_r$ (by the closure hypothesis on C_r). We conclude that C_{r+1} is closed, completing the proof.

1.32 (Improper Convex Functions)

Let $f : \mathfrak{R}^n \mapsto [-\infty, \infty]$ be a convex function with $\text{dom}(f) \neq \emptyset$.

(a) Show that if f is improper, then

$$f(x) = -\infty, \quad \forall x \in \text{ri}(\text{dom}(f)).$$

Furthermore,

$$(\text{cl } f)(x) = \begin{cases} -\infty & \text{if } x \in \text{cl}(\text{dom}(f)), \\ \infty & \text{otherwise.} \end{cases}$$

(b) Show that if $f(x) < \infty$ for all $x \in \mathfrak{R}^n$, then either $f(x) = -\infty$ for all $x \in \mathfrak{R}^n$ or $f(x) > -\infty$ for all $x \in \mathfrak{R}^n$.

Solution: (a) Since f is improper, there exists some $\bar{x} \in \text{dom}(f)$ such that $f(\bar{x}) = -\infty$. Let $x \in \text{ri}(\text{dom}(f))$. Then by the Prolongation Principle [Prop. 1.3.1(c)], there is a vector $y \in \text{ri}(\text{dom}(f))$ such that $y \neq x$ and x lies in the line segment connecting y and \bar{x} . Thus, for some $\alpha \in (0, 1)$, we have $x = \alpha y + (1-\alpha)\bar{x}$, so by convexity of f ,

$$f(x) \leq \alpha f(y) + (1-\alpha)f(\bar{x}).$$

Since $f(y) < \infty$ and $f(\bar{x}) = -\infty$, it follows that $f(x) = -\infty$.

We next note that if $x \notin \text{cl}(\text{dom}(f))$, then $(x, w) \notin \text{cl}(\text{epi}(f))$ for all $w \in \mathfrak{R}$, so that $(\text{cl } f)(x) = \infty$.

We finally show that $(\text{cl } f)(x) = -\infty$ for all $x \in \text{cl}(\text{dom}(f))$. Assume, to arrive at a contradiction, that for some $x \in \text{cl}(\text{dom}(f))$, we have $(\text{cl } f)(x) > -\infty$. By the Line Segment Principle, there exists a sequence $\{x_k\} \subset \text{ri}(\text{dom}(f))$ that converges to x . Since by part (a), we have $f(x_k) = -\infty$, we have that $(x_k, w) \in \text{epi}(f)$ for every k and $w \in \mathfrak{R}$. It follows that

$$(x, w) \in \text{cl}(\text{epi}(f)) = \text{epi}(\text{cl } f), \quad \forall w \in \mathfrak{R}.$$

This implies that $(\text{cl } f)(x) = -\infty$ for all $x \in \text{cl}(\text{dom}(f))$.

(b) We have $\text{dom}(f) = \mathfrak{R}^n$, so either f is improper, in which case by part (a) we have $f(x) = -\infty$ for all $x \in \mathfrak{R}^n$, or f is proper, in which case we have $f(x) > -\infty$ for all $x \in \mathfrak{R}^n$.

1.33 (Lipschitz Continuity of Convex Functions)

Let $f : \mathbb{R}^n \mapsto \mathbb{R}$ be a convex function and X be a bounded set in \mathbb{R}^n . Show that f is Lipschitz continuous over X , i.e., there exists a positive scalar L such that

$$|f(x) - f(y)| \leq L\|x - y\|, \quad \forall x, y \in X.$$

Note: This result is also shown with a different proof in Section 5.4, using the theory of subgradients.

Solution: Let ϵ be a positive scalar and let C_ϵ be the set given by

$$C_\epsilon = \left\{ z \mid \|z - x\| \leq \epsilon, \text{ for some } x \in \text{cl}(X) \right\}.$$

We claim that the set C_ϵ is compact. Indeed, since X is bounded, so is its closure, which implies that $\|z\| \leq \max_{x \in \text{cl}(X)} \|x\| + \epsilon$ for all $z \in C_\epsilon$, showing that C_ϵ is bounded. To show the closedness of C_ϵ , let $\{z_k\}$ be a sequence in C_ϵ converging to some z . By the definition of C_ϵ , there is a corresponding sequence $\{x_k\}$ in $\text{cl}(X)$ such that

$$\|z_k - x_k\| \leq \epsilon, \quad \forall k. \quad (1.14)$$

Because $\text{cl}(X)$ is compact, $\{x_k\}$ has a subsequence converging to some $x \in \text{cl}(X)$. Without loss of generality, we may assume that $\{x_k\}$ converges to $x \in \text{cl}(X)$. By taking the limit in Eq. (1.14) as $k \rightarrow \infty$, we obtain $\|z - x\| \leq \epsilon$ with $x \in \text{cl}(X)$, showing that $z \in C_\epsilon$. Hence, C_ϵ is closed.

We now show that f has the Lipschitz property over X . Let x and y be two distinct points in X . Then, by the definition of C_ϵ , the point

$$z = y + \frac{\epsilon}{\|y - x\|}(y - x)$$

is in C_ϵ . Thus

$$y = \frac{\|y - x\|}{\|y - x\| + \epsilon}z + \frac{\epsilon}{\|y - x\| + \epsilon}x,$$

showing that y is a convex combination of $z \in C_\epsilon$ and $x \in C_\epsilon$. By convexity of f , we have

$$f(y) \leq \frac{\|y - x\|}{\|y - x\| + \epsilon}f(z) + \frac{\epsilon}{\|y - x\| + \epsilon}f(x),$$

implying that

$$f(y) - f(x) \leq \frac{\|y - x\|}{\|y - x\| + \epsilon}(f(z) - f(x)) \leq \frac{\|y - x\|}{\epsilon} \left(\max_{u \in C_\epsilon} f(u) - \min_{v \in C_\epsilon} f(v) \right),$$

where in the last inequality we use Weierstrass' theorem (f is continuous over \mathbb{R}^n and C_ϵ is compact). By switching the roles of x and y , we similarly obtain

$$f(x) - f(y) \leq \frac{\|x - y\|}{\epsilon} \left(\max_{u \in C_\epsilon} f(u) - \min_{v \in C_\epsilon} f(v) \right),$$

which combined with the preceding relation yields $|f(x) - f(y)| \leq L\|x - y\|$, where

$$L = \left(\max_{u \in C_\epsilon} f(u) - \min_{v \in C_\epsilon} f(v) \right) / \epsilon.$$

1.34 (Uniform Approximation Lemma)

Let C be a convex and compact set, and let $\{f_i \mid i \in I\}$ be a family of convex functions $f_i : C \mapsto \Re$ such that

$$\sup_{i \in I} f_i(x) = 0, \quad \forall x \in C.$$

Then for every $\epsilon > 0$ there exists an index $\bar{i} \in I$ such that

$$-\epsilon \leq f_{\bar{i}}(x) \leq 0, \quad \forall x \in C.$$

Hint: Let \hat{x} be a point in the relative interior of C . For any $x \in C$ with $x \neq \hat{x}$, consider the line that starts at x and passes through \hat{x} , and let $r(x)$ be the point at which it meets D , the relative boundary of C . Choose an index \bar{i} such that

$$f_{\bar{i}}(\hat{x}) \geq - \left(\frac{\max_{x \in C} \|x - \hat{x}\|}{\min_{x \in D} \|x - \hat{x}\|} + 1 \right)^{-1} \epsilon.$$

From the paper: Yu, H., and Bertsekas, D. P., "On Near-Optimality of the Set of Finite-State Controllers for Average Cost POMDP," Mathematics of Operations Research, Vol. 33, pp. 1-11, 2008.

Solution: Let \hat{x} be a point in the relative interior of C . For any $x \in C$ with $x \neq \hat{x}$, consider the line that starts at x and passes through \hat{x} , and let $r(x)$ be the point at which it meets D , the relative boundary of C (D is the set of points in C that are not relative interior points of C). Choose an index \bar{i} such that

$$0 \geq f_{\bar{i}}(\hat{x}) \geq - \left(\frac{\max_{x \in C} \|x - \hat{x}\|}{\min_{x \in D} \|\hat{x} - x\|} + 1 \right)^{-1} \epsilon. \quad (1.15)$$

Using the convexity and nonpositivity of $f_{\bar{i}}$, we have

$$\begin{aligned} f_{\bar{i}}(\hat{x}) &\leq \frac{\|\hat{x} - r(x)\|}{\|x - \hat{x}\| + \|\hat{x} - r(x)\|} f_{\bar{i}}(x) + \frac{\|x - \hat{x}\|}{\|x - \hat{x}\| + \|\hat{x} - r(x)\|} f_{\bar{i}}(r(x)) \\ &\leq \frac{\|\hat{x} - r(x)\|}{\|x - \hat{x}\| + \|\hat{x} - r(x)\|} f_{\bar{i}}(x). \end{aligned}$$

From this relation and Eq. (1.15), we obtain for all $x \in C$ with $x \neq \hat{x}$

$$\begin{aligned} f_{\bar{i}}(x) &\geq \frac{\|x - \hat{x}\| + \|\hat{x} - r(x)\|}{\|\hat{x} - r(x)\|} f_{\bar{i}}(\hat{x}) \\ &= \left(\frac{\|x - \hat{x}\|}{\|\hat{x} - r(x)\|} + 1 \right) f_{\bar{i}}(\hat{x}) \\ &\geq \left(\frac{\max_{x \in C} \|x - \hat{x}\|}{\min_{x \in D} \|\hat{x} - x\|} + 1 \right) f_{\bar{i}}(\hat{x}) \\ &\geq -\epsilon. \end{aligned}$$

SECTION 1.4: Recession Cones

1.35 (Recession Cones of Nonclosed Sets)

Let C be a nonempty convex set.

(a) Show that

$$R_C \subset R_{\text{cl}(C)}, \quad \text{cl}(R_C) \subset R_{\text{cl}(C)}.$$

Give an example where the inclusion $\text{cl}(R_C) \subset R_{\text{cl}(C)}$ is strict.

(b) Let \overline{C} be a closed convex set such that $C \subset \overline{C}$. Show that $R_C \subset R_{\overline{C}}$. Give an example showing that the inclusion can fail if \overline{C} is not closed.

Solution: (a) Let $y \in R_C$. Then, by the definition of R_C , $x + \alpha y \in C$ for every $x \in C$ and every $\alpha \geq 0$. Since $C \subset \text{cl}(C)$, it follows that $x + \alpha y \in \text{cl}(C)$ for some $x \in \text{cl}(C)$ and every $\alpha \geq 0$, which, in view of part (b) of the Recession Cone Theorem (cf. Prop. 1.4.1), implies that $y \in R_{\text{cl}(C)}$. Hence

$$R_C \subset R_{\text{cl}(C)}.$$

By taking closures in this relation and by using the fact that $R_{\text{cl}(C)}$ is closed [part (a) of the Recession Cone Theorem], we obtain $\text{cl}(R_C) \subset R_{\text{cl}(C)}$.

To see that the inclusion $\text{cl}(R_C) \subset R_{\text{cl}(C)}$ can be strict, consider the set

$$C = \{(x_1, x_2) \mid 0 \leq x_1, 0 \leq x_2 < 1\} \cup \{(0, 1)\},$$

whose closure is

$$\text{cl}(C) = \{(x_1, x_2) \mid 0 \leq x_1, 0 \leq x_2 \leq 1\}.$$

The recession cones of C and its closure are

$$R_C = \{(0, 0)\}, \quad R_{\text{cl}(C)} = \{(x_1, x_2) \mid 0 \leq x_1, x_2 = 0\}.$$

Thus, $\text{cl}(R_C) = \{(0, 0)\}$, and $\text{cl}(R_C)$ is a strict subset of $R_{\text{cl}(C)}$.

(b) Let $y \in R_C$ and let x be a vector in C . Then we have $x + \alpha y \in C$ for all $\alpha \geq 0$. Thus for the vector x , which belongs to \overline{C} , we have $x + \alpha y \in \overline{C}$ for all $\alpha \geq 0$, and it follows from part (b) of the Recession Cone Theorem (cf. Prop. 1.4.1) that $y \in R_{\overline{C}}$. Hence, $R_C \subset R_{\overline{C}}$.

To see that the inclusion $R_C \subset R_{\overline{C}}$ can fail when \overline{C} is not closed, consider the sets

$$C = \{(x_1, x_2) \mid x_1 \geq 0, x_2 = 0\}, \quad \overline{C} = \{(x_1, x_2) \mid x_1 \geq 0, 0 \leq x_2 < 1\}.$$

Their recession cones are

$$R_C = C = \{(x_1, x_2) \mid x_1 \geq 0, x_2 = 0\}, \quad R_{\overline{C}} = \{(0, 0)\},$$

showing that R_C is not a subset of $R_{\overline{C}}$.

1.36 (Recession Cones of Relative Interiors)

Let C be a nonempty convex set.

- (a) Show that $R_{\text{ri}(C)} = R_{\text{cl}(C)}$.
- (b) Show that a vector y belongs to $R_{\text{ri}(C)}$ if and only if there exists a vector $x \in \text{ri}(C)$ such that $x + \alpha y \in \text{ri}(C)$ for every $\alpha \geq 0$.
- (c) Let \bar{C} be a convex set such that $\bar{C} = \text{ri}(\bar{C})$ and $C \subset \bar{C}$. Show that $R_C \subset R_{\bar{C}}$. Give an example showing that the inclusion can fail if $\bar{C} \neq \text{ri}(\bar{C})$.

Solution: (a) The inclusion $R_{\text{ri}(C)} \subset R_{\text{cl}(C)}$ follows from Exercise 1.35(b). Conversely, let $y \in R_{\text{cl}(C)}$, so that by the definition of $R_{\text{cl}(C)}$, $x + \alpha y \in \text{cl}(C)$ for every $x \in \text{cl}(C)$ and every $\alpha \geq 0$. In particular, $x + \alpha y \in \text{cl}(C)$ for every $x \in \text{ri}(C)$ and every $\alpha \geq 0$. By the Line Segment Principle, all points on the line segment connecting x and $x + \alpha y$, except possibly $x + \alpha y$, belong to $\text{ri}(C)$, implying that $x + \alpha y \in \text{ri}(C)$ for every $x \in \text{ri}(C)$ and every $\alpha \geq 0$. Hence, $y \in R_{\text{ri}(C)}$, showing that $R_{\text{cl}(C)} \subset R_{\text{ri}(C)}$.

(b) If $y \in R_{\text{ri}(C)}$, then by the definition of $R_{\text{ri}(C)}$ for every vector $x \in \text{ri}(C)$ and $\alpha \geq 0$, the vector $x + \alpha y$ is in $\text{ri}(C)$, which holds in particular for some $x \in \text{ri}(C)$ [note that $\text{ri}(C)$ is nonempty by Prop. 1.3.1(b)].

Conversely, let y be such that there exists a vector $x \in \text{ri}(C)$ with $x + \alpha y \in \text{ri}(C)$ for all $\alpha \geq 0$. Hence, there exists a vector $x \in \text{cl}(C)$ with $x + \alpha y \in \text{cl}(C)$ for all $\alpha \geq 0$, which, by part (b) of the Recession Cone Theorem (cf. Prop. 1.4.1), implies that $y \in R_{\text{cl}(C)}$. Using part (a), it follows that $y \in R_{\text{ri}(C)}$, completing the proof.

(c) Using Exercise 1.35(c) and the assumption that $C \subset \bar{C}$ [which implies that $C \subset \overline{\text{cl}(C)}$], we have

$$R_C \subset R_{\text{cl}(\bar{C})} = R_{\text{ri}(\bar{C})} = R_{\bar{C}},$$

where the equalities follow from part (a) and the assumption that $\bar{C} = \text{ri}(\bar{C})$.

To see that the inclusion $R_C \subset R_{\bar{C}}$ can fail when $\bar{C} \neq \text{ri}(\bar{C})$, consider the sets

$$C = \{(x_1, x_2) \mid x_1 \geq 0, 0 < x_2 < 1\}, \quad \bar{C} = \{(x_1, x_2) \mid x_1 \geq 0, 0 \leq x_2 < 1\},$$

for which we have $C \subset \bar{C}$ and

$$R_C = \{(x_1, x_2) \mid x_1 \geq 0, x_2 = 0\}, \quad R_{\bar{C}} = \{(0, 0)\},$$

showing that R_C is not a subset of $R_{\bar{C}}$.

1.37 (Closure Under Linear Transformations)

Let C be a nonempty convex subset of \mathfrak{R}^n and let A be an $m \times n$ matrix. Show that if $R_{\text{cl}(C)} \cap N(A) = \{0\}$, then

$$\text{cl}(A \cdot C) = A \cdot \text{cl}(C), \quad A \cdot R_{\text{cl}(C)} = R_{A \cdot \text{cl}(C)}.$$

Give an example showing that $A \cdot R_{\text{cl}(C)}$ and $R_{A \cdot \text{cl}(C)}$ may differ when $R_{\text{cl}(C)} \cap N(A) \neq \{0\}$.

Solution: Let y be in the closure of $A \cdot C$. We will show that $y = Ax$ for some $x \in \text{cl}(C)$. For every $\epsilon > 0$, the set

$$C_\epsilon = \text{cl}(C) \cap \{x \mid \|y - Ax\| \leq \epsilon\}$$

is closed. Since $A \cdot C \subset A \cdot \text{cl}(C)$ and $y \in \text{cl}(A \cdot C)$, it follows that y is in the closure of $A \cdot \text{cl}(C)$, so that C_ϵ is nonempty for every $\epsilon > 0$. Furthermore, the recession cone of the set $\{x \mid \|Ax - y\| \leq \epsilon\}$ coincides with the null space $N(A)$, so that $R_{C_\epsilon} = R_{\text{cl}(C)} \cap N(A)$. By assumption we have $R_{\text{cl}(C)} \cap N(A) = \{0\}$, and by part (c) of the Recession Cone Theorem (cf. Prop. 1.4.1), it follows that C_ϵ is bounded for every $\epsilon > 0$. Now, since the sets C_ϵ are nested nonempty compact sets, their intersection $\bigcap_{\epsilon > 0} C_\epsilon$ is nonempty. For any x in this intersection, we have $x \in \text{cl}(C)$ and $Ax - y = 0$, showing that $y \in A \cdot \text{cl}(C)$. Hence, $\text{cl}(A \cdot C) \subset A \cdot \text{cl}(C)$. The converse $A \cdot \text{cl}(C) \subset \text{cl}(A \cdot C)$ is clear, since for any $x \in \text{cl}(C)$ and sequence $\{x_k\} \subset C$ converging to x , we have $Ax_k \rightarrow Ax$, showing that $Ax \in \text{cl}(A \cdot C)$. Therefore,

$$\text{cl}(A \cdot C) = A \cdot \text{cl}(C). \quad (1.16)$$

We now show that $A \cdot R_{\text{cl}(C)} = R_{A \cdot \text{cl}(C)}$. Let $y \in A \cdot R_{\text{cl}(C)}$. Then, there exists a vector $u \in R_{\text{cl}(C)}$ such that $Au = y$, and by the definition of $R_{\text{cl}(C)}$, there is a vector $x \in \text{cl}(C)$ such that $x + \alpha u \in \text{cl}(C)$ for every $\alpha \geq 0$. Therefore, $Ax + \alpha Au \in A \cdot \text{cl}(C)$ for every $\alpha \geq 0$, which, together with $Ax \in A \cdot \text{cl}(C)$ and $Au = y$, implies that y is a direction of recession of the closed set $A \cdot \text{cl}(C)$ [cf. Eq. (1.16)]. Hence, $A \cdot R_{\text{cl}(C)} \subset R_{A \cdot \text{cl}(C)}$.

Conversely, let $y \in R_{A \cdot \text{cl}(C)}$. We will show that $y \in A \cdot R_{\text{cl}(C)}$. This is true if $y = 0$, so assume that $y \neq 0$. By definition of direction of recession, there is a vector $z \in A \cdot \text{cl}(C)$ such that $z + \alpha y \in A \cdot \text{cl}(C)$ for every $\alpha \geq 0$. Let $x \in \text{cl}(C)$ be such that $Ax = z$, and for every positive integer k , let $x_k \in \text{cl}(C)$ be such that $Ax_k = z + ky$. Since $y \neq 0$, the sequence $\{Ax_k\}$ is unbounded, implying that $\{x_k\}$ is also unbounded (if $\{x_k\}$ were bounded, then $\{Ax_k\}$ would be bounded, a contradiction). Because $x_k \neq x$ for all k , we can define

$$u_k = \frac{x_k - x}{\|x_k - x\|}, \quad \forall k.$$

Let u be a limit point of $\{u_k\}$, and note that $u \neq 0$. It can be seen that u is a direction of recession of $\text{cl}(C)$ [this can be done similar to the proof of part (c) of the Recession Cone Theorem (cf. Prop. 1.4.1)]. By taking an appropriate subsequence if necessary, we may assume without loss of generality that $\lim_{k \rightarrow \infty} u_k = u$. Then, by the choices of u_k and x_k , we have

$$Au = \lim_{k \rightarrow \infty} Au_k = \lim_{k \rightarrow \infty} \frac{Ax_k - Ax}{\|x_k - x\|} = \lim_{k \rightarrow \infty} \frac{k}{\|x_k - x\|} y,$$

implying that $\lim_{k \rightarrow \infty} \frac{k}{\|x_k - x\|}$ exists. Denote this limit by λ . If $\lambda = 0$, then u is in the null space $N(A)$, implying that $u \in R_{\text{cl}(C)} \cap N(A)$. By the given condition

$R_{\text{cl}(C)} \cap N(A) = \{0\}$, we have $u = 0$ contradicting the fact $u \neq 0$. Thus, λ is positive and $Au = \lambda y$, so that $A(u/\lambda) = y$. Since $R_{\text{cl}(C)}$ is a cone [part (a) of the Recession Cone Theorem] and $u \in R_{\text{cl}(C)}$, the vector u/λ is in $R_{\text{cl}(C)}$, so that y belongs to $A \cdot R_{\text{cl}(C)}$. Hence, $R_{A \cdot \text{cl}(C)} \subset A \cdot R_{\text{cl}(C)}$, completing the proof.

As an example showing that $A \cdot R_{\text{cl}(C)}$ and $R_{A \cdot \text{cl}(C)}$ may differ when $R_{\text{cl}(C)} \cap N(A) \neq \{0\}$, consider the set

$$C = \{(x_1, x_2) \mid x_1 \in \mathfrak{R}, x_2 \geq x_1^2\},$$

and the linear transformation A that maps $(x_1, x_2) \in \mathfrak{R}^2$ into $x_1 \in \mathfrak{R}$. Then, C is closed and its recession cone is

$$R_C = \{(x_1, x_2) \mid x_1 = 0, x_2 \geq 0\},$$

so that $A \cdot R_C = \{0\}$, where 0 is scalar. On the other hand, $A \cdot C$ coincides with \mathfrak{R} , so that $R_{A \cdot C} = \mathfrak{R} \neq A \cdot R_C$.

1.38

Let C be a nonempty convex subset of \mathfrak{R}^n , and A be an $m \times n$ matrix. Show that if $R_{\text{cl}(C)} \cap N(A)$ is a subspace of the lineality space of $\text{cl}(C)$, then

$$\text{cl}(A \cdot C) = A \cdot \text{cl}(C), \quad A \cdot R_{\text{cl}(C)} = R_{A \cdot \text{cl}(C)}.$$

Note: This is a refinement of Exercise 1.37.

Solution: Let S be defined by

$$S = R_{\text{cl}(C)} \cap N(A),$$

and note that S is a subspace of $L_{\text{cl}(C)}$ by the given assumption. Then, by Lemma 1.4.4, we have

$$\text{cl}(C) = (\text{cl}(C) \cap S^\perp) + S,$$

so that the images of $\text{cl}(C)$ and $\text{cl}(C) \cap S^\perp$ under A coincide [since $S \subset N(A)$], i.e.,

$$A \cdot \text{cl}(C) = A \cdot (\text{cl}(C) \cap S^\perp). \quad (1.17)$$

Because $A \cdot C \subset A \cdot \text{cl}(C)$, we have

$$\text{cl}(A \cdot C) \subset \text{cl}(A \cdot \text{cl}(C)),$$

which in view of Eq. (1.17) gives

$$\text{cl}(A \cdot C) \subset \text{cl}\left(A \cdot (\text{cl}(C) \cap S^\perp)\right).$$

Define

$$\overline{C} = \text{cl}(C) \cap S^\perp$$

so that the preceding relation becomes

$$\text{cl}(A \cdot C) \subset \text{cl}(A \cdot \overline{C}). \quad (1.18)$$

The recession cone of \overline{C} is given by

$$R_{\overline{C}} = R_{\text{cl}(C)} \cap S^\perp, \quad (1.19)$$

[cf. part (e) of the Recession Cone Theorem, Prop. 1.4.1], for which, since $S = R_{\text{cl}(C)} \cap N(A)$, we have

$$R_{\overline{C}} \cap N(A) = S \cap S^\perp = \{0\}.$$

Therefore, by Prop. 1.4.13, the set $A \cdot \overline{C}$ is closed, implying that $\text{cl}(A \cdot \overline{C}) = A \cdot \overline{C}$. By the definition of \overline{C} , we have $A \cdot \overline{C} \subset A \cdot \text{cl}(C)$, implying that $\text{cl}(A \cdot \overline{C}) \subset A \cdot \text{cl}(C)$ which together with Eq. (1.18) yields $\text{cl}(A \cdot C) \subset A \cdot \text{cl}(C)$. The converse $A \cdot \text{cl}(C) \subset \text{cl}(A \cdot C)$ is clear, since for any $x \in \text{cl}(C)$ and sequence $\{x_k\} \subset C$ converging to x , we have $Ax_k \rightarrow Ax$, showing that $Ax \in \text{cl}(A \cdot C)$. Therefore,

$$\text{cl}(A \cdot C) = A \cdot \text{cl}(C). \quad (1.20)$$

We next show that $A \cdot R_{\text{cl}(C)} = R_{A \cdot \text{cl}(C)}$. Let $y \in A \cdot R_{\text{cl}(C)}$. Then, there exists a vector $u \in R_{\text{cl}(C)}$ such that $Au = y$, and by the definition of $R_{\text{cl}(C)}$, there is a vector $x \in \text{cl}(C)$ such that $x + \alpha u \in \text{cl}(C)$ for every $\alpha \geq 0$. Therefore, $Ax + \alpha Au \in A \cdot \text{cl}(C)$ for some $x \in \text{cl}(C)$ and for every $\alpha \geq 0$, which together with $Ax \in A \cdot \text{cl}(C)$ and $Au = y$ implies that y is a recession direction of the closed set $A \cdot \text{cl}(C)$ [Eq. (1.20)]. Hence, $A \cdot R_{\text{cl}(C)} \subset R_{A \cdot \text{cl}(C)}$.

Conversely, in view of Eq. (1.17) and the definition of \overline{C} , we have

$$R_{A \cdot \text{cl}(C)} = R_{A \cdot \overline{C}}.$$

Since $R_{\overline{C}} \cap N(A) = \{0\}$ and \overline{C} is closed, by Exercise 1.37, it follows that

$$R_{A \cdot \overline{C}} = A \cdot R_{\overline{C}},$$

which combined with Eq. (1.19) implies that

$$A \cdot R_{\overline{C}} \subset A \cdot R_{\text{cl}(C)}.$$

The preceding three relations yield $R_{A \cdot \text{cl}(C)} \subset A \cdot R_{\text{cl}(C)}$, completing the proof.

1.39 (Recession Cones of Vector Sums)

This exercise is an extension of Prop. 1.4.14 to nonclosed sets. Let C_1, \dots, C_m be nonempty convex subsets of \mathfrak{R}^n such that the equality $d_1 + \dots + d_m = 0$ with $d_i \in R_{\text{cl}(C_i)}$ implies that $d_i \in L_{\text{cl}(C_i)}$ for all i . Then

$$\text{cl}(C_1 + \dots + C_m) = \text{cl}(C_1) + \dots + \text{cl}(C_m),$$

$$R_{\text{cl}(C_1 + \dots + C_m)} = R_{\text{cl}(C_1)} + \dots + R_{\text{cl}(C_m)}.$$

Solution: Let C be the Cartesian product $C_1 \times \dots \times C_m$. Then,

$$\text{cl}(C) = \text{cl}(C_1) \times \dots \times \text{cl}(C_m), \quad (1.21)$$

and using Exercise 1.38, its recession cone and lineality space are given by

$$R_{\text{cl}(C)} = R_{\text{cl}(C_1)} \times \dots \times R_{\text{cl}(C_m)}, \quad (1.22)$$

$$L_{\text{cl}(C)} = L_{\text{cl}(C_1)} \times \dots \times L_{\text{cl}(C_m)}.$$

Let A be a linear transformation that maps $(x_1, \dots, x_m) \in \mathfrak{R}^{mn}$ into $x_1 + \dots + x_m \in \mathfrak{R}^n$. Then, the intersection $R_{\text{cl}(C)} \cap N(A)$ consists of points (y_1, \dots, y_m) such that $y_1 + \dots + y_m = 0$ with $y_i \in R_{\text{cl}(C_i)}$ for all i . By the given condition, every vector (y_1, \dots, y_m) in the intersection $R_{\text{cl}(C)} \cap N(A)$ is such that $y_i \in L_{\text{cl}(C_i)}$ for all i , implying that (y_1, \dots, y_m) belongs to the lineality space $L_{\text{cl}(C)}$. Thus, $R_{\text{cl}(C)} \cap N(A) \subset L_{\text{cl}(C)} \cap N(A)$. On the other hand by definition of the lineality space, we have $L_{\text{cl}(C)} \subset R_{\text{cl}(C)}$, so that $L_{\text{cl}(C)} \cap N(A) \subset R_{\text{cl}(C)} \cap N(A)$. Hence, $R_{\text{cl}(C)} \cap N(A) = L_{\text{cl}(C)} \cap N(A)$, implying that $R_{\text{cl}(C)} \cap N(A)$ is a subspace of $L_{\text{cl}(C)}$. By Exercise 1.38, we have $\text{cl}(A \cdot C) = A \cdot \text{cl}(C)$ and $R_{A \cdot \text{cl}(C)} = A \cdot R_{\text{cl}(C)}$, from which by using the relation $A \cdot C = C_1 + \dots + C_m$, and Eqs. (1.21) and (1.22), we obtain

$$\text{cl}(C_1 + \dots + C_m) = \text{cl}(C_1) + \dots + \text{cl}(C_m),$$

$$R_{\text{cl}(C_1 + \dots + C_m)} = R_{\text{cl}(C_1)} + \dots + R_{\text{cl}(C_m)}.$$

1.40 (Retractiveness of Convex Cones)

- Show that a nonpolyhedral closed convex cone need not be retractive, by using as an example the cone $C = \{(u, v, w) \mid \|(u, v)\| \leq w\}$, the recession direction $d = (1, 0, 1)$, and the corresponding asymptotic sequence $\{(k, \sqrt{k}, \sqrt{k^2 + k})\}$. (This is the, so-called, second order cone, which plays an important role in conic programming; see Chapter 5.)
- Verify that the cone C of part (a) can be written as the intersection of an infinite number of closed halfspaces, thereby showing that a nested set sequence obtained by intersection of an infinite number of retractive nested set sequences need not be retractive.

Solution: (a) Clearly, $d = (1, 0, 1)$ is the recession direction associated with the asymptotic sequence $\{x_k\}$, where $x_k = (k, \sqrt{k}, \sqrt{k^2 + k})$. On the other hand, it can be verified by straightforward calculation that the vector

$$x_k - d = (k - 1, \sqrt{k}, \sqrt{k^2 + k} - 1)$$

does not belong to C . Indeed, denoting

$$u_k = k - 1, \quad v_k = \sqrt{k}, \quad w_k = \sqrt{k^2 + k} - 1,$$

we have

$$\|(u_k, v_k)\|^2 = (k - 1)^2 + k = k^2 - k + 1,$$

while

$$w_k^2 = (\sqrt{k^2 + k} - 1)^2 = k^2 + k + 1 - 2\sqrt{k^2 + k},$$

and it can be seen that

$$\|(u_k, v_k)\|^2 > w_k^2, \quad \forall k \geq 1.$$

(b) Since by the Schwarz inequality, we have

$$\max_{\|(w, y)\|=1} (uw + vy) = \|(u, v)\|,$$

it follows that the cone

$$C = \{(u, v, w) \mid \|(u, v)\| \leq w\}$$

can be written as

$$C = \cap_{\|(w, y)\|=1} \{(u, v, w) \mid uw + vy \leq w\}.$$

Hence C is the intersection of an infinite number of closed halfspaces.

1.41 (Radon's Theorem)

Let x_1, \dots, x_m be vectors in \mathfrak{R}^n , where $m \geq n + 2$. Show that there exists a partition of the index set $\{1, \dots, m\}$ into two disjoint sets I and J such that

$$\text{conv}(\{x_i \mid i \in I\}) \cap \text{conv}(\{x_j \mid j \in J\}) \neq \emptyset.$$

As an illustration, show that given four points in the plane, either the (possibly degenerate) triangle formed by three of the points contains the fourth, or else the four points define a (possibly degenerate) quadrilateral. *Hint:* The system of $n + 1$ equations in the m unknowns $\lambda_1, \dots, \lambda_m$,

$$\sum_{i=1}^m \lambda_i x_i = 0, \quad \sum_{i=1}^m \lambda_i = 0,$$

has a nonzero solution λ^* . Let $I = \{i \mid \lambda_i^* \geq 0\}$ and $J = \{j \mid \lambda_j^* < 0\}$.

Solution: Consider the system of $n + 1$ equations in the m unknowns $\lambda_1, \dots, \lambda_m$

$$\sum_{i=1}^m \lambda_i x_i = 0, \quad \sum_{i=1}^m \lambda_i = 0.$$

Since $m > n + 1$, there exists a nonzero solution, call it λ^* . Let

$$I = \{i \mid \lambda_i^* \geq 0\}, \quad J = \{j \mid \lambda_j^* < 0\},$$

and note that I and J are nonempty, and that

$$\sum_{k \in I} \lambda_k^* = \sum_{k \in J} (-\lambda_k^*) > 0.$$

Consider the vector

$$x^* = \sum_{i \in I} \alpha_i x_i,$$

where

$$\alpha_i = \frac{\lambda_i^*}{\sum_{k \in I} \lambda_k^*}, \quad i \in I.$$

In view of the equations $\sum_{i=1}^m \lambda_i^* x_i = 0$ and $\sum_{i=1}^m \lambda_i^* = 0$, we also have

$$x^* = \sum_{j \in J} \alpha_j x_j,$$

where

$$\alpha_j = \frac{-\lambda_j^*}{\sum_{k \in J} (-\lambda_k^*)}, \quad j \in J.$$

It is seen that the scalars α_i and α_j are nonnegative, and that

$$\sum_{i \in I} \alpha_i = \sum_{j \in J} \alpha_j = 1,$$

so x^* belongs to the intersection

$$\text{conv}(\{x_i \mid i \in I\}) \cap \text{conv}(\{x_j \mid j \in J\}).$$

Given four distinct points in the plane (i.e., $m = 4$ and $n = 2$), Radon's Theorem guarantees the existence of a partition into two subsets, the convex hulls of which intersect. Assuming, there is no subset of three points lying on the same line, there are two possibilities:

- (1) Each set in the partition consists of two points, in which case the convex hulls intersect and define the diagonals of a quadrilateral.

- (2) One set in the partition consists of three points and the other consists of one point, in which case the triangle formed by the three points must contain the fourth.

In the case where three of the points define a line segment on which they lie, and the fourth does not, the triangle formed by the two ends of the line segment and the point outside the line segment form a triangle that contains the fourth point. In the case where all four of the points lie on a line segment, the degenerate triangle formed by three of the points, including the two ends of the line segment, contains the fourth point.

1.42 (Helly's Theorem [Hel21])

Consider a collection \mathcal{S} of convex sets in \mathfrak{R}^n , with at least $n + 1$ members, and assume that the intersection of every subcollection of $n + 1$ sets has nonempty intersection.

- (a) Assuming that \mathcal{S} is a finite collection, show that the entire collection has nonempty intersection. *Hint:* Use induction. Assume that the conclusion holds for every collection of M sets, where $M \geq n + 1$, and show that the conclusion holds for every collection of $M + 1$ sets. In particular, let C_1, \dots, C_{M+1} be a collection of $M + 1$ convex sets, and consider the collection of $M + 1$ sets B_1, \dots, B_{M+1} , where

$$B_j = \bigcap_{\substack{i=1, \dots, M+1 \\ i \neq j}} C_i, \quad j = 1, \dots, M + 1.$$

Note that, by the induction hypothesis, each set B_j is the intersection of a collection of M sets that have the property that every subcollection of $n + 1$ (or fewer) sets has nonempty intersection. Hence each set B_j is nonempty. Let x_j be a vector in B_j . Apply Radon's Theorem (Exercise 1.41) to the vectors x_1, \dots, x_{M+1} . Show that any vector in the intersection of the corresponding convex hulls belongs to the intersection of C_1, \dots, C_{M+1} .

- (b) Assuming that the members of \mathcal{S} are compact, show that the entire collection has nonempty intersection. *Hint:* Use part (a) and the fact that if a collection of compact sets has empty intersection, so does one of its finite subcollections [cf. Prop. A.2.4(i)].
- (c) Use part (a) to show that given a finite family of vertical intervals on the plane every three of which can be intersected by a line, the entire family can be intersected by the same line.

Solution: (a) Consider the induction argument of the hint, let B_j be defined as in the hint, and for each j , let x_j be a vector in B_j . Since $M + 1 \geq n + 2$, we can apply Radon's Theorem to the vectors x_1, \dots, x_{M+1} . Thus, there exist nonempty and disjoint index subsets I and J such that $I \cup J = \{1, \dots, M + 1\}$, nonnegative scalars $\alpha_1, \dots, \alpha_{M+1}$, and a vector x^* such that

$$x^* = \sum_{i \in I} \alpha_i x_i = \sum_{j \in J} \alpha_j x_j, \quad \sum_{i \in I} \alpha_i = \sum_{j \in J} \alpha_j = 1.$$

It can be seen that for every $i \in I$, a vector in B_i belongs to the intersection $\cap_{j \in J} C_j$. Therefore, since x^* is a convex combination of vectors in B_i , $i \in I$, x^* also belongs to the intersection $\cap_{j \in J} C_j$. Similarly, by reversing the role of I and J , we see that x^* belongs to the intersection $\cap_{i \in I} C_i$. Thus, x^* belongs to the intersection of the entire collection C_1, \dots, C_{M+1} .

(b) Evident from the hint.

(c) Consider a finite family $\{S_i \mid i = 1, \dots, m\}$ of vertical line segments on the plane:

$$S_i = \{(x, y) \mid x = x_i, \underline{y}_i \leq y \leq \bar{y}_i\}, \quad i = 1, \dots, m,$$

where $m \geq 3$, and $x_i, \underline{y}_i, \bar{y}_i$, $i = 1, \dots, m$, are given scalars. For each i , consider the set of lines that intersect S_i :

$$C_i = \{(a, b) \mid \underline{y}_i \leq ax_i + b \leq \bar{y}_i\}.$$

The sets C_i are convex, and every three of them have a common point. By Helly's Theorem, it follows that all the sets C_i have a common point, which is a line that intersects all the intervals S_i .

1.43 (Minimization of Max Functions)

Consider the minimization over \mathbb{R}^n of the function

$$\max\{f_1(x), \dots, f_M(x)\},$$

where $f_i : \mathbb{R}^n \mapsto (-\infty, \infty]$, $i = 1, \dots, M$, are convex functions, and assume that the optimal value, denoted f^* , is finite. Show that there exists a subset I of $\{1, \dots, M\}$, containing no more than $n + 1$ indices, such that

$$\inf_{x \in \mathbb{R}^n} \left\{ \max_{i \in I} f_i(x) \right\} = f^*.$$

Hint: Consider the convex sets $X_i = \{x \mid f_i(x) < f^*\}$, argue by contradiction, and apply Helly's Theorem (Exercise 1.42). *Note:* The result of this exercise relates to the following question: what is the minimal number of functions f_i that we need to include in the cost function $\max_i f_i(x)$ in order to attain the optimal value f^* ? According to the result, the number is no more than $n + 1$. For applications of this result in structural design and Chebyshev approximation, see Ben Tal and Nemirovski [BeN01].

Solution: Assume the contrary, i.e., that for every index set $I \subset \{1, \dots, M\}$, which contains no more than $n + 1$ indices, we have

$$\inf_{x \in \mathbb{R}^n} \left\{ \max_{i \in I} f_i(x) \right\} < f^*.$$

This means that for every such I , the intersection $\cap_{i \in I} X_i$ is nonempty, where

$$X_i = \{x \mid f_i(x) < f^*\}.$$

From Helly's Theorem, it follows that the entire collection $\{X_i \mid i = 1, \dots, M\}$ has nonempty intersection, thereby implying that

$$\inf_{x \in \mathbb{R}^n} \left\{ \max_{i=1, \dots, M} f_i(x) \right\} < f^*.$$

This contradicts the definition of f^* .

1.44 (Set Intersections and Helly's Theorem)

Show that the conclusions of Prop. 1.4.11(b) hold if the assumption that the sets C_k are nonempty and nested is replaced by the weaker assumption that any subcollection of $n + 1$ (or fewer) sets from the sequence $\{C_k\}$ has nonempty intersection. *Hint:* Consider the sets \overline{C}_k given by

$$\overline{C}_k = \bigcap_{i=0}^k C_i, \quad k = 1, 2, \dots,$$

and use Helly's Theorem [Exercise 1.42(a)] to show that they are nonempty.

Solution: Helly's Theorem implies that the sets \overline{C}_k defined in the hint are nonempty. These sets are also nested and satisfy the assumptions of Prop. 1.4.11(b). Therefore, the intersection $\bigcap_{i=1}^{\infty} \overline{C}_i$ is nonempty. Since

$$\bigcap_{i=1}^{\infty} \overline{C}_i \subset \bigcap_{i=1}^{\infty} C_i,$$

the result follows.

1.45 (Kirchberger's Theorem [Kir1903])

Let S be a finite subset of \mathbb{R}^n with at least $n + 2$ points, and let $S = B \cup R$ be a partition of S in two disjoint subsets B (the "blue" points) and R (the "red" points). Suppose that every subset \overline{S} of $n + 2$ points of S can be linearly separated, in the sense that there is a vector \overline{a} and a scalar \overline{c} such that $\overline{a}'b + \overline{c} < 0$ for all $b \in \overline{S} \cap B$ and $\overline{a}'r + \overline{c} > 0$ for all $r \in \overline{S} \cap R$. Use Helly's Theorem (Exercise 1.42) to show that the entire set S can be linearly separated, i.e., that there is a vector a and a scalar c such that $a'b + c < 0$ for all $b \in B$ and $a'r + c > 0$ for all $r \in R$. *Hint:* For each $b \in B$ consider the set $G(b)$ of vectors (x_1, \dots, x_{n+1}) such that

$$\sum_{i=1}^n x_i b_i + x_{n+1} < 0,$$

and for each $r \in R$, consider the set $H(r)$ of vectors (x_1, \dots, x_{n+1}) such that

$$\sum_{i=1}^n x_i r_i + x_{n+1} > 0.$$

Let \mathcal{C} be the collection of the sets $G(b)$ and $H(r)$ as b and r ranges over B and R , respectively. Use Helly's Theorem (Exercise 1.42) to show that \mathcal{C} has nonempty intersection.

Solution: For each $b \in B$ consider the set $G(b)$ of vectors (x_1, \dots, x_{n+1}) such that

$$\sum_{i=1}^n x_i b_i + x_{n+1} < 0,$$

and for each $r \in R$, consider the set $H(r)$ of vectors (x_1, \dots, x_{n+1}) such that

$$\sum_{i=1}^n x_i r_i + x_{n+1} > 0.$$

Let \mathcal{C} be the collection of the convex sets $G(b)$ and $H(r)$ as b and r ranges over B and R , respectively. By assumption, for any subset $C \subset B \cup R$, consisting of $n+2$ points, the sets $B \cap C$ and $R \cap C$ can be linearly separated, so there exist $\bar{a} \in \mathfrak{R}^n$ and $\bar{c} \in \mathfrak{R}$ such that

$$\begin{aligned} \bar{a}'b + \bar{c} &< 0, & \forall b \in B \cap C, \\ \bar{a}'r + \bar{c} &> 0, & \forall r \in R \cap C. \end{aligned}$$

Thus, $(\bar{a}, \bar{c}) \in L(b)$ for all $b \in B \cap C$, and $(\bar{a}, \bar{c}) \in G(r)$ for all $r \in R \cap C$. It follows that \mathcal{C} is a finite family of convex sets in \mathfrak{R}^{n+1} , which contains at least $n+2$ members and every collection of $n+2$ of these members has nonempty intersection. By Helly's Theorem, there is a vector (a, c) that belongs to all members of \mathcal{C} , and for which we have $a'x + c < 0$ for all $x \in B$ and $a'x + c > 0$ for all $x \in R$. (Proof given in Webster [Web94], and credited to H. Rademacher and I. J. Schoenberg, "Helly's Theorem on Convex Domains and Tchebycheff's Approximation Problem," Canadian J. of Math., Vol. 2, 1950, pp. 245-256.)

1.46 (Krasnoselsky's Theorem [Kra46])

Let S be a nonempty compact subset of \mathfrak{R}^n . For any two points x and y of S , we say that x is *visible from* y if the line segment connecting x and y belongs to S . Assume that S has the property that for any subset of $n+1$ points of S , there is a point of S from which all $n+1$ points are visible. Show that there is a point in S from which all points of S are visible. *Hint:* For each $y \in S$, let S_y be the set of points of S that are visible from y . Show that the set $C_y = \text{conv}(S_y)$ is compact, and consider the family of sets $\{C_y \mid y \in S\}$. Use Helly's Theorem (Exercise 1.42) to show that there is a vector $a \in S$ that belongs to $\bigcap_{y \in S} \text{conv}(S_y)$. Show that $a \in \bigcap_{y \in S} S_y$ (this last part is not simple).

Solution: For each $y \in S$, let S_y be the set of points of S that are visible from y . The set S_y is easily seen to be closed, and hence its convex hull, $C_y = \text{conv}(S_y)$, is compact by Prop. 1.2.2. Consider the family of sets $\{C_y \mid y \in S\}$. Let y_0, \dots, y_n be points in S . By the hypothesis, there is a vector $x \in S$ from which y_0, \dots, y_n are visible. Thus $x \in S_{y_0} \cap \dots \cap S_{y_n}$, and hence also $x \in C_{y_0} \cap \dots \cap C_{y_n}$. It follows that any subcollection of $n+1$ sets from the family $\{C_y \mid y \in S\}$ is nonempty. By Helly's Theorem [Exercise 1.42(b)], the entire family is nonempty. Thus, there exists a vector a such that

$$a \in \text{conv}(S_y), \quad \forall y \in S.$$

We claim now that every $y \in S$ is visible from a . Assume the contrary, so there exists a vector $b \in S$ and a vector c in the line segment connecting a and b such that $c \notin S$. Let C be a closed ball of nonzero radius, which is centered at c and does not intersect S . Let

$$\alpha = \inf \{ \lambda \geq 0 \mid S \cap (C + \lambda(b - c)) \neq \emptyset \},$$

denote the closed ball $C + \alpha(b - c)$ by D and denote its center by d . Then by construction, S meets the boundary of D but not its interior. Let e a vector in $S \cap D$. We will show that $a \notin \text{conv}(S_e)$, thus arriving at a contradiction.

Indeed, consider the halfspaces

$$H^- = \{z \mid (z - e)'(e - d) < 0\}, \quad H^+ = \{z \mid (z - e)'(e - d) \geq 0\}.$$

Then, by elementary geometry, it follows that $a \in H^-$, while $S_e \subset H^+$ and hence also $\text{conv}(S_e) \subset H^+$. Since $H^- \cap H^+ = \emptyset$, it follows that $a \notin \text{conv}(S_e)$, a contradiction. (Proof given in Webster [Web94].)

SECTION 1.5: Hyperplanes

1.47

- (a) Let C_1 be a convex set with nonempty interior and C_2 be a nonempty convex set that does not intersect the interior of C_1 . Show that there exists a hyperplane such that one of the associated closed halfspaces contains C_2 , and does not intersect the interior of C_1 .
- (b) Show by an example that we cannot replace interior with relative interior in the statement of part (a).

Solution: (a) In view of the assumption that $\text{int}(C_1)$ and C_2 are disjoint and convex [cf Prop. 1.1.1(d)], it follows from the Separating Hyperplane Theorem that there exists a vector $a \neq 0$ such that

$$a'x_1 \leq a'x_2, \quad \forall x_1 \in \text{int}(C_1), \quad \forall x_2 \in C_2.$$

Let $b = \inf_{x_2 \in C_2} a'x_2$. Then, from the preceding relation, we have

$$a'x \leq b, \quad \forall x \in \text{int}(C_1). \tag{1.23}$$

We claim that the closed halfspace $\{x \mid a'x \geq b\}$, which contains C_2 , does not intersect $\text{int}(C_1)$.

Assume to arrive at a contradiction that there exists some $\bar{x}_1 \in \text{int}(C_1)$ such that $a'\bar{x}_1 \geq b$. Since $\bar{x}_1 \in \text{int}(C_1)$, we have that there exists some $\epsilon > 0$ such that $\bar{x}_1 + \epsilon a \in \text{int}(C_1)$, and

$$a'(\bar{x}_1 + \epsilon a) \geq b + \epsilon \|a\|^2 > b.$$

This contradicts Eq. (1.23). Hence, we have

$$\text{int}(C_1) \subset \{x \mid a'x < b\}.$$

(b) Consider the sets

$$\begin{aligned} C_1 &= \{(x_1, x_2) \mid x_1 = 0\}, \\ C_2 &= \{(x_1, x_2) \mid x_1 > 0, x_2x_1 \geq 1\}. \end{aligned}$$

These two sets are convex and C_2 is disjoint from $\text{ri}(C_1)$, which is equal to C_1 . The only separating hyperplane is the x_2 axis, which corresponds to having $a = (0, 1)$, as defined in part (a). For this example, there does not exist a closed halfspace that contains C_2 but is disjoint from $\text{ri}(C_1)$.

1.48

Let C be a nonempty convex set in \mathfrak{R}^n , and let M be a nonempty affine set in \mathfrak{R}^n . Show that $M \cap \text{ri}(C) = \emptyset$ is a necessary and sufficient condition for the existence of a hyperplane H containing M , and such that $\text{ri}(C)$ is contained in one of the open halfspaces associated with H .

Solution: If there exists a hyperplane H with the properties stated, then $M \cap C = \emptyset$, so the condition $M \cap \text{ri}(C) = \emptyset$ holds. Conversely, if $M \cap \text{ri}(C) = \emptyset$, then M and C can be properly separated by Prop. 1.5.6. This hyperplane can be chosen to contain M since M is affine. If this hyperplane contains a point in $\text{ri}(C)$, then it must contain all of C by Prop. 1.3.4. This contradicts the proper separation property, thus showing that $\text{ri}(C)$ is contained in one of the open halfspaces.

1.49

Let C_1 and C_2 be nonempty convex subsets of \mathfrak{R}^n such that C_2 is a cone.

- (a) Suppose that there exists a hyperplane that separates C_1 and C_2 properly. Show that there exists a hyperplane which separates C_1 and C_2 properly and passes through the origin.
- (b) Suppose that there exists a hyperplane that separates C_1 and C_2 strictly. Show that there exists a hyperplane that passes through the origin such that one of the associated closed halfspaces contains the cone C_2 and does not intersect C_1 .

Solution: (a) If C_1 and C_2 can be separated properly, we have from the Proper Separation Theorem that there exists a vector $a \neq 0$ such that

$$\inf_{x \in C_1} a'x \geq \sup_{x \in C_2} a'x, \quad (1.24)$$

$$\sup_{x \in C_1} a'x > \inf_{x \in C_2} a'x. \quad (1.25)$$

Let

$$b = \sup_{x \in C_2} a'x. \quad (1.26)$$

and consider the hyperplane

$$H = \{x \mid a'x = b\}.$$

Since C_2 is a cone, we have

$$\lambda a'x = a'(\lambda x) \leq b < \infty, \quad \forall x \in C_2, \forall \lambda > 0.$$

This relation implies that $a'x \leq 0$, for all $x \in C_2$, since otherwise it is possible to choose λ large enough and violate the above inequality for some $x \in C_2$. Hence, it follows from Eq. (1.26) that $b \leq 0$. Also, by letting $\lambda \rightarrow 0$ in the preceding relation, we see that $b \geq 0$. Therefore, we have that $b = 0$ and the hyperplane H contains the origin.

(b) If C_1 and C_2 can be separated strictly, we have by definition that there exists a vector $a \neq 0$ and a scalar β such that

$$a'x_2 < \beta < a'x_1, \quad \forall x_1 \in C_1, \quad \forall x_2 \in C_2. \quad (1.27)$$

We choose b to be

$$b = \sup_{x \in C_2} a'x, \quad (1.28)$$

and consider the closed halfspace

$$K = \{x \mid a'x \leq b\},$$

which contains C_2 . By Eq. (1.27), we have

$$b \leq \beta < a'x, \quad \forall x \in C_1,$$

so the closed halfspace K does not intersect C_1 .

Since C_2 is a cone, an argument similar to the one in part (a) shows that $b = 0$, and hence the hyperplane associated with the closed halfspace K passes through the origin, and has the desired properties.

1.50 (Separation Properties of Cones)

Define a *homogeneous halfspace* to be a closed halfspace associated with a hyperplane that passes through the origin. Show that:

- (a) A nonempty closed convex cone is the intersection of the homogeneous halfspaces that contain it.
- (b) The closure of the convex cone generated by a nonempty set X is the intersection of all the homogeneous halfspaces containing X .

Solution: (a) C is contained in the intersection of the homogeneous closed halfspaces that contain C , so we focus on proving the reverse inclusion. Let $x \notin C$. Since C is closed and convex by assumption, by using the Strict Separation Theorem, we see that the sets C and $\{x\}$ can be separated strictly. From Exercise 1.49(c), this implies that there exists a hyperplane that passes through the origin such that one of the associated closed halfspaces contains C , but is disjoint from x . Hence, if $x \notin C$, then x cannot belong to the intersection of the homogeneous closed halfspaces containing C , proving that C contains that intersection.

(b) A homogeneous halfspace is in particular a closed convex cone containing the origin, and such a cone includes X if and only if it includes $\text{cl}(\text{cone}(X))$. Hence, the intersection of all closed homogeneous halfspaces containing X and the intersection of all closed homogeneous halfspaces containing $\text{cl}(\text{cone}(X))$ coincide. From what has been proved in part(a), the latter intersection is equal to $\text{cl}(\text{cone}(X))$.

1.51 (Strong Separation)

Let C_1 and C_2 be nonempty convex subsets of \mathfrak{R}^n , and let B denote the unit ball in \mathfrak{R}^n , $B = \{x \mid \|x\| \leq 1\}$. A hyperplane H is said to *separate strongly* C_1 and C_2 if there exists an $\epsilon > 0$ such that $C_1 + \epsilon B$ is contained in one of the open halfspaces associated with H and $C_2 + \epsilon B$ is contained in the other. Show that:

- (a) The following three conditions are equivalent.
 - (i) There exists a hyperplane separating strongly C_1 and C_2 .
 - (ii) There exists a vector $a \in \mathfrak{R}^n$ such that $\inf_{x \in C_1} a'x > \sup_{x \in C_2} a'x$.
 - (iii) $\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 0$, i.e., $0 \notin \text{cl}(C_2 - C_1)$.
- (b) If C_1 and C_2 are disjoint, any one of the five conditions for strict separation, given in Prop. 1.5.3, implies that C_1 and C_2 can be strongly separated.

Solution: (a) We first show that (i) implies (ii). Suppose that C_1 and C_2 can be separated strongly. By definition, this implies that for some nonzero vector $a \in \mathfrak{R}^n$, $b \in \mathfrak{R}$, and $\epsilon > 0$, we have

$$C_1 + \epsilon B \subset \{x \mid a'x > b\},$$

$$C_2 + \epsilon B \subset \{x \mid a'x < b\},$$

where B denotes the closed unit ball. Since $a \neq 0$, we also have

$$\inf\{a'y \mid y \in B\} < 0, \quad \sup\{a'y \mid y \in B\} > 0.$$

Therefore, it follows from the preceding relations that

$$b \leq \inf\{a'x + \epsilon a'y \mid x \in C_1, y \in B\} < \inf\{a'x \mid x \in C_1\},$$

$$b \geq \sup\{a'x + \epsilon a'y \mid x \in C_2, y \in B\} > \sup\{a'x \mid x \in C_2\}.$$

Thus, there exists a vector $a \in \mathfrak{R}^n$ such that

$$\inf_{x \in C_1} a'x > \sup_{x \in C_2} a'x,$$

proving (ii).

Next, we show that (ii) implies (iii). Suppose that (ii) holds, i.e., there exists some vector $a \in \mathfrak{R}^n$ such that

$$\inf_{x \in C_1} a'x > \sup_{x \in C_2} a'x, \quad (1.29)$$

Using the Schwartz inequality, we see that

$$\begin{aligned} 0 &< \inf_{x \in C_1} a'x - \sup_{x \in C_2} a'x \\ &= \inf_{x_1 \in C_1, x_2 \in C_2} a'(x_1 - x_2), \\ &\leq \inf_{x_1 \in C_1, x_2 \in C_2} \|a\| \|x_1 - x_2\|. \end{aligned}$$

It follows that

$$\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 0,$$

thus proving (iii).

Finally, we show that (iii) implies (i). If (iii) holds, we have for some $\epsilon > 0$,

$$\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 2\epsilon > 0.$$

From this we obtain for all $x_1 \in C_1$, all $x_2 \in C_2$, and for all y_1, y_2 with $\|y_1\| \leq \epsilon$, $\|y_2\| \leq \epsilon$,

$$\|(x_1 + y_1) - (x_2 + y_2)\| \geq \|x_1 - x_2\| - \|y_1\| - \|y_2\| > 0,$$

which implies that $0 \notin (C_1 + \epsilon B) - (C_2 + \epsilon B)$. Therefore, the convex sets $C_1 + \epsilon B$ and $C_2 + \epsilon B$ are disjoint. By the Separating Hyperplane Theorem, we see that $C_1 + \epsilon B$ and $C_2 + \epsilon B$ can be separated, i.e., $C_1 + \epsilon B$ and $C_2 + \epsilon B$ lie in opposite closed halfspaces associated with the hyperplane that separates them. Then, the sets $C_1 + (\epsilon/2)B$ and $C_2 + (\epsilon/2)B$ lie in opposite open halfspaces, which by definition implies that C_1 and C_2 can be separated strongly.

(b) Since C_1 and C_2 are disjoint, we have $0 \notin (C_1 - C_2)$. Any one of conditions (2)-(5) of Prop. 1.5.3 imply condition (1) of that proposition (see the discussion in the proof of Prop. 1.5.3), which states that the set $C_1 - C_2$ is closed, i.e.,

$$\text{cl}(C_1 - C_2) = C_1 - C_2.$$

Hence, we have $0 \notin \text{cl}(C_1 - C_2)$, which implies that

$$\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 0.$$

From part (a), it follows that there exists a hyperplane separating C_1 and C_2 strongly.

1.52 (Characterization of Closed Convex Sets)

This exercise generalizes Prop. 1.5.4. Let C be a nonempty closed convex subset of \mathbb{R}^{n+1} . Show that if C contains no vertical lines, then C is the intersection of the closed halfspaces that contain it and correspond to nonvertical hyperplanes.

Solution: The set C is contained in the intersection of the closed halfspaces that contain C and correspond to nonvertical hyperplanes, so we focus on proving the reverse inclusion. Let $x \notin C$. Since by assumption C does not contain any vertical lines, by Prop. 1.5.8, there exists a closed halfspace that corresponds to a nonvertical hyperplane, contains C , but does not contain x . Hence, if $x \notin C$, then x cannot belong to the intersection of the closed halfspaces containing C and corresponding to nonvertical hyperplanes, proving that C contains that intersection.

SECTION 1.6: Conjugate Functions

1.53 (Logarithmic/Exponential Conjugacy)

Let $f : \mathbb{R} \mapsto \mathbb{R}$ be the exponential function

$$f(x) = e^x.$$

Show that the conjugate is

$$f^*(y) = \begin{cases} y \ln y - y & \text{if } y > 0, \\ 0 & \text{if } y = 0, \\ \infty & \text{if } y < 0. \end{cases}$$

Solution: The conjugate is

$$f^*(y) = \sup_{x \in \mathbb{R}} \{xy - e^x\}.$$

For $y < 0$, by taking $x \rightarrow -\infty$, we see that $xy - e^x$ can be made arbitrarily large, so $f^*(y) = \infty$. For $y = 0$, we have

$$f^*(0) = \sup_{x \in \mathbb{R}} \{-e^x\} = - \inf_{x \in \mathbb{R}} e^x = 0.$$

Finally, for $y > 0$, by setting the derivative of $xy - e^x$ to zero, we see that the supremum of $xy - e^x$ is obtained for $x = \ln y$, and by substitution, we obtain $f^*(y) = y \ln y - y$.

1.54 (Conjugates of p -Norms)

Let $f : \mathfrak{R}^n \mapsto \mathfrak{R}$ be the function

$$f(x) = \frac{1}{p} \sum_{i=1}^n |x_i|^p$$

where $1 < p$. Show that the conjugate is

$$f^*(y) = \frac{1}{q} \sum_{i=1}^n |y_i|^q,$$

where q is defined by the relation

$$\frac{1}{p} + \frac{1}{q} = 1.$$

Solution: Consider first the case $n = 1$. Let x and y be scalars. By setting the derivative of $xy - (1/p)|x|^p$ to zero, and we see that the supremum over x is attained when $\text{sgn}(x)|x|^{p-1} = y$, which implies that $xy = |x|^p$ and $|x|^{p-1} = |y|$. By substitution in the formula for the conjugate, we obtain

$$f^*(y) = |x|^p - \frac{1}{p}|x|^p = \left(1 - \frac{1}{p}\right)|x|^p = \frac{1}{q}|y|^{\frac{p}{p-1}} = \frac{1}{q}|y|^q. \quad (1.30)$$

We now note that for any function $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ that has the form

$$f(x) = f_1(x_1) + \cdots + f_n(x_n),$$

where $x = (x_1, \dots, x_n)$ and $f_i : \mathfrak{R} \mapsto (-\infty, \infty]$, $i = 1, \dots, n$, the conjugate is given by

$$f^*(y) = f_1^*(y_1) + \cdots + f_n^*(y_n),$$

where $f_i^* : \mathfrak{R} \mapsto (-\infty, \infty]$ is the conjugate of f_i , $i = 1, \dots, n$. By combining this fact with the formula (1.30), we obtain the desired result.

1.55 (Conjugate of a Quadratic)

Let

$$f(x) = \frac{1}{2}x'Qx + a'x + b,$$

where Q is a symmetric positive semidefinite $n \times n$ matrix, a is a vector in \mathfrak{R}^n , and b is a scalar. Derive the conjugate of f .

Solution: Let us assume first that Q is nonsingular. Then the maximum of $x'y - f(x)$ over x is attained when $Qx + a = \nabla f(x) = y$. By substitution, we obtain

$$f^*(y) = x'y - f(x) = x'(Qx + a) - \frac{1}{2}x'Qx - a'x - b = \frac{1}{2}x'Qx - b,$$

and finally, using $x = Q^{-1}(y - a)$,

$$f^*(y) = \frac{1}{2}(y - a)'Q^{-1}(y - a) - b.$$

Consider now the general case where Q may be singular. Then if the equation $y = Qx + a$ has no solution, i.e., $y - a$ does not belong to the range $R(Q)$ of Q , we have $f^*(y) = \infty$. Otherwise, let x be the solution of the equation $y = Qx + a$ that has minimum Euclidean norm. Then x is linearly related to $y - a$ and can be written as $x = Q^\dagger(y - a)$ where Q^\dagger is a symmetric positive semidefinite matrix (this is the *pseudoinverse* of Q ; see e.g., Luenberger, Optimization by Vector Space Methods, 1969, p. 165). Similar to the case where Q is invertible, we have $f^*(y) = (1/2)x'Qx - b$, and it follows, using $x = Q^\dagger(y - a)$, that

$$f^*(y) = \begin{cases} \frac{1}{2}(y - a)'Q^\dagger(y - a) - b & \text{if } y - a \in R(Q), \\ \infty & \text{otherwise.} \end{cases}$$

1.56

- (a) Show that if $f_1 : \mathbb{R}^n \mapsto (-\infty, \infty]$ and $f_2 : \mathbb{R}^n \mapsto (-\infty, \infty]$ are closed proper convex functions, with conjugates denoted by f_1^* and f_2^* , respectively, we have

$$f_1(x) \leq f_2(x), \quad \forall x \in \mathbb{R}^n,$$

if and only if

$$f_1^*(y) \geq f_2^*(y), \quad \forall y \in \mathbb{R}^n.$$

- (b) Show that if C_1 and C_2 are nonempty closed convex sets, we have

$$C_1 \subset C_2,$$

if and only if

$$\sigma_{C_1}(y) \leq \sigma_{C_2}(y), \quad \forall y \in \mathbb{R}^n.$$

Construct an example showing that closedness of C_1 and C_2 is a necessary assumption.

Solution: (a) If $f_1(x) \leq f_2(x)$ for all x , we have for all $y \in \mathbb{R}^n$,

$$f_1^*(y) = \sup_{x \in \mathbb{R}^n} \{x'y - f_1(x)\} \geq \sup_{x \in \mathbb{R}^n} \{x'y - f_2(x)\} = f_2^*(y).$$

The reverse implication follows from the fact that f_1 and f_2 are the conjugates of f_1^* and f_2^* , respectively.

- (b) Consider the indicator functions δ_{C_1} and δ_{C_2} of C_1 and C_2 . We have

$$C_1 \subset C_2 \quad \text{if and only if} \quad \delta_{C_1}(x) \geq \delta_{C_2}(x), \quad \forall x \in \mathbb{R}^n.$$

Since σ_{C_1} and σ_{C_2} are the conjugates of δ_{C_1} and δ_{C_2} , respectively, the result follows from part (a).

To see that the assumption of closedness of C_1 and C_2 is needed, consider two convex sets that have the same closure, but none of the two is contained in the other, such as for example $(0, 1]$ and $[0, 1)$.

1.57 (Essentially One-Dimensional Functions)

We say that a closed proper convex function $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ is *essentially one-dimensional* if it has the form

$$f(x) = \bar{f}(a'x),$$

where a is a vector in \mathfrak{R}^n and $\bar{f} : \mathfrak{R} \mapsto (-\infty, \infty]$ is a scalar closed proper convex function. We say that a closed proper convex function $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ is *domain one-dimensional* if the affine hull of $\text{dom}(f)$ is either a single point or a line, i.e.,

$$\text{aff}(\text{dom}(f)) = \{\gamma a + b \mid \gamma \in \mathfrak{R}\},$$

where a and b are some vectors in \mathfrak{R}^n .

- (a) The conjugate of an essentially one-dimensional function is a domain one-dimensional function such that the affine hull of its domain is a subspace.
- (b) The conjugate of a domain one-dimensional function is the sum of an essentially one-dimensional function and a linear function.

Solution: (a) Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be essentially one-dimensional, so that

$$f(x) = \bar{f}(a'x),$$

where a is a vector in \mathfrak{R}^n and $\bar{f} : \mathfrak{R} \mapsto (-\infty, \infty]$ is a scalar closed proper convex function. If $a = 0$, then f is a constant function, so its conjugate is domain one-dimensional, since its domain is $\{0\}$. We may thus assume that $a \neq 0$. We claim that the conjugate

$$f^*(y) = \sup_{x \in \mathfrak{R}^n} \{y'x - \bar{f}(a'x)\}, \tag{1.31}$$

takes infinite values if y is outside the one-dimensional subspace spanned by a , implying that f^* is domain one-dimensional with the desired property. Indeed, let y be of the form $y = \xi a + v$, where ξ is a scalar, and v is a nonzero vector with $v \perp a$. If we take $x = \gamma a + \delta v$ in Eq. (1.31), where γ is such that $\gamma \|a\|^2 \in \text{dom}(\bar{f})$, we obtain

$$\begin{aligned} f^*(y) &= \sup_{x \in \mathfrak{R}^n} \{y'x - \bar{f}(a'x)\} \\ &\geq \sup_{\delta \in \mathfrak{R}} \{(\xi a + v)'(\gamma a + \delta v) - \bar{f}(\gamma \|a\|^2)\} \\ &= \xi \gamma \|a\|^2 - \bar{f}(\gamma \|a\|^2) + \sup_{\delta \in \mathfrak{R}} \{\delta \|v\|^2\}, \end{aligned}$$

so it follows that $f^*(y) = \infty$.

(b) Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be domain one-dimensional, so that

$$\text{aff}(\text{dom}(f)) = \{\gamma a + b \mid \gamma \in \mathfrak{R}\},$$

for some vectors a and b . If $a = b = 0$, the domain of f is $\{0\}$, so its conjugate is the function taking the constant value $-f(0)$ and is essentially one-dimensional. If $b = 0$ and $a \neq 0$, then the conjugate is

$$f^*(y) = \sup_{x \in \mathbb{R}^n} \{y'x - f(x)\} = \sup_{\gamma \in \mathbb{R}} \{\gamma a'y - f(\gamma a)\},$$

so $f^*(y) = \bar{f}^*(a'y)$ where \bar{f}^* is the conjugate of the scalar function $\bar{f}(\gamma) = f(\gamma a)$. Since \bar{f} is closed, convex, and proper, the same is true for \bar{g} , and it follows that f^* is essentially one-dimensional. Finally, consider the case where $b \neq 0$. Then we use a translation argument and write $f(x) = \hat{f}(x - b)$, where \hat{f} is a function such that the affine hull of its domain is the subspace spanned by a . The conjugate of \hat{f} is essentially one-dimensional (by the preceding argument), and the conjugate of f is obtained by adding $b'y$ to it.

1.58

Calculate the support functions of the following sets:

- (1) $C = \{x \mid \|x\| \leq 1\}$.
- (2) $C = \{a\} + \gamma\{x \mid \|x\| \leq 1\}$, where $a \in \mathbb{R}^n$ and $\gamma > 0$.
- (3) $C = \{x \mid x_1 + \cdots + x_n = 1, x_i \geq 0, i = 1, \dots, n\}$.
- (4) $C = \{x \mid |x_i| \leq 1, i = 1, \dots, n\}$.
- (5) $C = \{x \mid \sum_{i=1}^n |x_i| \leq 1\}$.

Solution: Answers:

- (1) $\sigma_C(y) = \|y\|$.
- (2) $\sigma_C(y) = a'y + \gamma\|y\|$.
- (3) $\sigma_C(y) = \max_{i=1, \dots, n} y_i$.
- (4) $\sigma_C(y) = \sum_{i=1}^n |y_i|$.
- (5) $\sigma_C(y) = \max_{i=1, \dots, n} |y_i|$.

1.59

Show that a subset of \mathbb{R}^n is bounded if and only if its support function is everywhere finite.

Solution:

1.60

Show that the support function of the unit ball of \mathbb{R}^n with respect to a norm $\|\cdot\|_a$ is another norm $\|\cdot\|_b$. Verify that if $\|\cdot\|_a$ is the l_1 -norm then $\|\cdot\|_b$ is the l_∞ -norm, and that if $\|\cdot\|_a$ is the l_∞ -norm then $\|\cdot\|_b$ is the l_1 -norm.

Solution:

1.61 (Support Function of a Bounded Ellipsoid)

Let X be an ellipsoid of the form

$$X = \{x \mid (x - \bar{x})'Q(x - \bar{x}) \leq b\},$$

where Q is a symmetric positive definite matrix, \bar{x} is a vector, and b is a positive scalar. Show that the support function of X is

$$\sigma_X(y) = y'\bar{x} + (by'Q^{-1}y)^{1/2}, \quad \forall y \in \mathbb{R}^n.$$

Solution: To calculate $\sigma_X(y)$, we write

$$X = \{\bar{x}\} + \bar{X},$$

where

$$\bar{X} = \{x \mid x'Qx \leq b\},$$

we calculate the support function $\sigma_{\bar{X}}(y)$, and we use the equation

$$\sigma_X(y) = y'\bar{x} + \sigma_{\bar{X}}(y). \quad (1.32)$$

To calculate

$$\sigma_{\bar{X}}(y) = \sup_{x'Qx \leq b} y'x,$$

we introduce the transformation $z = Q^{1/2}x$ and we write

$$\sigma_{\bar{X}}(y) = \sup_{\|z\| \leq b^{1/2}} y'Q^{-1/2}z.$$

It can be seen that for $y \neq 0$, the supremum over z above is attained at

$$z(y) = b^{1/2} \frac{Q^{-1/2}y}{\|Q^{-1/2}y\|},$$

and by substitution in the expression for $\sigma_{\bar{X}}(y)$, we have

$$\sigma_{\bar{X}}(y) = (by'Q^{-1}y)^{1/2}.$$

Thus, using Eq. (1.32), we finally obtain

$$\sigma_X(y) = y'\bar{x} + (by'Q^{-1}y)^{1/2}, \quad \forall y \in \mathbb{R}^n.$$

1.62 (Support Function of Sum and Union)

Let X_1, \dots, X_r , be nonempty subsets of \mathfrak{R}^n . Derive formulas for $X_1 + \dots + X_r$, $\text{conv}(X_1) + \dots + \text{conv}(X_r)$, $\cup_{j=1}^r X_j$, and $\text{conv}(\cup_{j=1}^r X_j)$.

Solution: Let $X = X_1 + \dots + X_r$. We have for all $y \in \mathfrak{R}^n$,

$$\begin{aligned}\sigma_X(y) &= \sup_{x \in X_1 + \dots + X_r} x'y \\ &= \sup_{x_1 \in X_1, \dots, x_r \in X_r} (x_1 + \dots + x_r)'y \\ &= \sup_{x_1 \in X_1} x_1'y + \dots + \sup_{x_r \in X_r} x_r'y \\ &= \sigma_{X_1}(y) + \dots + \sigma_{X_r}(y).\end{aligned}$$

Since X_j and $\text{conv}(X_j)$ have the same support function, it follows that

$$\sigma_{X_1}(y) + \dots + \sigma_{X_r}(y)$$

is also the support function of

$$\text{conv}(X_1) + \dots + \text{conv}(X_r).$$

Let also $X = \cup_{j=1}^r X_j$. We have

$$\sigma_X(y) = \sup_{x \in X} y'x = \max_{j=1, \dots, r} \sup_{x \in X_j} y'x = \max_{j=1, \dots, r} \sigma_{X_j}(y).$$

This is also the support function of $\text{conv}(\cup_{j=1}^r X_j)$.

1.63 (Positively Homogeneous Functions and Support Functions)

A function $f : \mathfrak{R}^n \mapsto [-\infty, \infty]$ is called *positively homogeneous* if its epigraph is a cone in \mathfrak{R}^{n+1} . Equivalently, f is positively homogeneous if and only if

$$f(\gamma x) = \gamma f(x), \quad \forall \gamma > 0, \forall x \in \mathfrak{R}^n.$$

- (a) Show that if f is a proper convex positively homogeneous function, then for all $x_1, \dots, x_m \in \mathfrak{R}^n$ and $\gamma_1, \dots, \gamma_m > 0$, we have

$$f(\gamma_1 x_1 + \dots + \gamma_m x_m) \leq \gamma_1 f(x_1) + \dots + \gamma_m f(x_m).$$

- (b) Clearly, the support function σ_X of a nonempty set X is closed proper convex and positively homogeneous. Show that the reverse is also true, namely that the closure of any proper convex positively homogeneous function is the support function of some set. In particular, show that if $\sigma : \mathfrak{R}^n \mapsto$

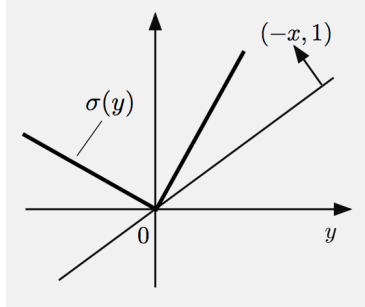


Figure 1.1. Visualization of the conjugate of a positively homogeneous function σ (cf. Exercise 1.63). The values of its conjugate correspond to crossing levels of the vertical axis by hyperplanes supporting the epigraph of σ . There can be only one such level, namely 0, so the conjugate of σ takes only the values 0 and ∞ , i.e., it is the indicator function of the set whose elements x correspond to normals $(-x, 1)$ of hyperplanes supporting $\text{epi}(f)$ at 0, as shown in the figure.

$(-\infty, \infty]$ is a proper convex positively homogeneous function, then the conjugate of σ is the indicator function of the closed convex set

$$X = \{x \mid y'x \leq \sigma(y), \forall y \in \mathbb{R}^n\},$$

and $\text{cl } \sigma$ is the support function of X (see Fig. 1.1 for a geometric interpretation of this result).

Solution: (a) Let

$$\bar{\gamma} = \gamma_1 + \cdots + \gamma_m,$$

and

$$\bar{\gamma}_i = \frac{\gamma_i}{\bar{\gamma}}, \quad i = 1, \dots, m.$$

By convexity of f , we have

$$f(\bar{\gamma}_1 x_1 + \cdots + \bar{\gamma}_m x_m) \leq \bar{\gamma}_1 f(x_1) + \cdots + \bar{\gamma}_m f(x_m).$$

Since f is positively homogeneous, this inequality can be written as

$$\frac{1}{\bar{\gamma}} f(\gamma_1 x_1 + \cdots + \gamma_m x_m) \leq \frac{1}{\bar{\gamma}} (\gamma_1 f(x_1) + \cdots + \gamma_m f(x_m)),$$

and the result follows.

(b) Let δ be the conjugate of σ :

$$\delta(x) = \sup_{y \in \mathbb{R}^n} \{y'x - \sigma(y)\}.$$

Since σ is positively homogeneous, we have for any $\gamma > 0$,

$$\gamma \delta(x) = \sup_{y \in \mathbb{R}^n} \{\gamma y'x - \gamma \sigma(y)\} = \sup_{y \in \mathbb{R}^n} \{(\gamma y)'x - \sigma(\gamma y)\}.$$

The right-hand sides of the preceding two relations are equal, so we obtain

$$\delta(x) = \gamma \delta(x), \quad \forall \gamma > 0,$$

which implies that δ takes only the values 0 and ∞ (since σ and hence also its conjugate δ is proper). Thus, δ is the indicator function of a set, call it X , and we have

$$\begin{aligned} X &= \{x \mid \delta(x) \leq 0\} \\ &= \left\{x \mid \sup_{y \in \mathfrak{R}^n} \{y'x - \sigma(y)\} \leq 0\right\} \\ &= \{x \mid y'x \leq \sigma(y), \forall y \in \mathfrak{R}^n\}. \end{aligned}$$

Finally, since δ is the conjugate of σ , we see that $\text{cl } \sigma$ is the conjugate of δ ; cf. the Conjugacy Theorem [Prop. 1.6.1(c)]. Since δ is the indicator function of X , it follows that $\text{cl } \sigma$ is the support function of X .

1.64 (Support Function of Domain)

Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a proper convex function, and let f^* be its conjugate.

- (a) The support function of $\text{dom}(f)$ is the recession function of f^* .
- (b) If f is closed, the support function of $\text{dom}(f^*)$ is the recession function of f .

Solution: (a) From the definition

$$f^*(y) = \sup_{x \in \mathfrak{R}^n} \{x'y - f(x)\},$$

we see that f^* is the pointwise supremum of the affine functions

$$h_{(x,w)}(y) = x'y - w,$$

as (x, w) ranges over $\text{epi}(f)$. Therefore, $\text{epi}(f^*)$ is the intersection of the epigraphs of $h_{(x,w)}$ as (x, w) ranges over $\text{epi}(f)$. Hence, by the Recession Cone Theorem [Prop. 1.4.1(e)], the recession cone of $\text{epi}(f^*)$ is the intersection of the recession cones of the epigraphs of $h_{(x,w)}$ as (x, w) ranges over $\text{epi}(f)$. Since the epigraph of $h_{(x,w)}$ is $\{(y, u) \mid x'y - w \leq u\}$, its recession cone is $\{(y, u) \mid x'y \leq u\}$, and we have

$$R_{\text{epi}(f^*)} = \bigcap_{(x,w) \in \text{epi}(f)} \{(y, u) \mid x'y \leq u\}.$$

Since $R_{\text{epi}(f^*)}$ is the epigraph of the recession function r_f^* of f^* , it follows that

$$r_f^*(y) = \sup_{x \in \text{dom}(f)} x'y,$$

so r_f^* is the support function of $\text{dom}(f)$.

(b) If f is closed, then by the Conjugacy Theorem [Prop. 1.6.1(c)], it is the conjugate of f^* , and the result follows by applying part (a) with the roles of f and f^* interchanged.

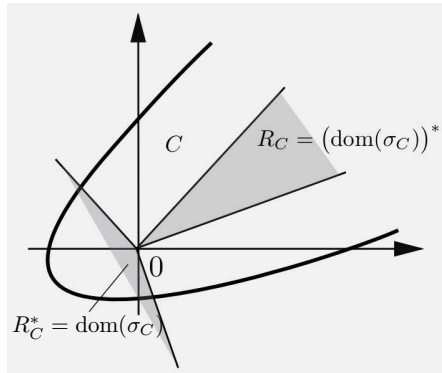


Figure 1.2. Illustration of the polarity of the recession cone R_C of a closed convex set C and the domain $\text{dom}(\sigma_C)$ of its support function (cf. Exercise 1.65). For any $y \notin R_C^*$ we have

$$\sup_{x \in C} y'x = \infty,$$

while for any $y \in R_C^*$, we must have $\sup_{x \in C} y'x < \infty$.

1.65 (Polarity of Recession Cone and Domain of Support Function)

Let C be a nonempty closed convex set in \mathfrak{R}^n . Then, the recession cone of C is equal to the polar cone of the domain of σ_C :

$$R_C = (\text{dom}(\sigma_C))^*.$$

Solution: We apply the result of Exercise 1.64(b) with f and f^* equal to the indicator function δ_C and support function σ_C of C , respectively. We obtain

$$r_{\delta_C} = \sigma_{\text{dom}(\sigma_C)},$$

from which, by using also the formula of Example 1.6.2,

$$R_C = \{y \mid r_{\delta_C}(y) \leq 0\} = \{y \mid \sigma_{\text{dom}(\sigma_C)}(y) \leq 0\} = (\text{dom}(\sigma_C))^*.$$

1.66 (Generated Functions - Support Function of 0-Level Set)

Given a proper convex function $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$, the closure of the cone generated by $\text{epi}(f)$, is the epigraph of a closed convex positively homogeneous function, called the *closed function generated by f* , and denoted by $\text{gen } f$. The epigraph of $\text{gen } f$ is the intersection of all the halfspaces that contain $\text{epi}(f)$ and contain 0 in their boundary. Alternatively, the epigraph of $\text{gen } f$ is the intersection of all the closed cones that contain $\text{epi}(f)$.

- Show that if the level set $\{y \mid h(y) \leq 0\}$ is nonempty, the generated function $\text{gen } f$ is proper.
- Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a closed proper convex function and let f^* be its conjugate. Show that if the level set $\{y \mid f^*(y) \leq 0\}$ [respectively $\{x \mid f(x) \leq 0\}$] is nonempty, its support function is the closed function generated by f (respectively f^*).

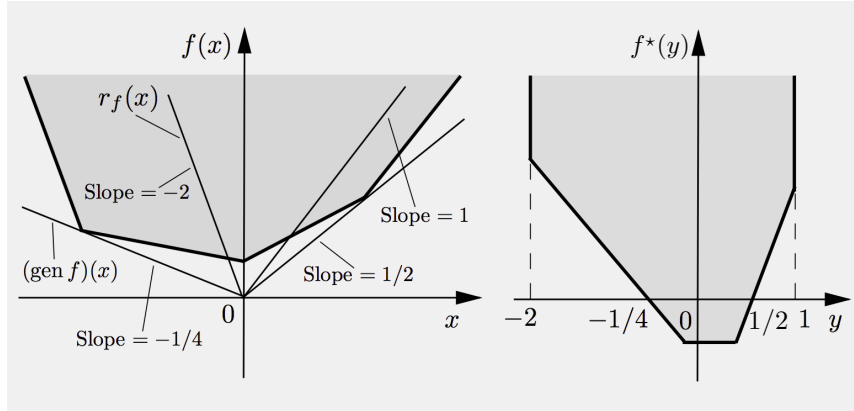


Figure 1.3. Illustration of Exercise 1.66. The recession function r_f and the closed generated function $\text{gen } f$ are the support functions of the sets $\text{dom}(f^*)$ and $\{y \mid f^*(y) \leq 0\}$, respectively.

Solution: (a) If $(\text{cl } f)(0) > 0$, by the Nonvertical Hyperplane Theorem (Prop. 1.5.8), there exists a nonvertical hyperplane passing through the origin and containing $\text{epi}(f)$ in one of its closed halfspaces, implying that the epigraph of $\text{gen } f$ does not contain a line, so $\text{gen } f$ is proper. Any y such that $f^*(y) \leq 0$, or equivalently $y'x \leq f(x)$ for all x , defines a nonvertical hyperplane that separates the origin from $\text{epi}(f)$.

(b) Let σ be the closed function generated by f . Then, since $\{y \mid f^*(y) \leq 0\} \neq \emptyset$, by part (a), σ is proper, and by Exercise 1.63, σ is the support function of the set

$$Y = \{y \mid y'x \leq \sigma(x), \forall x \in \mathbb{R}^n\}.$$

Since $\text{epi}(\sigma)$ is the intersection of all the halfspaces that contain $\text{epi}(f)$ and contain 0 in their boundary, the set Y can be written as

$$Y = \{y \mid y'x \leq f(x), \forall x \in \mathbb{R}^n\},$$

from which we obtain

$$Y = \left\{y \mid \sup_{x \in \mathbb{R}^n} \{y'x - f(x)\} \leq 0\right\} = \{y \mid f^*(y) \leq 0\}.$$

Note that the method used to characterize the 0-level set of f can be applied to any level set. In particular, a nonempty level set $L_\gamma = \{x \mid f(x) \leq \gamma\}$ is the 0-level set of the function f_γ defined by $f_\gamma(x) = f(x) - \gamma$, and its support function is the closed function generated by h_γ , the conjugate of f_γ , which is given by $h_\gamma(y) = h(y) + \gamma$. The preceding analysis is illustrated in Fig. 1.3.

1.67 (Conjugates Involving Invertible Linear Transformations)

Let $p : \mathfrak{R}^n \mapsto [-\infty, \infty]$ be a convex function, let A be an invertible $n \times n$ matrix, let a and c be vectors in \mathfrak{R}^n , and let b be a scalar. Calculate the conjugate of the convex function

$$f(x) = p(A(x - c)) + a'x + b.$$

Solution: Using the transformation $z = A(x - c)$, we can write the conjugate as

$$\begin{aligned} f^*(y) &= \sup_{x \in \mathfrak{R}^n} \{x'y - p(A(x - c)) - a'x - b\} \\ &= \sup_{z \in \mathfrak{R}^n} \{(A^{-1}z + c)'y - p(z) - (A^{-1}z + c)'a - b\} \\ &= \sup_{z \in \mathfrak{R}^n} \{(A^{-1}z)'(y - a) - p(z)\} + c'(y - a) - b \end{aligned}$$

and finally

$$f^*(y) = p^*((A')^{-1}(y - a)) + c'y + d,$$

where p^* is the conjugate of p and

$$d = -(c'a + b).$$

Note the symmetry between f and f^* .

1.68 (Conjugate of a Partially Affine Function)

A *partially affine* function f is defined as a function such that $\text{dom}(f)$ is an affine set, and f is affine on $\text{dom}(f)$. Show that the conjugate of a partially affine function is another partially affine function. *Hint:* Write f as

$$f(x) = \delta_{\text{aff}(f)}(x) + a'x + b$$

and apply the result of Exercise 1.67.

Solution: See p. 107 of [Roc70]. Write f as

$$f(x) = \delta_{\text{aff}(f)}(x) + a'x + b$$

and apply the result of Exercise 1.67.

1.69

Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a proper convex function and let f^* be its conjugate. Show that:

- The constancy space of f^* is the orthogonal complement of the subspace parallel to $\text{aff}(\text{dom}(f))$. *Hint:* Use Exercise 1.64.
- If f is closed, the subspace parallel to $\text{aff}(\text{dom}(h))$ is the orthogonal complement of the constancy space of f .

Solution: *Hint:* Use Exercise 1.64. See Theorem 13.3 of [Roc70].

1.70

Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a closed proper convex function, and let f^* be its conjugate. Show that $\text{dom}(f)$ is an affine set if and only if the recession function of f^* satisfies $r_h(d) = \infty$ for all d that are not in the constancy space of f . *Hint:* Use Exercise 1.64.

Solution: *Hint:* Use Exercise 1.64. See Theorem 13.3 of [Roc70].

1.71 (Co-finite Functions)

A closed proper convex function $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ is said to be co-finite if its recession function is

$$r_f(d) = \begin{cases} 0 & \text{if } d = 0, \\ \infty & \text{if } d \neq 0. \end{cases}$$

Show that f is co-finite if and only if its conjugate is real-valued. *Hint:* Use Exercise 1.64.

Solution: See Theorem 13.3 of [Roc70].

1.72 (Lipschitz Continuity and Domain Boundedness)

Let $f : \mathfrak{R}^n \mapsto (-\infty, \infty]$ be a proper convex function, and let f^* be its conjugate. Show that $\text{dom}(f^*)$ is bounded if and only if f is real-valued and there exists a scalar $L \geq 0$ such that

$$|f(x) - f(y)| \leq L\|x - y\|, \quad \forall x \in \mathfrak{R}^n, y \in \mathfrak{R}^n.$$

Furthermore, the minimal value of L for which this Lipschitz condition holds is

$$\sup_{y \in \text{dom}(f^*)} \|y\|.$$

Hint: Use Exercise 1.64.

Solution: See Theorem 13.3, Corollary 13.3.3 of [Roc70].

1.73

Let C be a nonempty convex subset of \mathfrak{R}^n and let σ_C be its support function.

- $x \in \text{cl}(C)$ if and only if $y'x \leq \sigma_C(y)$ for all $y \in \mathfrak{R}^n$.
- $x \in \text{ri}(C)$ if and only if $y'x \leq \sigma_C(y)$ for all $y \in \mathfrak{R}^n$ and $y'x < \sigma_C(y)$ for all $y \in \mathfrak{R}^n$ with $-\sigma_C(-y) \neq \sigma_C(y)$.
- x is an interior point of C if and only if $y'x < \sigma_C(y)$ for all $y \neq 0$.

Solution: See Theorem 13.1 of [Roc70].

REFERENCES

- [Hel21] Helly, E., 1921. "Über Systeme Linearer Gleichungen mit Unendlich Vielen Unbekannten," Monatschr. Math. Phys., Vol. 31, pp. 60-91.
- [Kir1903] Kirchberger, P., 1903. "Über Tschebyscheffsche Annäherungsmethoden," Mathematische Annalen, Vo. 57, pp. 509-540.
- [Kra46] Krasnoselsky, M. A., 1946. "Sur un Critere pur qu'un Domain Soit Etoile," Matematiceskii Sbornik, (Russian with French summary), Novaja Serija, Vol. 19, pp. 309-310.
- [Roc70] Rockafellar, R. T., 1970. Convex Analysis, Princeton Univ. Press, Princeton, N. J.
- [Web94] Webster, R., 1994. Convexity, Oxford Univ. Press, Oxford, UK.