

# Homework Assignment 2

2025.11.14

## Problem 1. (15 points)

(a) 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  $\left\{ \left( k, \sqrt{k}, \sqrt{k^2 + k} \right) \right\}$ .

(b) 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 1** (a) First, we show that  $\left\{ x_k = (k, \sqrt{k}, \sqrt{k^2 + k}) \right\}$  is an asymptotic sequence for the cone  $C$  in the direction  $d = (1, 0, 1)$ .

We have

$$\|x_k\| = \sqrt{k^2 + k + k^2 + k} = \sqrt{2k^2 + 2k} = \sqrt{2}\sqrt{k^2 + k}$$

Thus,

$$\frac{x_k}{\|x_k\|} = \left( \frac{k}{\sqrt{2}\sqrt{k^2 + k}}, \frac{\sqrt{k}}{\sqrt{2}\sqrt{k^2 + k}}, \frac{\sqrt{k^2 + k}}{\sqrt{2}\sqrt{k^2 + k}} \right) = \left( \frac{k}{\sqrt{2}\sqrt{k^2 + k}}, \frac{\sqrt{k}}{\sqrt{2}\sqrt{k^2 + k}}, \frac{1}{\sqrt{2}} \right)$$

As  $k \rightarrow \infty$ , we have

$$\lim_{k \rightarrow \infty} \frac{x_k}{\|x_k\|} = \left( \frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right) = \frac{1}{\sqrt{2}}(1, 0, 1) = \frac{d}{\|d\|}$$

Next, we check that  $x_k - d \in C$  not true for all  $k$  sufficiently large. Which equivalently means checking if  $\left\| (k-1, \sqrt{k}, \sqrt{k^2+k}-1) \right\| \leq \sqrt{k^2+k}-1 \iff -2k \leq -2\sqrt{k^2+k}$  which is not true for all  $k$  sufficiently large.

(b) We want to show  $C = \bigcap_{\theta \in [0, 2\pi]} H_\theta$ , where  $H_\theta = \{(u, v, w) \mid u \cos \theta + v \sin \theta - w \leq 0\}$  are closed halfspaces. This is true since the inequality  $\|(u, v)\| \leq w$  is equivalent to

$$u \cos \theta + v \sin \theta \leq w \quad \text{for all } \theta \in [0, 2\pi]$$

Each  $H_\theta$  is a closed halfspace defined by the normal  $(\cos \theta, \sin \theta, -1)$ . Each  $H_\theta$  is a polyhedral set and therefore retractive.

## Problem 2. (15 points)

Let  $C_1$  and  $C_2$  be nonempty convex subsets of  $\mathbf{R}^n$ , and let  $B$  denote the unit ball in  $\mathbf{R}^n$ ,  $B = \{x \mid \|x\| \leq 1\}$ . 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 the following three conditions are equivalent.

- (i) There exists a hyperplane separating strongly  $C_1$  and  $C_2$ .
- (ii) There exists a vector  $\alpha \in \mathbf{R}^n$  such that  $\inf_{x \in C_1} \alpha'x > \sup_{x \in C_2} \alpha'x$ .
- (iii)  $\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 0$ , i.e.,  $0 \notin \text{cl}(C_2 - C_1)$ .

**Solution 2** (i)  $\implies$  (ii): There exists a hyperplane  $H = \{x \mid \alpha'x = b\}$  and an  $\epsilon > 0$  such that  $C_1 + \epsilon B$  and  $C_2 + \epsilon B$  are in opposite open halfspaces. Assume  $C_1 + \epsilon B \subset \{x \mid \alpha'x > b\}$  and  $C_2 + \epsilon B \subset \{x \mid \alpha'x < b\}$ .

For any  $x_1 \in C_1$  and  $z \in B$ , we have  $\alpha'(x_1 + \epsilon z) > b$ , which means  $\alpha'x_1 + \inf_{z \in B} (\epsilon \alpha'z) > b$ .  $\inf_{z \in B} (\alpha'z) = -\sup_{z \in B} (-\alpha'z) = -\|\alpha\|$ . So,  $\alpha'x_1 - \epsilon \|\alpha\| > b$  for all  $x_1 \in C_1$ . Taking the infimum over  $x_1$ :

$$\inf_{x_1 \in C_1} \alpha'x_1 \geq b + \epsilon \|\alpha\|$$

Similarly,

$$\sup_{x_2 \in C_2} \alpha'x_2 \leq b - \epsilon \|\alpha\|$$

Combining these, we have:

$$\sup_{x_2 \in C_2} \alpha'x_2 \leq b - \epsilon \|\alpha\| < b + \epsilon \|\alpha\| \leq \inf_{x_1 \in C_1} \alpha'x_1$$

(ii)  $\implies$  (iii): There exists  $\alpha \neq 0$  such that  $\inf_{x_1 \in C_1} \alpha'x_1 > \sup_{x_2 \in C_2} \alpha'x_2$ . Let  $\gamma_1 = \inf_{x_1 \in C_1} \alpha'x_1$  and  $\gamma_2 = \sup_{x_2 \in C_2} \alpha'x_2$ . Let  $\delta = \gamma_1 - \gamma_2 > 0$ . Then  $\|\alpha\| \|x_1 - x_2\| \geq \alpha'(x_1 - x_2) = \alpha'x_1 - \alpha'x_2 \geq \gamma_1 - \gamma_2 = \delta$ . This implies  $\|x_1 - x_2\| \geq \frac{\delta}{\|\alpha\|}$  for all  $x_1 \in C_1, x_2 \in C_2$ . Therefore,  $\inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| \geq \frac{\delta}{\|\alpha\|} > 0$ .

(iii)  $\implies$  (i): Let  $\delta = \inf_{x_1 \in C_1, x_2 \in C_2} \|x_1 - x_2\| > 0$ . Let  $D = C_1 - C_2 = \{x_1 - x_2 \mid x_1 \in C_1, x_2 \in C_2\}$  is convex since  $C_1$  and  $C_2$  are convex. We have  $\inf_{d \in D} \|d\| = \delta > 0$ . This implies that the convex set  $D$  is disjoint from the open ball  $B(0, \delta) = \{x \mid \|x\| < \delta\}$ . By the Separating Hyperplane Theorem, there exists a hyperplane  $H = \{x \mid \alpha'x = b\}$  that separates  $D$  and  $B(0, \delta)$ . Let  $\epsilon = \frac{\delta}{2\|\alpha\|} > 0$ . Then, for any  $x_1 \in C_1$  and  $z \in B$ , we have:

$$\alpha'(x_1 + \epsilon z - x_2) = \alpha'(x_1 - x_2) + \epsilon \alpha'z \geq b + \epsilon(-\|\alpha\|) = b - \frac{\delta}{2}$$

Similarly, for any  $x_2 \in C_2$  and  $z \in B$ , we have:

$$\alpha'(x_1 - (x_2 + \epsilon z)) = \alpha'(x_1 - x_2) - \epsilon \alpha'z \leq b + \epsilon \|\alpha\| = b + \frac{\delta}{2}$$

Thus,  $C_1 + \epsilon B$  and  $C_2 + \epsilon B$  are contained in opposite open halfspaces defined by  $H$ .

### Problem 3. (15 points)

Let  $F : \mathbf{R}^{n+m} \mapsto (-\infty, \infty]$  be a closed proper convex function of two vectors  $x \in \mathbf{R}^n$  and  $z \in \mathbf{R}^m$ , and let

$$X = \left\{ x \mid \inf_{z \in \mathbf{R}^m} F(x, z) < \infty \right\}.$$

Assume that the function  $F(x, \cdot)$  is closed for each  $x \in X$ . Show that:

(a) If for some  $\bar{x} \in X$ , the minimum of  $F(\bar{x}, \cdot)$  over  $\mathbf{R}^m$  is attained at a nonempty and compact set, the same is true for all  $x \in X$ .

(b) If the functions  $F(x, \cdot)$  are differentiable for all  $x \in X$ , they have the same asymptotic slopes along all directions, i.e., for each  $d \in \mathbf{R}^m$ , the value of  $\lim_{\alpha \rightarrow \infty} \nabla_z F(x, z + \alpha d)'d$  is the same for all  $x \in X$  and  $z \in \mathbf{R}^m$ .

**Solution 3** (a) Let  $g_x(z) = F(x, z)$ . We are given that for a specific  $\tilde{x} \in X$ ,  $g_{\tilde{x}}(z)$  is a closed convex function and its set of minimizers  $M(\tilde{x}) = \arg \min_z g_{\tilde{x}}(z)$  is nonempty and compact.

For a closed proper convex function  $g$ , the set of minimizers is nonempty and compact if and only if  $g$  recession function  $r_g(d)$  satisfies  $r_g(d) > 0$  for all  $d \neq 0$ . The recession function of  $g_x(z) = F(x, z)$  is

$$r_{g_x}(d_z) = \lim_{\alpha \rightarrow \infty} \frac{F(x, z_0 + \alpha d_z) - F(x, z_0)}{\alpha}$$

for any  $z_0$  in the domain of  $g_x$ .

So we have  $r_F(0, d_z) = r_{g_{\tilde{x}}}(d_z) > 0$  for all  $d_z \neq 0$ . Since  $x \in X$ , we know  $\inf_z F(x, z) < \infty$ . Thus  $M(x) = \arg \min_z F(x, z)$  is nonempty. The set of minimizers  $M(x)$  is a level set of  $g_x$ , and since  $g_x$  is closed,  $M(x)$  is closed. A closed and bounded set in  $\mathbf{R}^m$  is compact. Thus, for all  $x \in X$ ,  $M(x)$  is nonempty and compact.

(b) Let  $g_x(z) = F(x, z)$ . We have

$$\lim_{\alpha \rightarrow \infty} \nabla_z F(x, z + \alpha d)'d = r_{g_x}(d)$$

This is because  $\nabla g_x(z + \alpha d)'d$  is non-decreasing in  $\alpha$  and its limit is  $\sup_{\alpha} \nabla g_x(z + \alpha d)'d$ , which equals  $r_{g_x}(d)$ . As we established in part (a),  $r_{g_x}(d) = r_F(0, d)$ . This value  $r_F(0, d)$  depends only on  $F$  and  $d$ , and is independent of  $x$  and  $z$ . Therefore, the value of the asymptotic slope is the same for all  $x \in X$  and  $z \in \mathbf{R}^m$ .

### Problem 4. (15 points)

Let  $f : \mathbf{R}^n \mapsto \mathbf{R}$  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 4**

$$\begin{aligned} f^*(y) &= \sup_{x \in \mathbb{R}^n} \{y'x - f(x)\} = \sup_{x \in \mathbb{R}^n} \left\{ \sum_{i=1}^n y_i x_i - \frac{1}{p} \sum_{i=1}^n |x_i|^p \right\} \\ &= \sum_{i=1}^n \sup_{x_i \in \mathbb{R}} \left\{ y_i x_i - \frac{1}{p} |x_i|^p \right\} \\ (x_i^* &= |y_i|^{q-1}) = \sum_{i=1}^n \left( 1 - \frac{1}{p} \right) |y_i|^q \\ &= \frac{1}{q} \sum_{i=1}^n |y_i|^q \end{aligned}$$

**Problem 5. (15 points)**

(a) Show that if  $f_1 : \mathbf{R}^n \mapsto (-\infty, \infty]$  and  $f_2 : \mathbf{R}^n \mapsto (-\infty, \infty]$  are closed proper convex functions, with conjugates denoted by  $f_1^*$  and  $f_2^*$ , respectively, we have

$$f_1(x) \leq f_2(x), \quad \forall x \in \mathbf{R}^n$$

if and only if

$$f_1^*(y) \geq f_2^*(y), \quad \forall y \in \mathbf{R}^n.$$

(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

$$\sigma_{C_1}(y) \leq \sigma_{C_2}(y), \quad \forall y \in \mathbf{R}^n.$$

Construct an example showing that closedness of  $C_1$  and  $C_2$  is a necessary assumption.

**Solution 5** (a) “ $\implies$ ”:

$$f_1^*(y) = \sup_x \{y'x - f_1(x)\} \geq \sup_x \{y'x - f_2(x)\} = f_2^*(y)$$

“ $\impliedby$ ”: Since  $f_1$  and  $f_2$  are closed proper convex functions,  $f = f^{**}$ .

$$f_1(x) = f_1^{**}(x) = \sup_y \{x'y - f_1^*(y)\} \leq \sup_y \{x'y - f_2^*(y)\} = f_2^{**}(x) = f_2(x)$$

(b) Let  $\delta_{C_1}(x)$  and  $\delta_{C_2}(x)$  be the indicator functions for the sets  $C_1$  and  $C_2$ .

$$C_1 \subset C_2 \iff \delta_{C_2}(x) \leq \delta_{C_1}(x), \forall x \iff \delta_{C_2}^*(y) \geq \delta_{C_1}^*(y), \forall y \iff \sigma_{C_2}(y) \geq \sigma_{C_1}(y) \quad \forall y$$

**Counterexample:** Let  $C_1 = [0, 1]$ ,  $C_2 = (0, 1)$ .

$$\sigma_{C_1}(y) = \sup_{x \in [0,1]} \{yx\} = \begin{cases} y & \text{if } y > 0 \\ 0 & \text{if } y \leq 0 \end{cases} = \max(0, y)$$

$$\sigma_{C_2}(y) = \sup_{x \in (0,1)} \{yx\} = \begin{cases} y & \text{if } y > 0 \\ 0 & \text{if } y \leq 0 \end{cases} = \max(0, y)$$

Here,  $\sigma_{C_1}(y) = \sigma_{C_2}(y)$ , so  $\sigma_{C_1}(y) \leq \sigma_{C_2}(y)$  is satisfied. However,  $C_1 \not\subseteq C_2$ .

### Problem 6. (15 points)

Consider a function  $\phi$  of two real variables  $x$  and  $z$  taking values in compact intervals  $X$  and  $Z$ , respectively. Assume that for each  $z \in Z$ , the function  $\phi(\cdot, z)$  is minimized over  $X$  at a unique point denoted  $\hat{x}(z)$ . Similarly, assume that for each  $x \in X$ , the function  $\phi(x, \cdot)$  is maximized over  $Z$  at a unique point denoted  $\hat{z}(x)$ . Assume further that the functions  $\hat{x}(z)$  and  $\hat{z}(x)$  are continuous over  $Z$  and  $X$ , respectively. Show that  $\phi$  has a saddle point  $(x^*, z^*)$ . Use this to investigate the existence of saddle points of  $\phi(x, z) = x^2 + z^2$  over  $X = [0, 1]$  and  $Z = [0, 1]$ .

#### Solution 6

$$\phi(x^*, z) \leq \phi(x^*, z^*) \leq \phi(x, z^*)$$

which gives:

$$x^* = \hat{x}(z^*)$$

$$z^* = \hat{z}(x^*)$$

Define  $T : X \times Z \rightarrow X \times Z$  as

$$T(x, z) = (\hat{x}(z), \hat{z}(x))$$

Since  $X$  and  $Z$  are compact intervals, they are compact and convex,  $X \times Z$  is also compact, convex. Since  $\hat{x}(z)$  and  $\hat{z}(x)$  are continuous functions,  $T(x, z)$  is a continuous function from  $X \times Z$  to  $X \times Z$ . By the Brouwer Fixed-Point Theorem, there exists a point  $(x^*, z^*) \in X \times Z$  such that  $T(x^*, z^*) = (x^*, z^*)$ .

For  $\phi$ , we have  $(x^*, z^*) = (0, 1)$  since:

$$\begin{aligned} \phi(0, z) &\leq \phi(0, 1) \leq \phi(x, 1) \\ \iff 0^2 + z^2 &\leq 0^2 + 1^2 \leq x^2 + 1^2 \\ \iff z^2 &\leq 1 \leq x^2 + 1 \end{aligned}$$

**Problem 7. (10 points)**

let  $F(x, u) = f_1(x) + f_2(Ax + u)$ , where  $A$  is an  $m \times n$  matrix, and  $f_1 : \mathbf{R}^n \mapsto (-\infty, \infty]$  and  $f_2 : \mathbf{R}^m \mapsto (-\infty, \infty]$  are closed convex functions. Show that the dual function is

$$q(\mu) = -f_1^*(A'\mu) - f_2^*(-\mu)$$

where  $f_1^*$  and  $f_2^*$  are the conjugate functions of  $f_1$  and  $f_2$ , respectively.

**Solution 7** Let  $p = \inf_x F(x, 0) = \inf_x \{f_1(x) + f_2(Ax)\}$ ,  $L(x, u, \mu) = F(x, u) + \mu'u$ :

$$\begin{aligned} q(\mu) &= \inf_{x \in \mathbf{R}^n, u \in \mathbf{R}^m} \{L(x, u, \mu)\} = \inf_{x, u} \{f_1(x) + f_2(Ax + u) + \mu'u\} \\ &= \inf_{x \in \mathbf{R}^n} \left\{ f_1(x) + \inf_{u \in \mathbf{R}^m} \{f_2(Ax + u) + \mu'u\} \right\} \end{aligned}$$

Set  $z = Ax + u$ . Then  $u = z - Ax$ .

$$\begin{aligned} \inf_u \{f_2(Ax + u) + \mu'u\} &= \inf_z \{f_2(z) + \mu'(z - Ax)\} \\ &= \inf_z \{f_2(z) + \mu'z - \mu'Ax\} \\ &= -f_2^*(-\mu) - \mu'Ax \end{aligned}$$

So the dual function is:

$$\begin{aligned} q(\mu) &= \inf_{x \in \mathbf{R}^n} \{f_1(x) + (-\mu'Ax - f_2^*(-\mu))\} \\ &= \inf_{x \in \mathbf{R}^n} \{f_1(x) - \mu'Ax\} - f_2^*(-\mu) \\ &= \inf_{x \in \mathbf{R}^n} \{f_1(x) - (A'\mu)'x\} - f_2^*(-\mu) \\ &= -f_1^*(A'\mu) - f_2^*(-\mu) \end{aligned}$$