

# Linear and Convex Optimization Homework 02

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## 1. Proof:

By the definition,  $\forall \mathbf{x}_1, \mathbf{x}_2 \in f^{-1}(C), \forall \theta \in [0,1], \exists \mathbf{y}_1, \mathbf{y}_2 \in C$  s.t.  $\mathbf{y}_1 = f(\mathbf{x}_1), \mathbf{y}_2 = f(\mathbf{x}_2)$ .

Let  $\bar{\theta} = 1 - \theta$ . Since  $C$  is convex,  $\theta \mathbf{y}_1 + \bar{\theta} \mathbf{y}_2 \in C$ .

Since  $f(\mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{b}$  is an affine function,

$$f(\theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2) = \mathbf{A}(\theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2) + \mathbf{b} = \theta \mathbf{A}\mathbf{x}_1 + \theta \mathbf{b} + \bar{\theta} \mathbf{A}\mathbf{x}_2 + \bar{\theta} \mathbf{b} = \theta \mathbf{y}_1 + \bar{\theta} \mathbf{y}_2 \in C,$$
$$\text{i.e. } \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2 \in f^{-1}(C).$$

Thus,  $f^{-1}(C)$  is also convex.

*Qed.* ■

## 2. Proof:

First we prove  $\mathbf{0} \notin C$  by contradiction. If  $\mathbf{0} \in C$ , by definition we know  $\exists \mathbf{x}_1 \in C_1, \mathbf{x}_2 \in C_2$  s.t.  $\mathbf{x}_1 - \mathbf{x}_2 = \mathbf{0}$ , i.e.  $C_1 \cap C_2 = \{\mathbf{x}_1\} \neq \emptyset$ . Contradiction.

Thus,  $\mathbf{0} \notin C$ .

Then we prove  $C$  is a nonempty set.

Since  $C_1$  and  $C_2$  are both nonempty sets, there exist at least one  $\mathbf{x}_1 \in C_1$  and one  $\mathbf{x}_2 \in C_2$ . By definition, we have  $\mathbf{x}_1 - \mathbf{x}_2 \in C$ . Thus,  $C$  is a nonempty set.

Now we prove  $C$  is a convex set.

$\forall \mathbf{x}, \mathbf{y} \in C, \forall \theta \in [0,1]$ , by definition,  $\exists \mathbf{x}_1, \mathbf{y}_1 \in C_1, \mathbf{x}_2, \mathbf{y}_2 \in C_2$  s.t.  $\mathbf{x} = \mathbf{x}_1 - \mathbf{x}_2, \mathbf{y} = \mathbf{y}_1 - \mathbf{y}_2$ .

Let  $\bar{\theta} = 1 - \theta$ .

Since  $C_1$  and  $C_2$  are both convex sets,  $\mathbf{z}_1 \triangleq \theta \mathbf{x}_1 + \bar{\theta} \mathbf{y}_1 \in C_1, \mathbf{z}_2 \triangleq \theta \mathbf{x}_2 + \bar{\theta} \mathbf{y}_2 \in C_2$ .

Thus,  $\theta \mathbf{x} + \bar{\theta} \mathbf{y} = \theta \mathbf{x}_1 - \theta \mathbf{x}_2 + \bar{\theta} \mathbf{y}_1 - \bar{\theta} \mathbf{y}_2 = (\theta \mathbf{x}_1 + \bar{\theta} \mathbf{y}_1) - (\theta \mathbf{x}_2 + \bar{\theta} \mathbf{y}_2) = \mathbf{z}_1 - \mathbf{z}_2 \in C$ .

Therefore,  $C$  is a convex set.

In conclusion,  $C$  is a nonempty convex set and  $\mathbf{0} \notin C$ .

*Qed.* ■

## 3. (a) Proof:

$\forall \mathbf{x}_1, \mathbf{x}_2 \in \text{int } C, \forall \theta \in [0,1]$ , Since  $\text{int } C \subset C, \mathbf{x}_1, \mathbf{x}_2 \in C$ .

Since  $C$  is convex,  $\forall \mathbf{x}_1, \mathbf{x}_2 \in C, \forall \theta \in [0,1]$ , let  $\bar{\theta} = 1 - \theta, \mathbf{y} \triangleq \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2 \in C$ .

Since  $\mathbf{x}_1, \mathbf{x}_2 \in \text{int } C, \exists \varepsilon > 0$  s.t.  $B(\mathbf{x}_1, \varepsilon) \subset C, B(\mathbf{x}_2, \varepsilon) \subset C$ .

$\forall \mathbf{z} \in B(\mathbf{y}, \varepsilon), \|\mathbf{z} - \mathbf{y}\| < \varepsilon$ ,

$$\text{i.e. } \mathbf{z} - \mathbf{y} + \mathbf{x}_1 \in B(\mathbf{x}_1, \varepsilon) \subset C, \mathbf{z} - \mathbf{y} + \mathbf{x}_2 \in B(\mathbf{x}_2, \varepsilon) \subset C,$$

$$\mathbf{z} = \mathbf{y} + (\mathbf{z} - \mathbf{y}) = \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2 + (\theta + \bar{\theta})(\mathbf{z} - \mathbf{y}) = \theta(\mathbf{z} - \mathbf{y} + \mathbf{x}_1) + \bar{\theta}(\mathbf{z} - \mathbf{y} + \mathbf{x}_2) \in C.$$

Thus,  $B(\mathbf{y}, \varepsilon) \subset C$ , i.e.  $\mathbf{y} \in \text{int } C$ .

In other words,  $\text{int } C$  is convex.

*Qed.* ■

(b) *Proof:*

$\forall \mathbf{x}_1, \mathbf{x}_2 \in \bar{C}, \forall \theta \in [0,1]$ , let  $\bar{\theta} = 1 - \theta, \mathbf{y} = \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2, \partial C = \bar{C} \setminus \text{int } C \triangleq \{\mathbf{x}: \mathbf{x} \in \bar{C}, \mathbf{x} \notin \text{int } C\}$ .

There are three cases:

**CASE 1.**  $\mathbf{x}_1, \mathbf{x}_2 \in \text{int } C$ . Since  $C$  is convex, from (a) we know  $\mathbf{y} \in \text{int } C \subset \bar{C}$ .

**CASE 2.**  $\mathbf{x}_i \in \text{int } C, \mathbf{x}_j \in \partial C, \{i, j\} = \{1, 2\}$ . Let  $\mathbf{x}_1 \in C, \mathbf{x}_2 \in \partial C$ .

(1) When  $\theta = 0$ ,  $\mathbf{y} = \mathbf{x}_2 \in \bar{C}$ .

(2) When  $\theta \in [0,1)$ , since  $\mathbf{x}_2 \in \partial C$ , for given  $\varepsilon > 0$ , we can always find  $\tilde{\mathbf{x}}_2 \in B(\mathbf{x}_2, \varepsilon)$  s.t.  $\tilde{\mathbf{x}}_2 \in \text{int } C$ . We have  $\|\tilde{\mathbf{x}}_2 - \mathbf{x}_2\| < \varepsilon$ .

From (a) we know  $\theta \mathbf{x}_1 + \bar{\theta} \tilde{\mathbf{x}}_2 \in \text{int } C$  and  $B(\theta \mathbf{x}_1 + \bar{\theta} \tilde{\mathbf{x}}_2, \varepsilon) \subset C \subset \bar{C}$ .

Meanwhile,  $\mathbf{y} = \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2 = (\theta \mathbf{x}_1 + \bar{\theta} \tilde{\mathbf{x}}_2) + \bar{\theta}(\mathbf{x}_2 - \tilde{\mathbf{x}}_2) \in B(\theta \mathbf{x}_1 + \bar{\theta} \tilde{\mathbf{x}}_2, \theta \varepsilon) \subset B(\theta \mathbf{x}_1 + \bar{\theta} \tilde{\mathbf{x}}_2, \varepsilon) \subset \bar{C}$ .

**CASE 3.**  $\mathbf{x}_1, \mathbf{x}_2 \in \partial C$ .

(1) When  $\theta = 0$ ,  $\mathbf{y} = \mathbf{x}_2 \in \bar{C}$ .

(2) When  $\theta = 1$ ,  $\mathbf{y} = \mathbf{x}_1 \in \bar{C}$ .

(3) When  $\theta \in (0,1)$ , since  $\mathbf{x}_1, \mathbf{x}_2 \in \partial C$ , for given  $\varepsilon > 0$ ,

we can always find  $\tilde{\mathbf{x}}_1 \in B(\mathbf{x}_1, \varepsilon), \tilde{\mathbf{x}}_2 \in B(\mathbf{x}_2, \varepsilon)$  s.t.  $\tilde{\mathbf{x}}_1, \tilde{\mathbf{x}}_2 \in \text{int } C$ .

We have  $\|\tilde{\mathbf{x}}_1 - \mathbf{x}_1\| < \varepsilon, \|\tilde{\mathbf{x}}_2 - \mathbf{x}_2\| < \varepsilon$ .

From (a) we know  $\theta \tilde{\mathbf{x}}_1 + \bar{\theta} \tilde{\mathbf{x}}_2 \in \text{int } C$  and  $B(\theta \tilde{\mathbf{x}}_1 + \bar{\theta} \tilde{\mathbf{x}}_2, \varepsilon) \subset C \subset \bar{C}$ .

Meanwhile,  $\mathbf{y} = \theta \mathbf{x}_1 + \bar{\theta} \mathbf{x}_2 = (\theta \tilde{\mathbf{x}}_1 + \bar{\theta} \tilde{\mathbf{x}}_2) + \bar{\theta}(\mathbf{x}_2 - \tilde{\mathbf{x}}_2) + \theta(\mathbf{x}_1 - \tilde{\mathbf{x}}_1)$ .

Since  $\|\bar{\theta}(\mathbf{x}_2 - \tilde{\mathbf{x}}_2) + \theta(\mathbf{x}_1 - \tilde{\mathbf{x}}_1)\| \leq \theta \|\tilde{\mathbf{x}}_1 - \mathbf{x}_1\| + \bar{\theta} \|\tilde{\mathbf{x}}_2 - \mathbf{x}_2\| \leq \theta \varepsilon + \bar{\theta} \varepsilon \leq \varepsilon$ ,  
 $\mathbf{y} \in B(\theta \tilde{\mathbf{x}}_1 + \bar{\theta} \tilde{\mathbf{x}}_2, \varepsilon) \subset \bar{C}$ .

Thus,  $\mathbf{y} \in \bar{C}$ .

In other words,  $\bar{C}$  is convex.

*Qed.* ■

4. (a) *Proof:*

$\forall \mathbf{y}_1, \mathbf{y}_2 \in C$ , by definition we know  $\exists \varphi_1, \dots, \varphi_m, \mu_1, \dots, \mu_m$  s.t.

$$\sum_{i=1}^m \varphi_i \mathbf{x}_i = \mathbf{y}_1, \sum_{i=1}^m \mu_i \mathbf{x}_i = \mathbf{y}_2, \sum_{i=1}^m \varphi_i = 1, \sum_{i=1}^m \mu_i = 1, \varphi_i \geq 0, \mu_i \geq 0 \ (i = 1, 2, \dots, m).$$

$\forall \theta \in [0,1]$ , let  $\bar{\theta} = 1 - \theta$ ,

$$\theta \mathbf{y}_1 + \bar{\theta} \mathbf{y}_2 = \sum_{i=1}^m \theta \varphi_i \mathbf{x}_i + \sum_{i=1}^m \bar{\theta} \mu_i \mathbf{x}_i = \sum_{i=1}^m (\theta \varphi_i + \bar{\theta} \mu_i) \mathbf{x}_i,$$

$$\sum_{i=1}^m \theta \varphi_i + \bar{\theta} \mu_i = \theta \sum_{i=1}^m \varphi_i + \bar{\theta} \sum_{i=1}^m \mu_i = \theta + \bar{\theta} = 1,$$

$$\theta \varphi_i + \bar{\theta} \mu_i \geq 0 \ (i = 1, 2, \dots, m).$$

Thus,  $\theta \mathbf{y}_1 + \bar{\theta} \mathbf{y}_2 \in C$ .

In other words,  $C$  is convex.

*Qed.* ■

(b) *Proof:*

First we prove  $C \subset \mathbf{conv} S$  by contradiction.

If there exists  $x \in C$  s.t.  $x \notin \mathbf{conv} S$ , by definition we know  $x = \sum_{i=1}^m \theta_i x_i$ .

Meanwhile, since  $x_1, \dots, x_m \in \mathbf{conv} S$  and  $\mathbf{conv} S$  is convex, by theorem, convex combination  $x \in \mathbf{conv} S$ . (The theorem will be proved below.)

Contradiction.

Thus,  $\forall x \in C, x \in \mathbf{conv} S$ , i.e.  $C \subset \mathbf{conv} S$ .

By definition we know  $S \subset C$  (let  $\theta_i = \begin{cases} 0, & i \neq k \\ 1, & i = k \end{cases}$  ( $i \in \{1, \dots, m\}$ ), we get  $x_k \in C$ ).

Since  $\mathbf{conv} S$  is the smallest convex set containing  $S$  (by the definition of convex hull) and  $C$  is a convex set containing  $S$ , we know  $\mathbf{conv} S \subset C$ .

Therefore,  $C = \mathbf{conv} S$ .

*Qed.* ■

(The theorem given in the ppt and used above can be proved as follows.)

**Thm.** If  $C$  is convex and  $x_1, \dots, x_m \in C$ , any convex combination  $x = \sum_{i=1}^m \theta_i x_i \in C$ .

*Proof.* Prove the theorem by induction.

We prove

$$y_n = \frac{\sum_{i=1}^n \theta_i x_i}{\sum_{i=1}^n \theta_i} \in C \quad (n \in \{1, \dots, m\}).$$

**BASE STEP.** When  $n = 1$ , obviously  $y_1 = x_1 \in C$ .

When  $n = 2$ , obviously  $y_2 \in C$  (by the definition of convex sets).

**INDUCTIVE STEP.**

Suppose when  $n = k < m$ ,  $y_n \in C$ .

Let  $\bar{\theta}_{k+1} = 1 - \theta_{k+1}$ , then  $\bar{\theta}_{k+1} \geq 0, \theta_{k+1} \geq 0$

(by the definition of convex combination).

Since  $\frac{\sum_{i=1}^k \theta_i}{\sum_{i=1}^{k+1} \theta_i} + \frac{\theta_{k+1}}{\sum_{i=1}^{k+1} \theta_i} = 1$ , by the definition of convex sets, we have

$$\frac{\sum_{i=1}^k \theta_i}{\sum_{i=1}^{k+1} \theta_i} y_k + \frac{\theta_{k+1}}{\sum_{i=1}^{k+1} \theta_i} x_{k+1} \in C.$$

i.e.

$$\begin{aligned} y_{k+1} &= \frac{\sum_{i=1}^{k+1} \theta_i x_i}{\sum_{i=1}^{k+1} \theta_i} = \frac{\sum_{i=1}^k \theta_i x_i + \theta_{k+1} x_{k+1}}{\sum_{i=1}^{k+1} \theta_i} = \frac{\sum_{i=1}^k \theta_i}{\sum_{i=1}^{k+1} \theta_i} \frac{\sum_{i=1}^k \theta_i x_i}{\sum_{i=1}^k \theta_i} + \frac{\theta_{k+1}}{\sum_{i=1}^{k+1} \theta_i} x_{k+1} \\ &= \frac{\sum_{i=1}^k \theta_i}{\sum_{i=1}^{k+1} \theta_i} y_k + \frac{\theta_{k+1}}{\sum_{i=1}^{k+1} \theta_i} x_{k+1} \in C \end{aligned}$$

Thus,  $y_n \in C$  still holds when  $n = k + 1$  ( $n \leq m$ ).

Therefore,  $y_m \in C$ .

Since  $\sum_{i=1}^m \theta_i = 1$  (by the definition of convex combination), we have

$$x = \sum_{i=1}^m \theta_i x_i = \frac{\sum_{i=1}^m \theta_i x_i}{\sum_{i=1}^m \theta_i} = y_m \in C.$$

□

### 5. Proof:

Consider the case of  $\|\mathbf{x} - \mathbf{x}_0\|_2 \leq \|\mathbf{x} - \mathbf{x}_i\|_2$ .

Let  $\mathbf{x} = (y_1, \dots, y_n)$ ,  $\mathbf{x}_i = (x_{i1}, \dots, x_{in})$  ( $i = 0, 1, \dots, K$ ).

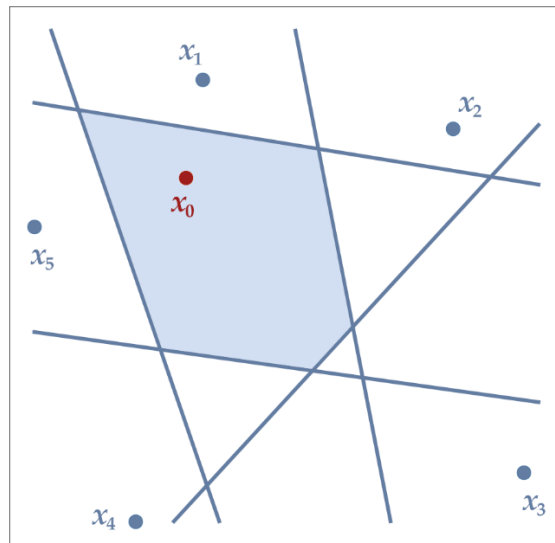
$$\begin{aligned} \|\mathbf{x} - \mathbf{x}_0\|_2 \leq \|\mathbf{x} - \mathbf{x}_i\|_2 &\Leftrightarrow \sqrt{\sum_{j=1}^n (y_j - x_{0j})^2} \leq \sqrt{\sum_{j=1}^n (y_j - x_{ij})^2} \\ \text{可以尝试 } \|\mathbf{x} - \mathbf{x}_0\|_2^2 &= (\mathbf{x} - \mathbf{x}_0)^T (\mathbf{x} - \mathbf{x}_0). \\ &\Leftrightarrow 2 \sum_{j=1}^n (x_{ij} - x_{0j}) y_j \leq 2 \sum_{j=1}^n (x_{ij}^2 - x_{0j}^2) \Leftrightarrow 2(\mathbf{x}_i - \mathbf{x}_0)^T \mathbf{x} \leq \mathbf{x}_i^T \mathbf{x}_i - \mathbf{x}_0^T \mathbf{x}_0 \end{aligned}$$

Thus, we can find

$$\mathbf{A} = \begin{pmatrix} 2(\mathbf{x}_1 - \mathbf{x}_0)^T \\ \vdots \\ 2(\mathbf{x}_K - \mathbf{x}_0)^T \end{pmatrix}, \mathbf{b} = \begin{pmatrix} \mathbf{x}_1^T \mathbf{x}_1 - \mathbf{x}_0^T \mathbf{x}_0 \\ \vdots \\ \mathbf{x}_K^T \mathbf{x}_K - \mathbf{x}_0^T \mathbf{x}_0 \end{pmatrix}$$

s.t.  $V = \{\mathbf{x} : \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$ , i.e.  $V$  is a polyhedron.

A visualization of  $V$  when  $n = 2$  is as follows.



*Qed.* ■