

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

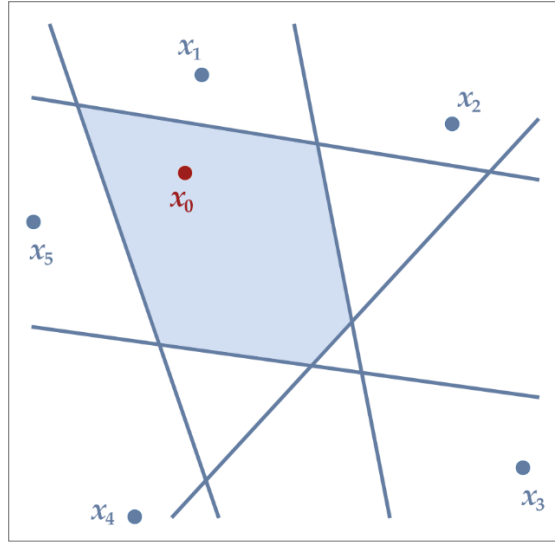
$$\begin{aligned} \|\mathbf{x} - \mathbf{x}_0\|_2 \leq \|\mathbf{x} - \mathbf{x}_i\|_2 &\Leftrightarrow \sqrt{\sum_{j=1}^n (y_j - x_{0j})^2} \leq \sqrt{\sum_{j=1}^n (y_j - x_{ij})^2} \\ &\Leftrightarrow 2 \sum_{j=1}^n (x_{ij} - x_{0j}) y_j \leq \sum_{j=1}^n (x_{ij}^2 - x_{0j}^2) \Leftrightarrow (\mathbf{x}_i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

Thus, we can find

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

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. ■