

Convex Optimization.

1. Convex Sets.

(a) Assume $x, y \in \{x \in \mathbb{R}^n : a_1^T x \leq b_1, a_2^T x \leq b_2\} =: S$

$\forall \theta \in [0, 1]$ check $\theta x + (1-\theta)y \in S$.

$$a_1^T(\theta x + (1-\theta)y) = \theta a_1^T x + (1-\theta) a_1^T y \leq \theta b_1 + (1-\theta) b_1 = b_1$$

$$a_2^T(\theta x + (1-\theta)y) = \theta a_2^T x + (1-\theta) a_2^T y \leq \theta b_2 + (1-\theta) b_2 = b_2$$

Hence $\theta x + (1-\theta)y \in S$.

Hence S is convex.

(b). Assume $x, y \in B(x_0, r)$

$\forall \theta \in [0, 1]$.

$$\begin{aligned} \|\theta x + (1-\theta)y - x_0\| &= \|\theta(x - x_0) + (1-\theta)(y - x_0)\| \\ &\leq \|\theta(x - x_0)\| + \|(1-\theta)(y - x_0)\| \\ &= \theta \|x - x_0\| + (1-\theta) \|y - x_0\| \\ &\leq \theta \cdot r + (1-\theta) r \\ &= r \end{aligned}$$

Hence $\theta x + (1-\theta)y \in B(x_0, r)$

$\Rightarrow B(x_0, r)$ is convex.

(c) For fixed y , the set $C_y := \{x \in \mathbb{R}^n : \|x - x_0\|_2 \leq \|x - y\|_2\}$
 $= \{x \in \mathbb{R}^n : (x - x_0)^T (x - x_0) \leq (x - y)^T (x - y)\}$
 $= \{x \in \mathbb{R}^n : 2(y - x_0)^T x \leq y^T y - x_0^T x_0\}$

is an affine halfspace

$$\begin{aligned} \forall x_1, x_2 \in C := \{x \in \mathbb{R}^n : \|x - x_0\|_2 \leq \|x - y\|_2 \text{ for all } y \in S\} \\ = \{x \in \mathbb{R}^n : 2(y - x_0)^T x \leq y^T y - x_0^T x_0, \forall y \in S\}. \end{aligned}$$

$\forall \theta \in [0, 1]$. check $\theta x_1 + (1-\theta)x_2$.

$$\begin{aligned} \forall y \in S. \quad 2(y - x_0)^T [\theta x_1 + (1-\theta)x_2] &= \theta \cdot 2(y - x_0)^T x_1 + (1-\theta) \cdot 2(y - x_0)^T x_2 \\ &\leq \theta \cdot y^T y - x_0^T x_0 + (1-\theta) \cdot (y^T y - x_0^T x_0) \\ &= y^T y - x_0^T x_0 \end{aligned}$$

Hence $\theta x_1 + (1-\theta)x_2 \in C$.

Hence C is convex.

2. Convex Combinations and Convex Hulls.

(a) By induction.

$k=1$ is trivial. As C contains every its elements.

$k=2 \dots \forall x, y \in C. \quad \theta x + (1-\theta)y \in C, \theta \in [0, 1]$

Assume for $k=n$. It holds.

then $\forall x_1, x_2 \dots x_n$.

$\theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n \in C, \sum_{i=1}^n \theta_i = 1$

Consider $k=n+1, \forall x_1, x_2 \dots x_n, x_{n+1}$

$\forall \lambda \in [0, 1]$. we have

$\lambda x_n + (1-\lambda)x_{n+1} \in C$.

Let $y = \lambda x_n + (1-\lambda)x_{n+1} \in C$.

Consider $x_1, x_2 \dots x_{n+1}, y$.

then $\forall \sum_{i=1}^n \theta_i = 1, \theta_i \geq 0$.

we have by induction

$\theta_1 x_1 + \theta_2 x_2 + \dots + \theta_{n+1} x_{n+1} + \theta_n y \in C$.

$\Leftrightarrow \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_{n+1} x_{n+1} + \lambda \theta_n x_n + (1-\lambda)\theta_n x_{n+1} \in C$.

and we have

$$\sum_{i=1}^{n+1} \theta_i + \lambda \theta_n + (1-\lambda)\theta_n = \sum_{i=1}^n \theta_i = 1$$

Let $\tilde{\theta}_i = \theta_i, i=1, \dots, n-1$

$\tilde{\theta}_n = \lambda \theta_n$

$\tilde{\theta}_{n+1} = (1-\lambda)\theta_n$

$\Rightarrow \sum_{i=1}^{n+1} \tilde{\theta}_i = 1, \tilde{\theta}_i \geq 0$.

Hence. for $k=n+1$, it holds.

By induction. the original statement holds true. \square

(b) Denote $Q := \bigcap \{C : C \text{ is convex and } S \subseteq C\}$.

As Q is intersection of convex sets. $\Rightarrow Q$ is convex.

Also it's obvious. $S \subseteq Q$, as all its element contains S .

$S \subseteq Q$ and $Q \subseteq \text{Conv } S$.

(b) Denote $\mathbb{Q} := \cap \{C : C \text{ is convex and SCC}\}$

As C is convex and SCC, \mathbb{Q} is intersection of all such C_s .
we have $S \subset \mathbb{Q}$, and \mathbb{Q} is convex.

We first show that $\text{ConvS} \subset \mathbb{Q}$

$\forall x \in \text{ConvS}$.

$$x = \sum_{i=1}^k \theta_i x_i, \quad x_i \in S, \quad \theta_i \geq 0, \quad \sum_{i=1}^k \theta_i = 1.$$

Because \mathbb{Q} is convex and $S \subset \mathbb{Q}$

$$\text{by (a) we know } x = \sum_{i=1}^k \theta_i x_i \in \mathbb{Q}.$$

Hence $\text{ConvS} \subset \mathbb{Q}$

Then assume $\exists y \in \mathbb{Q} \setminus \text{ConvS}$.

Let $\tilde{C} = \text{ConvS}$, as $S \subset \tilde{C}$ and \tilde{C} is convex.

~~$\tilde{Q} = \mathbb{Q} \cap \tilde{C}$~~ . Hence \tilde{C} is also in C_s that constructs \mathbb{Q}

Hence $\mathbb{Q} \subset \tilde{C} = \text{ConvS}$

Hence $\exists y \in \mathbb{Q} \setminus \text{ConvS}$

Therefore $\text{ConvS} = \mathbb{Q}$ \square

(c) $1^\circ \text{ConvSC} \mathbb{Q} := \left\{ \sum_{i=1}^k a_i x_i : x_i \in C_i, a_i \geq 0, i=1, \dots, k, \sum_{i=1}^k a_i = 1 \right\}$

By (a). We know $\forall x \in \text{ConvS}$.

$$x = \sum_{i=1}^n \theta_i x_i, \quad x_i \in S, \quad \theta_i \geq 0, \quad \sum_{i=1}^n \theta_i = 1.$$

Wlog, rearrange the order of x_i :

$$\text{Let } x = \sum_{j=1}^K \sum_{x_i \in C_j} \theta_i x_i$$

$$\text{denote } a_j = \sum_{x_i \in C_j} \theta_i. \quad y_j = \frac{1}{a_j} \sum_{x_i \in C_j} \theta_i x_i.$$

$$\text{Because } \sum_j \frac{\sum_{x_i \in C_j} \theta_i}{a_j} = \frac{\sum_{x_i \in C_j} \theta_i}{\sum_{x_i \in C_j} \theta_i} = 1, \quad \frac{\theta_i}{a_j} \geq 0, \quad x_i \in C_j$$

then we have $y_j \in C_j$.

$$\text{Hence } x = \sum_{j=1}^K a_j y_j, \quad \sum_{j=1}^K a_j = \sum_{j=1}^K \sum_{x_i \in C_j} \theta_i = 1. \quad a_j \geq 0. \quad y_j \in C_j$$

Hence $x \in \mathbb{Q} \Rightarrow \text{ConvS} \subset \mathbb{Q}$.

Hence by 1° and 2° .
 $\Rightarrow \text{ConvS} = \mathbb{Q}$.

$2^\circ \mathbb{Q} \subset \text{ConvS}$.

$$\forall x \in \mathbb{Q}, \quad x = \sum_{i=1}^k a_i x_i, \quad \text{as } x_i \in S, \quad \sum a_i = 1, \quad a_i \geq 0.$$

$$\Rightarrow x \in \text{ConvS}, \Rightarrow \mathbb{Q} \subset \text{ConvS}.$$

\square

3. Polar Cone and Separation of Convex Sets

(a) By definition, we have. $\bar{K}^o = \{ s \in \mathbb{R}^n : s^T x \leq 0, \text{ for all } x \in K \}$

We prove the statement by showing $K^o \subseteq \bar{K}^o$ and $\bar{K}^o \subseteq K^o$.

1° $\bar{K}^o \subseteq K^o$.

$\forall s \in \bar{K}^o$. s satisfies. $s^T x \leq 0$, for all $x \in K$.

$$\Rightarrow s^T \left(\sum_{i=1}^m a_i x_i \right) \leq 0, \quad a_i \geq 0.$$

$$\Leftrightarrow \sum_{i=1}^m a_i s^T x_i \leq 0. \quad \forall a_i \geq 0.$$

Let $a = (a_1, \dots, a_m)^T$.

Choose $a = e_j$. then we have $s^T x_j \leq 0 \quad j=1, 2, \dots, m$

Hence $s \in K^o$.

$$\Rightarrow \bar{K}^o \subseteq K^o$$

2° $K^o \subseteq \bar{K}^o$

$\forall s \in K^o$. we have $s^T x_j \leq 0$, for $j=1, \dots, m$

$$\Rightarrow \forall a_j \geq 0, \quad a_j s^T x_j \leq 0$$

$$\Rightarrow \sum_{j=1}^m a_j s^T x_j \leq 0$$

Hence $s \in \bar{K}^o$.

$$\Rightarrow K^o \subseteq \bar{K}^o$$

Hence, we conclude that $\bar{K}^o = K^o$. \square

(b). $K^o = \{ s \in \mathbb{R}^n : s^T x \leq 0, \text{ for all } x \in K \}$

We first show that K^o is always closed.

As K^o is the intersection of closed halfspaces, so K^o is closed.

Hence, K^{oo} is also closed.

1° show $C \cap K \subseteq K^{oo}$.

$\forall x \in K$. $\forall s \in K^o$. $s^T x \leq 0$.

$$\Rightarrow x \in K^{oo}$$

$$\Rightarrow C \cap K \subseteq K^{oo}$$

As K^{oo} is closed. $\Rightarrow C \cap K \subseteq K^{oo}$

2° If $\exists x_0 \in K^{oo} \setminus C \cap K$. we show by contradiction that if $x_0 \notin C \cap K$, then $x_0 \notin K^{oo}$

Because $x_0 \notin C \cap K$. by strict separation theorem. (Because $C \cap K$ is closed and convex)

\exists halfplane $a^T x = b$ such that $a^T x_0 > b$, $a^T x \leq b, \forall x \in C \cap K$.

As $0 \in K$. $\Rightarrow b \geq 0$

Because $K^o = \{ s \in \mathbb{R}^n : s^T x \leq 0, \text{ for all } x \in K \}$.

We can exactly take one of its supporting planes as the separating hyperplane.
Otherwise $x_0 \in K$.

Say. $s_i^T x = 0, s_i \in K^\circ$.

Hence we have $\forall x \in K, s_i^T x \leq 0$,

$$s_i^T x_0 > 0.$$

As $s_i \in K^\circ$, we then conclude $x_0 \notin K^\circ \quad \square$

Hence. $C\ell K = K^\circ$

- To prove $C\ell K$ is closed and convex.

Closeness is trivial.

To show convexity, because K is convex. $\forall x, y \in K, \forall \theta \in [0, 1], \theta x + (1-\theta)y \in K$.

For any $x, y \in C\ell K$. Construct two sequences $\{x_i\}, \{y_i\}$.

which satisfies. $\lim_{n \rightarrow \infty} x_n = x, \lim_{n \rightarrow \infty} y_n = y$.

and $x_i \in K, y_i \in K$.

Hence. $\theta x_i + (1-\theta)y_i = z_i \in K \subseteq C\ell K$.

Take the limit, and because $C\ell K$ is closed.

We have $\theta x + (1-\theta)y = z \in C\ell K$.

$\Rightarrow C\ell K$ is convex \square .

(c). Using results of (a) and (b)

$$K^\circ = C\ell K.$$

$$\begin{aligned} \text{Hence } \{s \in \mathbb{R}^n : s^T x_j \leq 0 \text{ for } j=1, \dots, m\}^\circ &= C\ell K \\ &= C\ell \cdot \left\{ \sum_{j=1}^m a_j x_j : a_j \geq 0 \text{ for } j=1, \dots, m \right\} \\ &= \left\{ \sum_{j=1}^m a_j x_j : a_j \geq 0 \text{ for } j=1, \dots, m \right\} \\ &= K. \end{aligned}$$

4. Normal cone and Tangent Cone

$$(a) N(x) = \{s \in \mathbb{R}^n \mid s^T(y-x) \leq 0 \text{ for all } y \in C\}$$

Because $N(x)$ is the intersection of closed convex halfspaces.

so $N(x)$ is closed, convex

if $s \in N(x)$, then $\forall \theta \geq 0 \quad (\theta s)^T(y-x) = \theta s^T(y-x) \leq 0, \text{ for all } y \in C$

Hence $N(x)$ is a cone.

$\Rightarrow N(x)$ is a closed convex cone.

(b) If $x \in \text{int } C$, then $N_C(x) = 0$. $T_C(x) = N_C^\circ(x) = \mathbb{R}^n$.

If $x \in \text{Bd } C$.

denote the tight constraints at point x as $J(x) = \{j : s_j^T x = r_j, j=1, \dots, m\}$

Let $N_C(x) = \left\{ \sum_{j \in J(x)} a_j s_j, a_j \geq 0 \right\}$

We show this is the normal cone at the point x .

For one single tight constraint $s_j^T x = r_j$,

the corresponding normal cone vector is s_j . as

$$s_j^T(y-x) \leq r_j - r_j = 0.$$

When the tight constraint increase by 1, the generator increases by s_j as well.

And
$$\begin{aligned} N_C(x) &= \left(\sum_{j \in J(x)} a_j s_j \right)^T (y-x), \quad \text{for all } y \in C \\ &= \sum_{j \in J(x)} [a_j s_j^T (y-x)] \\ &\leq \sum_{j \in J(x)} a_j (r_j - r_j) \\ &= 0. \end{aligned}$$

Hence $N_C(x) = \left\{ \sum_{j \in J(x)} a_j s_j, a_j \geq 0 \right\}$ is the description of the normal cone.

From 3(a) we know

$$T_C(x) = N_C^\circ(x) = \left\{ y \in \mathbb{R}^n : y^T s_j \leq 0 \quad \text{for } j \in J(x) \right\}.$$