

Ex1 Prove that the quadratic cone is convex

proof: quadratic cone: $C = \{(x, t) \in \mathbb{R}^{n+1} \mid \|x\|_2 \leq t\}$
 $= \{[x]^\top [t] \mid [x]^\top \begin{bmatrix} I & 0 \\ 0 & -1 \end{bmatrix} [x] \leq 0, t \geq 0\}$

suppose $c_1, c_2 \in C$ $c_1 = \begin{bmatrix} x_1 \\ t_1 \end{bmatrix}$ $c_2 = \begin{bmatrix} x_2 \\ t_2 \end{bmatrix}$

Want to Show: $\theta c_1 + (1-\theta) c_2 \in C$ for $\theta \in (0, 1)$ firstly as $\theta \in (0, 1)$, $t_1, t_2 \geq 0$

it's clear $\theta t_1 + (1-\theta) t_2 \geq 0$

consider $\begin{bmatrix} \theta x_1 + (1-\theta) x_2 \\ \theta t_1 + (1-\theta) t_2 \end{bmatrix}^\top \begin{bmatrix} I & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} \theta x_1 + (1-\theta) x_2 \\ \theta t_1 + (1-\theta) t_2 \end{bmatrix}$ $\xrightarrow{(*)}$

$$\begin{aligned} &= (\theta x_1 + (1-\theta) x_2)^\top (\theta x_1 + (1-\theta) x_2) - (\theta t_1 + (1-\theta) t_2)^2 \\ &= \underbrace{\theta^2 x_1^\top x_1}_{(I)} - \underbrace{\theta^2 t_1^2}_{(II)} + \underbrace{(1-\theta)^2 x_2^\top x_2}_{(III)} - (1-\theta)^2 t_2^2 + 2\theta(1-\theta) x_1^\top x_2 - 2\theta(1-\theta) t_1 t_2. \end{aligned}$$

As $\begin{bmatrix} x_1 \\ t_1 \end{bmatrix}, \begin{bmatrix} x_2 \\ t_2 \end{bmatrix} \in C$ so $(I) \leq 0$ $(II) \leq 0$

Consider Cauchy-Schwarz Inequality:

$$|\langle x_1, x_2 \rangle| = |x_1^\top x_2| \leq \|x_1\|_2 \|x_2\|_2 \Leftrightarrow |x_1^\top x_2| \leq t_1 t_2$$

$$\text{so } 2\theta(1-\theta)(x_1^\top x_2 - t_1 t_2) \leq 0 \Rightarrow (*) \leq 0$$

so $\theta c_1 + (1-\theta) c_2 \in C$ which means C is convex

Ex2 Prove that the intersection of two convex sets is convex

proof: Suppose there are two convex sets C_1 & C_2

choose $x_1, x_2 \in C_1 \cap C_2$ consider $\theta x_1 + (1-\theta) x_2$ $\theta \in (0, 1)$

$x_1, x_2 \in C_1 \cap C_2$ means $x_1, x_2 \in C_1$ and $x_1, x_2 \in C_2$

As C_1, C_2 are convex so $\theta x_1 + (1-\theta) x_2 \in C_1$ and $\theta x_1 + (1-\theta) x_2 \in C_2$

so $\theta x_1 + (1-\theta) x_2 \in C_1 \cap C_2$ when $\theta \in (0, 1)$

$\Rightarrow C_1 \cap C_2$ is still convex

Ex 3 Prove that the image of a convex set under an affine function is convex, and that the inverse image is also convex

proof: Suppose $f(x) = Ax + b$ is an affine function and C is a convex set.

consider $f(C) = \{f(x) \mid x \in C\}$

suppose $x_1, x_2 \in C$, $f(x_1), f(x_2) \in f(C)$, as C is convex so

$$\theta x_1 + (1-\theta)x_2 \in C \text{ for } \theta \in (0,1) \Rightarrow f(\theta x_1 + (1-\theta)x_2) \in f(C)$$

$$f(\theta x_1 + (1-\theta)x_2) = A(\theta x_1 + (1-\theta)x_2) + b$$

$$= \theta Ax_1 + \theta b + (1-\theta)Ax_2 + (1-\theta)b$$

$$= \theta f(x_1) + (1-\theta)f(x_2) \in f(C) \text{ for } \theta \in (0,1)$$

so $f(C)$ is also convex.

Next, I want to show the inverse image is also convex

$$f^{-1}(C) = \{x \mid f(x) \in C\}$$

Want to show $\theta x_1 + (1-\theta)x_2 \in f^{-1}(C)$ if $f(x_1) \in C$, $f(x_2) \in C$ and $\theta \in (0,1)$

As C is convex so $\theta f(x_1) + (1-\theta)f(x_2) \in C$

$$\text{i.e. } \theta f(x_1) + (1-\theta)f(x_2) \in C$$

$$\Leftrightarrow \theta(Ax_1 + b) + (1-\theta)(Ax_2 + b) \in C$$

$$\Leftrightarrow A(\theta x_1 + (1-\theta)x_2) + b \in C$$

$$\Leftrightarrow f(\theta x_1 + (1-\theta)x_2) \in C \Leftrightarrow \theta x_1 + (1-\theta)x_2 \in f^{-1}(C)$$

so $f^{-1}(C)$ is also convex

Ex 4 2.1 Let $C \subseteq \mathbf{R}^n$ be a convex set, with $x_1, \dots, x_k \in C$, and let $\theta_1, \dots, \theta_k \in \mathbf{R}$ satisfy $\theta_i \geq 0$, $\theta_1 + \dots + \theta_k = 1$. Show that $\theta_1 x_1 + \dots + \theta_k x_k \in C$. (The definition of convexity is that this holds for $k = 2$; you must show it for arbitrary k .) Hint. Use induction on k .

proof: when $k=2$, if C is a convex set if $\theta_1 + \theta_2 = 1$

$$\theta_1 x_1 + \theta_2 x_2 \in C$$

Suppose that $\sum_{i=1}^k \theta_i x_i \in C$ if $\sum_{i=1}^{k-1} \theta_i = 1$ and $x_1, \dots, x_k \in C$

Now want to show $\sum_{i=1}^{k+1} \theta_i x_i \in C$ if $\sum_{i=1}^{k+1} \theta_i = 1$

$$y = \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k + \theta_{k+1} x_{k+1}$$

$$= (1 - \theta_{k+1}) \left(\frac{\theta_1}{1 - \theta_{k+1}} x_1 + \frac{\theta_2}{1 - \theta_{k+1}} x_2 + \dots + \frac{\theta_k}{1 - \theta_{k+1}} x_k \right) + \theta_{k+1} x_{k+1}$$

Recall that $\sum_{i=1}^k \frac{\theta_i}{1 - \theta_{k+1}}$

$$= \frac{\sum_{i=1}^k \theta_i}{1 - \theta_{k+1}} = \frac{1 - \theta_{k+1}}{1 - \theta_{k+1}} = 1$$

and as $\theta_i \in (0, 1)$ $i=1, \dots, k+1$ $\theta_i < 1 - \theta_{k+1}$ and $\theta_i > 0$ $1 - \theta_{k+1} > 0$

$$\text{so } \frac{\theta_i}{1 - \theta_{k+1}} \in (0, 1) \quad i=1, \dots, k$$

By assumption, $\frac{\theta_1}{1 - \theta_{k+1}} x_1 + \frac{\theta_2}{1 - \theta_{k+1}} x_2 + \dots + \frac{\theta_k}{1 - \theta_{k+1}} x_k \in C$

since C is convex so $\theta_{k+1} y_{k+1} + (1 - \theta_{k+1}) \left(\frac{\theta_1}{1 - \theta_{k+1}} x_1 + \dots + \frac{\theta_k}{1 - \theta_{k+1}} x_k \right) \in C$

QED.

Ex5 2.2 Show that a set is convex if and only if its intersection with any line is convex. Show that a set is affine if and only if its intersection with any line is affine.

Proof: Part 1: " \Rightarrow " if a set is convex then its intersection with any line is convex

recall that a line is convex as any $x_1, x_2 \in l : x_0 + t\vec{r} \quad \theta x_1 + (1 - \theta)x_2 \in l$.

when $\theta \in (0, 1)$ so the intersection of two convex sets is also convex

" \Leftarrow " Suppose that the intersection of a set C and any line $l_{x_0, \vec{r}}$ is convex
i.e. $C \cap l_{x_0, \vec{r}}$ is convex and $l_{x_0, \vec{r}} = x_0 + t\vec{r} \quad t \in \mathbb{R}$.

now choose $x_1, x_2 \in C \quad \theta x_1 + (1 - \theta)x_2 = x_2 + \theta(x_1 - x_2) \in l_{x_1, \vec{r}}$ where $\theta \in (0, 1)$

$\vec{r} = x_1 - x_2$ which means we can find a line $l_{x_1, \vec{r}} = x_1 + t\vec{r}$ contains x_1, x_2

so $x_1, x_2 \in C \cap l_{x_1, \vec{r}}$ as $C \cap l_{x_1, \vec{r}}$ is convex so $\theta x_1 + (1 - \theta)x_2 \in C \cap l_{x_1, \vec{r}}$

$\Rightarrow \theta x_1 + (1 - \theta)x_2 \in C \quad \theta \in (0, 1)$ so C is convex

Part (2): We use the same idea, firstly prove the intersection of two affine sets is affine. Assume C_1 and C_2 are affine. so choose $x_1, x_2 \in C_1 \cap C_2$
 $\theta x_1 + (1-\theta)x_2 \in C_1, \theta \in (0,1)$ as C_1 is affine and $x_1, x_2 \in C_1$.
also $\theta x_1 + (1-\theta)x_2 \in C_2, \theta \in (0,1)$ as C_2 is affine
so $\theta x_1 + (1-\theta)x_2 \in C_1 \cap C_2$ i.e. $C_1 \cap C_2$ is affine
it's clear any line l is affine so for any affine set C , $C \cap l$ is also affine.
" \Leftarrow ". suppose for a set C , $C \cap l$ is affine for any line l
choose $x_1, x_2 \in C$ $\theta x_1 + (1-\theta)x_2 = x_2 + \theta(x_1 - x_2), \theta \in \mathbb{R}$.
so consider the line $l: x_2 + t(x_1 - x_2), t \in \mathbb{R}, x_1, x_2 \in l$
so $x_1, x_2 \in C \cap l \Rightarrow \theta x_1 + (1-\theta)x_2 \in C \cap l$ as $C \cap l$ is affine
so $\theta x_1 + (1-\theta)x_2 \in C$ for $\theta \in \mathbb{R}$ so C is affine

Ex 6 Solution set of a quadratic inequality. Let $C \subseteq \mathbf{R}^n$ be the solution set of a quadratic inequality,

$$C = \{x \in \mathbf{R}^n \mid x^T A x + b^T x + c \leq 0\},$$

with $A \in \mathbf{S}^n$, $b \in \mathbf{R}^n$, and $c \in \mathbf{R}$.

- (a) Show that C is convex if $A \succeq 0$.
- (b) Show that the intersection of C and the hyperplane defined by $g^T x + h = 0$ (where $g \neq 0$) is convex if $A + \lambda gg^T \succeq 0$ for some $\lambda \in \mathbf{R}$.

Are the converses of these statements true?

proof: (a) Use the conclusion in Ex.5. I want to prove for any line l $C \cap l$ is convex

consider the line $l = \{x \mid x = x_0 + t\vec{r}, t \in \mathbb{R}\}$

$$\begin{aligned} \text{so } l \cap C &= \left\{ t \in \mathbb{R} \mid (x_0 + t\vec{r})^T A (x_0 + t\vec{r}) + b^T (x_0 + t\vec{r}) + c \leq 0 \right\} \\ &= \left\{ t \in \mathbb{R} \mid \vec{r}^T A \vec{r} t^2 + (2x_0^T A \vec{r} + b^T \vec{r}) t + (x_0^T A x_0 + b^T x_0 + c) \leq 0 \right\}. (*) \end{aligned}$$

This is a quadratic function. consider any $f(x) = ax^2 + bx + c \leq 0$ ($a > 0$)
as $f''(x) = 2a > 0$ so $f(x)$ is always convex when $a > 0$

so the epigraph $\{x | f(x) \leq 0\}$ is always a convex set
when $a=0$ $f(x)$ degenerates to a line, which is affine, so also convex
so the epigraph $\{x | f(x) \leq 0\}$ when $a \geq 0$ is also convex.

go back to equation (*) as $A \succeq 0$ so $r^T A r \geq 0$ for any \vec{r}
which means the parameter of second-order term is nonnegative
so $C \cap H$ is convex for any line H

use the conclusion in Ex. 5. $\Rightarrow C$ is convex

converse of the statement is false: counterexample consider $x \in \mathbb{R}$. $\{x | -x^2 \leq 0\} = \mathbb{R}$.

It's clear \mathbb{R} is convex but $A = -1 \leq 0$ so $A \succeq 0$ doesn't hold but $\{x | -x^2 \leq 0\}$ is convex

b) consider the explicit form of $C \cap H$ where $H = \{x | g^T x + h = 0\}$
choose $x_0 \in H$. let $x = x_0 + t\vec{r}$ where $t \in \mathbb{R}$ \vec{r} is any direction

if x in H $g^T(x_0 + t\vec{r}) + h = 0$

$$\Leftrightarrow g^T x_0 + h + t g^T \vec{r} = 0 \quad g^T x_0 + h = 0 \quad \text{as } x_0 \in H$$

$$\Leftrightarrow t g^T \vec{r} = 0$$

so I have two cases (i) $g^T \vec{r} \neq 0 \Rightarrow t = 0$

so $C \cap H = \{x_0\}$ this is the trivial case

$$\& x_0 + (1-\theta)x_0 = x_0 \in C \cap H$$

(ii) $g^T \vec{r} = 0$ the t could be any real value

$$\text{so } C \cap H = \{x_0 + t\vec{r} \mid \vec{r}^T A \vec{r} t^2 + (2x_0^T A \vec{r} + b^T \vec{r})t + (x_0^T A x_0 + b^T x_0 + c) \leq 0\}$$

by the conclusion in (a) we know that if $\vec{r}^T A \vec{r} \geq 0$ then $C \cap H$ is convex
as we can find $\lambda \in \mathbb{R}$ such that $A + \lambda gg^T \succeq 0$

$$\text{so } \vec{r}^T (A + \lambda gg^T) \vec{r} = \vec{r}^T A \vec{r} + \underbrace{\lambda \vec{r}^T g g^T \vec{r}}_{\text{as } g^T \vec{r} = 0} = \vec{r}^T A \vec{r} \geq 0$$

so $C \cap H$ is convex.

However, the converse is false, as we can get from the above proof

$r^T A r \geq 0$ only when $g^T r = 0$ not for all r . If $g^T r \neq 0$, we can not get any conclusion of A as $H \cap C$ is a single set. So the converse of the statement is false.

Ex7 2.16 Show that if S_1 and S_2 are convex sets in \mathbf{R}^{m+n} , then so is their partial sum

$$S = \{(x, y_1 + y_2) \mid x \in \mathbf{R}^m, y_1, y_2 \in \mathbf{R}^n, (x, y_1) \in S_1, (x, y_2) \in S_2\}.$$

Proof: choose $(x, y_1 + y_2), (\mu, v_1 + v_2) \in S$

which means $(x, y_1), (\mu, v_1) \in S_1$,

$$(x, y_1), (\mu, v_1) \in S_1$$

as S_1, S_2 are convex so $\theta(x, y_1) + (1-\theta)(\mu, v_1)$

$$= (\theta x + (1-\theta)\mu, \theta y_1 + (1-\theta)v_1) \in S_1$$

Similarly $\theta(x, y_2) + (1-\theta)(\mu, v_2)$

$$= (\theta x + (1-\theta)\mu, \theta y_2 + (1-\theta)v_2) \in S_2$$

so $\theta(x, y_1 + y_2) + (1-\theta)(\mu, v_1 + v_2)$

$$= (\theta x + (1-\theta)\mu, \theta y_1 + (1-\theta)v_1 + (1-\theta)(v_1 + v_2))$$

$$= (\theta x + (1-\theta)\mu, \theta y_1 + (1-\theta)v_1 + \theta y_2 + (1-\theta)v_2) \in S$$

as $(\theta x + (1-\theta)\mu, \theta y_1 + (1-\theta)v_1) \in S_1$,

$$(\theta x + (1-\theta)\mu, \theta y_2 + (1-\theta)v_2) \in S_2$$

so S is convex

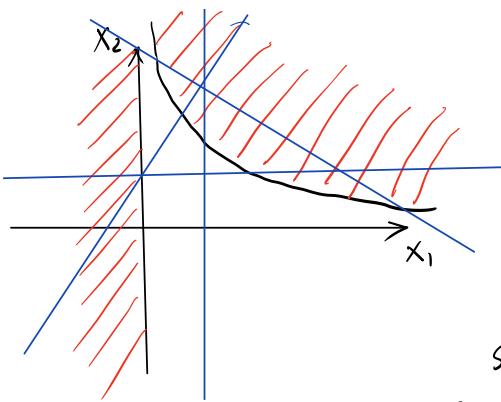
Ex 8 2.23 Give an example of two closed convex sets that are disjoint but cannot be strictly separated.

Consider two set $S_1 = \{(x_1, x_2) \mid x_1 \leq 0\}$ $S_2 = \{(x_1, x_2) \mid x_1 x_2 \geq 1, x_1 > 0, x_2 > 0\}$

Firstly show S_1 and S_2 are convex choose $x, y \in S_1$ $x = (x_1, x_2)$ $y = (y_1, y_2)$

$$\theta x + (1-\theta)y = (\theta x_1 + (1-\theta)y_1, \theta x_2 + (1-\theta)y_2) \text{ as } \theta \in (0, 1) \quad x_1, y_1 \leq 0$$

$$\text{so } \theta x_1 \leq 0 \quad (1-\theta)y_1 \leq 0 \quad \text{so } \theta x_1 + (1-\theta)y_1 \leq 0 \Rightarrow \theta x + (1-\theta)y \in S_1$$



S_2 is also convex

$$\text{consider } f(x) = \frac{1}{x} \quad (x > 0)$$

$$f''(x) = \frac{2}{x^3} > 0 \quad \text{when } x > 0$$

so $f(x) = \frac{1}{x}$ is a convex function

so for $y > 0$ $\{(x_1, y) \mid \frac{1}{x_1} \leq y\}$ is the epigraph of $f(x)$

so $\{(x_1, y) \mid \frac{1}{x_1} \leq y\}$ is convex because $f(x)$ is convex

which means $S_2 = \{(x_1, x_2) \mid x_1, x_2 \geq 1, x_1 > 0, x_2 > 0\}$ is convex

It's clear, $S_1 \cap S_2 = \emptyset$ as $x_1, x_2 > 0$ in S_2 and $x_1 \leq 0$ in S_1 .

It's obvious S_1 and S_2 are closed as it's easy to show S_1^c, S_2^c are open.

Now for any line $l: x_2 = ax_1 + b$, if l is not vertical, we can always find enough big y_1 such that for (x_1, y_1) $ax_1 + b < y_1$,

and enough small y_2 $ax_2 + b > y_2$ and $(x_1, y_1), (x_2, y_2) \in S_1$,

so l cannot be separating hyperplane

for vertical line $l: x = a$ if $a = 0$ this not the strict separating

hyperplane. if $a > 0$ $x = a$ and $f(x) = \frac{1}{x}$ always has intersection

if $a < 0$ $x = a \in S_1$

so we never find the strict separating hyperplane

- Ex 9 (b) Let $C = \{x \in \mathbf{R}^n \mid \|x\|_\infty \leq 1\}$, the ℓ_∞ -norm unit ball in \mathbf{R}^n , and let \hat{x} be a point in the boundary of C . Identify the supporting hyperplanes of C at \hat{x} explicitly.

Solution: I want to find a such that $a^\top x \leq a^\top \hat{x}$

Suppose \hat{x} is on the boundary, so there is at least one element of \hat{x} is 1 or -1

suppose \hat{x} only have one element with absolute value 1.

$|\hat{x}_i| = 1$ and $-1 < \hat{x}_i < 1$ for all $i \neq i$, case (i) $\hat{x}_i = 1$ for $x_i > 0$ and

$$a_j = 0 \quad j \neq i \quad a^T \hat{x} = a_i \geq a^T x = a_i x_i \text{ for all } x \in C$$

as if $x \in C$ $\|x\|_\infty \leq 1$ $x_i \leq 1$ so $a^T = (0, 0, \dots, \underset{i}{a_i}, 0, \dots, 0)$ $a^T x$ is
i-th element, $a_i > 0$

the supporting hyperplane

case (ii) $\hat{x}_i = -1$ when $a_i < 0$ and $a_j = 0$ for $j \neq i$

$a^T x = -a_i \geq a^T x = a_i x_i$ for all $x \in C$ as $-1 \leq x_i$ so the supporting hyperplane
is $a^T x$ $a^T = (0, \dots, \underset{i}{a_i}, 0, \dots, 0)$
i-th element $a_i < 0$

Next for the general case . if \hat{x} has j elements with absolute value 1

$\hat{x} = (x_{i_1}, x_{i_2}, \dots, x_{i_k}, \dots, x_{i_j})$ where $|x_{i_k}| = 1$ when $i_k \in \{i_1, \dots, i_j\}$

$-1 < x_{i_k} < 1$ when $i_k \in \{1, \dots, n\} \setminus \{i_1, \dots, i_j\}$

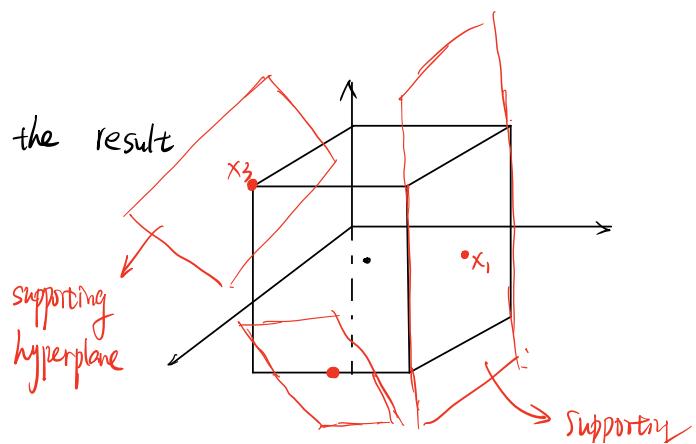
we only need to assign the element of a^T as

$$\begin{cases} a_i > 0 & \text{if } \hat{x}_i = 1 \quad \text{so } a^T \hat{x} = \sum_{k=1}^j |a_i| \geq a^T x = \sum_{k=1}^j a_{i_k} x_{i_k} \text{ when} \\ a_i < 0 & \text{if } \hat{x}_i = -1 \\ a_i = 0 & \text{if } -1 < \hat{x}_i < 1 \quad \text{as if } \|x\|_\infty \leq 1 \quad -1 \leq x_i \leq 1 \text{ for every } i=1, \dots, n. \end{cases}$$

so the supporting hyperplane $a^T x$ is

$$\begin{cases} a_i > 0 & \text{if } \hat{x}_i = 1 \\ a_i < 0 & \text{if } \hat{x}_i = -1 \\ a_i = 0 & \text{if } -1 < \hat{x}_i < 1 \end{cases}$$

I plot a figure to illustrate the result



Ex 10 Verify that as stated on BV p.39, the hyperplane cone is the inverse image of the second order cone under the given affine transformation

II O
hyperplane.

solution : C_1 : second order cone: $\{x \mid x^T P x \leq (C^T x)^2, C^T x \geq 0\} \quad P \in S_+^n \quad C \in \mathbb{R}^n$

C_2 : second order cone: $\{(z, t) \mid z^T z \leq t^2, t \geq 0\}$

affine function: $f(x) = (P^{1/2}x, C^T x)$

The inverse image of the second order cone C_2 .

$\{x \mid f(x) \in C_2\}$

$$\Leftrightarrow \{x \mid (P^{1/2}x, C^T x) \in C_2\} \Leftrightarrow \{x \mid (P^{1/2}x)^T (P^{1/2}x) \leq (C^T x)^2, C^T x \geq 0\} \\ \Leftrightarrow \{x \mid x^T P x \leq (C^T x)^2, C^T x \geq 0\} = C_1$$