

# Convex Sets

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**Definition 1.** Convex set:  $C \subset \mathbb{R}^n$  such that if  $x, y \in C$  then  $tx + (1 - t)y \in C$  for all  $0 \leq t \leq 1$ .

**Definition 2.** Convex combination of  $x_1, \dots, x_k \in \mathbb{R}^n$  is any linear combination

$$\theta_1 x_1 + \dots + \theta_k x_k$$

with  $\theta_i \geq 0$   $i = 0, \dots, k$ , and  $\sum_{i=1}^k \theta_i = 1$ .

**Definition 3.** Convex hull of a set  $C$ , denoted  $\text{conv}(C)$ , is the set of all the convex combinations of elements. It is the smallest convex set that contains set  $C$  (note  $C$  is not necessarily convex).

Examples of convex sets:

- Norm ball:  $\{x : \|x\| \leq r\}$ . It is a convex set by using the triangular inequality of the norm

$$\begin{aligned} \|tx + (1 - t)y\| &\leq t\|x\| + (1 - t)\|y\| \\ &\leq r \end{aligned}$$

- Affine space:

$$Cx = d \iff Cx \leq d - Cx \leq -d$$

- Polyhedron:

$$\{x : Ax \leq b\} \iff \{x : a_i^T x \leq b_i \mid i = 1, \dots, m\}$$

$a_i$  is the  $i$ -th row of  $A$  (note: this is an intersection of  $m$  halfspaces);

- This definition generalizes to  $\{x : Ax \leq b, Cx = d\}$  since we can rewrite the  $Cx = d$  constraint using inequality constraints.

- Simplex: special case of polyhedra, given by

$$\text{conv}\{x_0, \dots, x_k\}$$

where these points are affinely independent. The canonical example is the probability simplex

$$w^T e = \text{conv}\{e_1, \dots, e_n\} = \{w : w \geq 0, 1^T w = 1\}$$

where

$$e = [e_1, e_2, \dots, e_n]$$

where

$$e_1 = (1, 0, \dots, 0)^T$$

where  $e_1, \dots, e_n$  are the standard basis vectors in  $\mathbb{R}^n$  and  $w \in \mathbb{R}^n$ .

- **Note:**  $x_0, \dots, x_k$  are affine independent  $\iff x_1 - x_0, \dots, x_k - x_0$  are linearly independent.

**Definition 4.** Cone  $C \subset \mathbb{R}^n$  such that  $x \in C \implies tx \in C$  for all  $t \geq 0$ .

- **Note:** 0 must be lies in the cone.

**Definition 5.** Normal cone:  $N_c(x)$  is a normal cone to set  $C$  at the point  $x \in C$  which satisfies

$$N_c(x) = \{g : \langle g, y - x \rangle \leq 0 \text{ for all } y \in C\}$$

**Proposition 1.** *Normal cone is convex cone.*

*Proof.* We show  $N_c(x)$  is a cone and convex □

1. To show  $N_c(x)$  is a cone

- (a) Fix any  $g \in N_c(x)$  and  $t \geq 0$
- (b) By definition

$$\langle g, y - x \rangle \leq 0 \text{ for all } y \in C$$

then

$$\langle tg, y - x \rangle = t \langle g, y - x \rangle \leq 0 \text{ for all } y \in C$$

Thus  $tg \in N_c(x)$ . Therefore  $N_c(x)$  is a cone

2. To show  $N_c(x)$  is a convex set  $g_1, g_2 \in N_c(x)$ , we want to show  $tg_1 + (1 - t)g_2 \in N_c(x)$

- (a) Fix  $g_1, g_2 \in N_c(x)$  and  $t \in [0, 1]$

$$\langle g_1, y - x \rangle \leq 0 \text{ for all } y \in C$$

$$\langle g_2, y - x \rangle \leq 0 \text{ for all } y \in C$$

Thus

$$\begin{aligned}\langle tg_1 + (1-t)g_2, y-x \rangle &= t \langle g_1, y-x \rangle + (1-t) \langle g_2, y-x \rangle \\ &\leq 0 \text{ for all } y \in C\end{aligned}$$

Therefore

$$tg_1 + (1-t)g_2 \in N_c(x)$$

Thus  $N_c(x)$  is a convex set.

**Note:**

- For  $N_c(x)$ , set  $C$  can be any set, not necessarily convex
- $N_c(x) = \{0\}$  for any  $x$  inside  $C$

Basic linear algebra facts:

- $X \in \mathbb{S}^n \implies \lambda(X) \in \mathbb{R}^n$
- $X \in \mathbb{S}_+^n \implies \lambda(X) \in \mathbb{R}_+^n$
- $X \in \mathbb{S}_{++}^n \implies \lambda(X) \in \mathbb{R}_{++}^n$

We can define an inner product over  $\mathbb{S}^n$ : given  $X, Y \in \mathbb{S}^n$ ,

$$\langle X, Y \rangle = \text{tr}(XY)$$

We can define a partial ordering over  $\mathbb{S}^n$ : given  $X, Y \in \mathbb{S}^n$ ,

$$X \succeq Y \iff X - Y \in \mathbb{S}_+^n$$

**Proposition 2.** *The set of all positive semidefinite matrices  $x \in \mathbb{S}_+^n$  is a convex cone*

1. If  $x \in \mathbb{S}_+^n$  then  $tx \in \mathbb{S}_+^n$  for  $t \geq 0$
2.  $x, y \in \mathbb{S}_+^n$  then  $tx + (1 - t)y \in \mathbb{S}_+^n$  for  $t \geq 0$

**Proposition 3.** *The set of all the points  $x$  that satisfies*

$$\{x : x_1 A_1 + \cdots x_k A_k \preceq B\}$$

*is a convex set.*

*Proof.* Method 1: Let  $C = \{x : x_1 A_1 + \cdots x_k A_k \preceq B\}$ .

Assume  $\forall x, y \in C$  and let

$$z = tx + (1 - t)y, t \in [0, 1]$$

We want to show  $z \in C$

$$\begin{aligned} & B - (z_1 A_1 + \cdots + z_k A_k) \\ &= B - ((tx_1 + (1 - t)y_1)A_1 + \cdots + (tx_k + (1 - t)y_k)A_k) \\ &= t(B - (x_1 A_1 + \cdots + x_k A_k)) + (1 - t)(B - (y_1 A_1 + \cdots + y_k A_k)) \\ &\succeq 0 \end{aligned}$$

Method 2:

$$\{x : x_1 A_1 + \cdots x_k A_k \preceq B\} \iff \{x : f(x) \succeq 0\}, \mathbb{S}_+^n \text{ is convex so } f^{-1}(\mathbb{S}_+^n) \text{ is convex}$$

□

**Question:** We have two disjoint convex sets. Do we always have a hyperplane which strictly separate two sets.

**Answer:** No

**Proof.** For example let  $C = \{x : a^T x \leq b\}$ ,  $D = \{x : a^T x > b\}$ .  $C$  and  $D$  can not be strictly separated by a hyperplane. One of the set is open.  $\square$