

# Convex Optimization

## Lecture 2 - Convex Sets

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# Today's Lecture

- ① Basic Concepts – Affine and Convex Sets
- ② Important Examples
- ③ Operations That Preserve Convexity
- ④ Separating and Supporting Hyperplanes

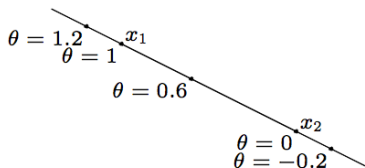
# Outline

- ① Basic Concepts – Affine and Convex Sets
- ② Important Examples
- ③ Operations That Preserve Convexity
- ④ Separating and Supporting Hyperplanes

# Affine Sets

**line** through  $x_1$  and  $x_2$ : all the points  $x$  such that

$$x = \theta x_1 + (1 - \theta)x_2, \theta \in \mathbb{R}$$



**affine set**: contains the line through any 2 distinct points in the set

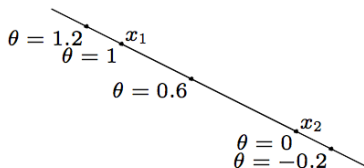
## Questions

- What is the affine set containing 3 points in 2-D space?  
the whole space (or a line if these 3 points are on a line)
- Is the solution set of linear equations  $\{x | Ax = b\}$  affine?  
yes

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# Convex Sets

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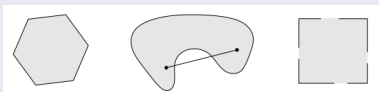
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**convex set**: contains line segment between any 2 distinct points in the set

$$x_1, x_2 \in C, \theta \in [0, 1] \Rightarrow \theta x_1 + (1 - \theta)x_2 \in C$$

## Questions

- Are the following sets convex?



yes; no; no

- Is convex set affine? Is affine set convex?

no; yes

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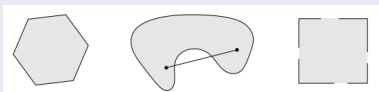
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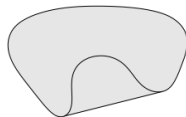
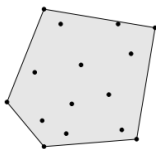
# Convex Combination, Convex Hull

**convex combination** of  $x_1, \dots, x_k$ : all points  $x$  such that

$$x = \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k$$

where  $\theta_i \geq 0, \theta_1 + \dots + \theta_k = 1$

**convex hull  $\text{conv}(S)$** : set of all convex combinations of points in  $S$



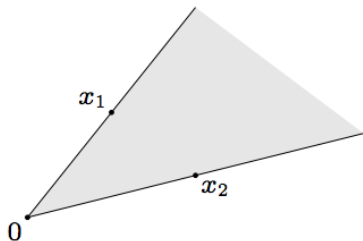


# Convex Cone

**conic combination** of  $x_1$  and  $x_2$ : all points  $x$  such that

$$x = \theta_1 x_1 + \theta_2 x_2$$

where  $\theta_1 \geq 0, \theta_2 \geq 0$



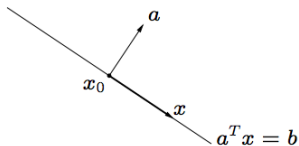
**convex cone**: set that contains all conic combinations of points in the set

# Outline

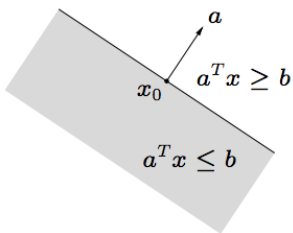
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# Hyperplanes, Halfspaces

**hyperplane:** set of the form  $\{x | a^T x = b\}$ ,  $a \neq 0$



**halfspace:** set of the form  $\{x | a^T x \leq b\}$ ,  $a \neq 0$



# Hyperplanes, Halfspaces

## Questions

- Is a hyperplane affine? Is it convex?  
yes; yes
- Is a halfspace affine? Is it convex?  
no; yes

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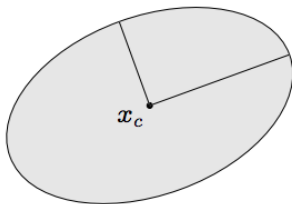
# Euclidean Balls, Ellipsoids

**Euclidean ball** with center  $x_c$  and radius  $r$ :

$$B(x_c, r) = \{x \mid \|x - x_c\|_2 \leq r\} = \{x_c + r \cdot u \mid \|u\|_2 \leq 1\}$$

**Ellipsoid**: set of the form

$$\{x \mid (x - x_c)^T P^{-1} (x - x_c) \leq 1\}, \quad P \text{ is symmetric positive definite}^1$$



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<sup>1</sup>positive definite: for all  $x \in \mathbb{R}^n$ , we have  $x^T P x \geq 0$

# Norm Balls, Norm Cones

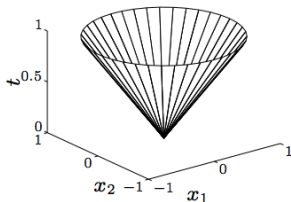
**Norm:** a function  $\|\cdot\|$  that satisfies

- $\|x\| \geq 0$ ;  $\|x\| = 0$  if and only if  $x = 0$
- $\|tx\| = |t|\|x\|$  for  $t \in \mathbb{R}$
- $\|x + y\| \leq \|x\| + \|y\|$

**Norm ball** with center  $x_c$  and radius  $r$ :

$$\{x \mid \|x - x_c\| \leq r\}$$

**Norm cone:** set of the form  $\{(x, t) \mid \|x\| \leq t\}$



# Norm Balls, Norm Cones

## Questions

- What is a norm ball with 1-norm?  
a square center at the origin and rotated by  $45^\circ$



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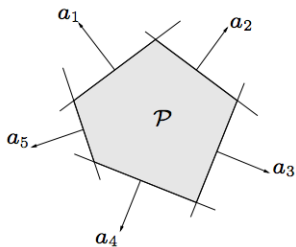
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# Polyhedra

**Polyhedra:** the solution set of finitely many linear inequalities and equalities

$$Ax \leq b, \quad Cx = d$$

where  $A \in \mathbb{R}^{m \times n}$  and  $C \in \mathbb{R}^{p \times n}$



intersection of finite number of halfspaces and hyperplanes

# Positive Semidefinite Cones

- $\mathbb{S}^n$ : set of  $n \times n$  symmetric matrices
- $\mathbb{S}_+^n$ : symmetric positive semidefinite  $n \times n$  matrices

$$X \in \mathbb{S}_+^n \Leftrightarrow z^T X z \geq 0$$

- $\mathbb{S}_{++}^n$ : symmetric positive definite  $n \times n$  matrices

$$X \in \mathbb{S}_{++}^n \Leftrightarrow z^T X z > 0$$

## Questions

- Is  $\mathbb{S}_+^n$  convex? Is  $\mathbb{S}_{++}^n$  convex?

yes; yes

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# Operations That Preserve Convexity

How to decide whether a set  $C$  is convex?

Method 1: By definition

$$x_1, x_2 \in C, \theta \in [0, 1] \Rightarrow \theta x_1 + (1 - \theta)x_2 \in C$$

Method 2: Show that  $C$  is obtained from convex sets (e.g., hyperplanes, norm balls) by operations that preserve convexity

- intersection
- affine functions
- perspective functions
- linear-fractional functions

# Intersection

the intersection of (*finite or infinite* number of) convex sets is convex

## Questions

- How to prove the intersection rule?  
by definition (exercise on board)
- Prove a polyhedron is convex by the intersection rule?  
intersection of halfspaces and hyperplanes
- Prove positive semidefinite cone  $\mathbb{S}_+^n$  is convex by the intersection rule?

$$\mathbb{S}_+^n = \bigcap_{z \neq 0} \{X \in \mathbb{S}^n \mid z^T X z \geq 0\}$$

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# Affine Function

**affine function:**  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  of the form

$$f(x) = Ax + b, \quad A \in \mathbb{R}^{m \times n}, \quad B \in \mathbb{R}^m$$

- the image of a convex set  $S$  under  $f$  is convex

$$S \text{ convex} \Rightarrow f(S) = \{f(x) | x \in S\} \text{ convex}$$

- the inverse image of a convex set  $S$  under  $f$  is convex

$$S \text{ convex} \Rightarrow f^{-1}(S) = \{x | f(x) \in S\} \text{ convex}$$

useful special cases:

- for  $S \subseteq \mathbb{R}^n$ ,  $\alpha \in \mathbb{R}$ , **scaling** of  $S$  is  $\alpha S = \{\alpha x | x \in S\}$
- for  $S \subseteq \mathbb{R}^n$ ,  $\alpha \in \mathbb{R}^n$ , **translation** of  $S$  is  
 $S + \alpha = \{x + \alpha | x \in S\}$
- for  $S \subseteq \mathbb{R}^m \times \mathbb{R}^n$ ,  $\alpha \in \mathbb{R}^n$ , **projection** of  $S$  is  
 $T = \{x_1 \in \mathbb{R}^m | (x_1, x_2) \in S, x_2 \in \mathbb{R}^n\}$

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- Prove a polyhedron is convex by the affine rule?  
 $f(x) = (b - Ax, d - Cx)$ , polyhedron is  $f^{-1}(\mathbb{R}_+^m \times \{0\})$
- Is the solution set to linear matrix inequality (LMI)

$$A(x) = x_1 A_1 + \cdots + x_n A_n \preceq B, \quad A_i, B \in \mathbb{S}^m$$

convex?  $f(x) = B - A(x)$ , the set is  $f^{-1}(\mathbb{S}_+^m)$

- Is hyperbolic cone

$$\left\{ x \mid x^T P x \leq (c^T x)^2, \quad c^T x \geq 0 \right\}, \quad P \in \mathbb{S}_+^n, \quad c \in \mathbb{R}^n$$

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# Perspective

**perspective function:**  $P : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$  of the form

$$P(x, t) = x/t, \text{ with domain } \mathbb{R}^n \times \mathbb{R}_{++}$$

- the image of a convex set under  $P$  is convex
- the inverse image of a convex set under  $P$  is convex

## Questions

- How to prove the above rules?  
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where  $A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$ ,  $c \in \mathbb{R}^n$ ,  $d \in \mathbb{R}$

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composition of affine and perspective functions

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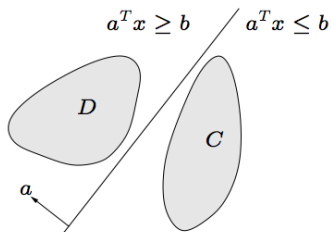
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# Separating Hyperplane Theorem

## Separating Hyperplane Theorem

if  $C$  and  $D$  are nonempty disjoint convex sets, there exist  $a \neq 0, b$  such that

$$a^T x \leq b \text{ for } x \in C, \text{ and } a^T x \geq b \text{ for } x \in D$$

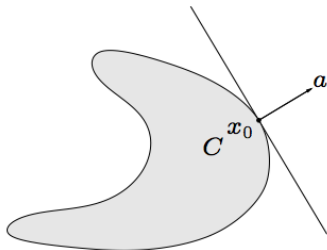


# Supporting Hyperplane Theorem

**supporting hyperplane** to set  $C$  at boundary point  $x_0$ :

$$\{x \mid a^T x = a^T x_0\}$$

where  $a \neq 0$ , and  $a^T x \leq a^T x_0$  for all  $x \in C$



the “tangent” line

# Supporting Hyperplane Theorem

## Supporting Hyperplane Theorem

If  $C$  is a convex set, then there exists a supporting hyperplane at every boundary point of  $C$ .

## Questions

- Prove supporting hyperplane theorem?  
use separating hyperplan theorem on the interior of  $C$  and the boundary point  $x_0$



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