

# Convex sets

Lecture 12

October 26, 2021

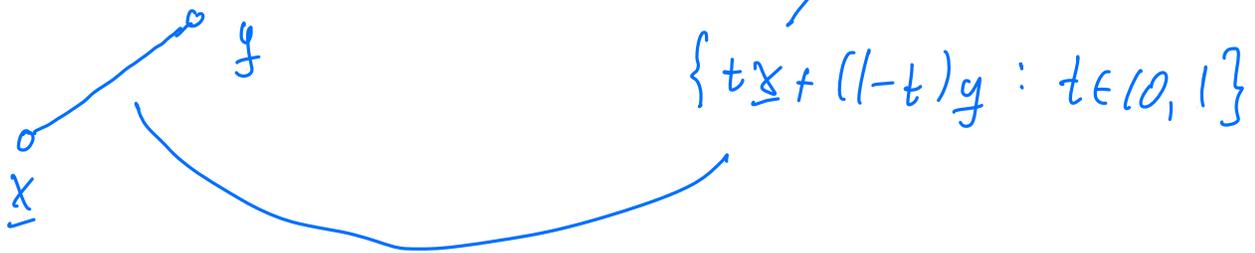
Beck, sections 6.1-6.4

# Convex sets

L12-S01

Definition: A set  $C \subseteq \mathbb{R}^n$  is convex if  $\forall \underline{x}, \underline{y} \in C$  and any  $\lambda \in (0, 1)$ , then  $\lambda \underline{x} + (1-\lambda) \underline{y} \in C$

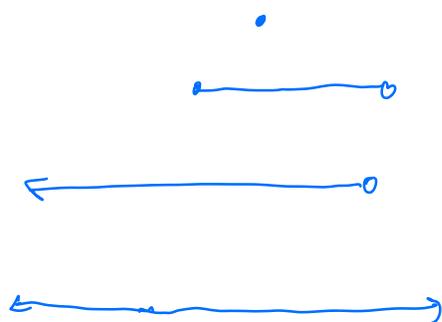
Also:  $C$  is convex if  $\forall \underline{x}, \underline{y} \in C$ , then  $(\underline{x}, \underline{y}) \in C$ .



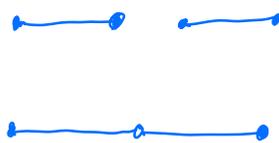
$$[\underline{x}, \underline{y}] = (\underline{x}, \underline{y}) \cup \{\underline{x}, \underline{y}\}$$

Also:  $C$  is convex if,  $\forall x, y \in C$ , the line segment connecting  $x$  and  $y$  lies entirely in  $C$ .

Examples: in  $\mathbb{R}$ : convex sets are intervals (possibly infinite)



convex



not convex

$$C = \{x\} \text{ on } \mathbb{R}$$

$\forall a, b \in C, \forall \lambda \in (0, 1)$  then

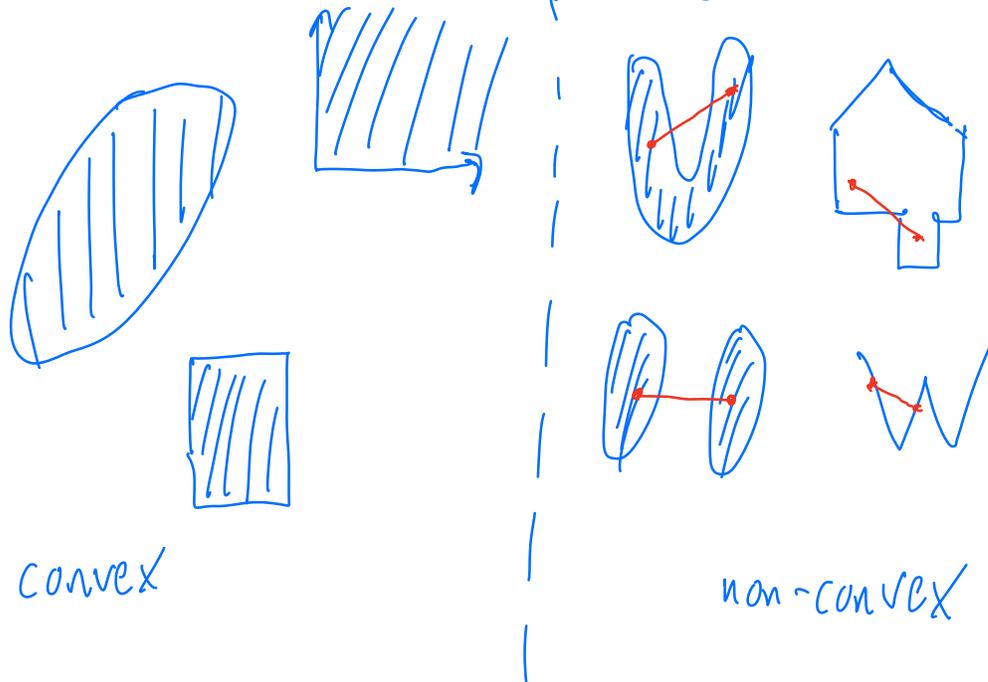
$$\lambda a + (1-\lambda)b \stackrel{?}{\in} C$$

$a=x, b=x$  (no other choices)

$$\lambda x + (1-\lambda)x = x \in C$$

$\Rightarrow C$  is convex.

Examples in  $\mathbb{R}^2$  using "geometry".



Example: Show that a line in  $\mathbb{R}^n$  is convex.

$$L = \{ \underline{x} + t\underline{y} \mid t \in \mathbb{R} \}, \text{ (given } \underline{x}, \underline{y} \text{)}$$

Suppose  $\underline{z}_1, \underline{z}_2 \in L$ . Let  $d \in (0, 1)$ .

want to show:  $d\underline{z}_1 + (1-d)\underline{z}_2 \in L$

Since  $\underline{z}_1, \underline{z}_2 \in L$ , then  $\exists t_1, t_2 \in \mathbb{R}$

s.t.  $\underline{z}_1 = \underline{x} + t_1\underline{y}$ ,  $\underline{z}_2 = \underline{x} + t_2\underline{y}$ .

$$\text{Then: } \lambda \underline{z}_1 + (1-\lambda) \underline{z}_2 = \lambda (\underline{x} + t_1 \underline{y}) + (1-\lambda) (\underline{x} + t_2 \underline{y}) \stackrel{?}{\in} L$$

$$= \underbrace{\underline{x} [\lambda + (1-\lambda)]}_{1} + \underbrace{\underline{y} [t_1 \lambda + (1-\lambda) t_2]}_{t_3}$$

by def. of  $L$

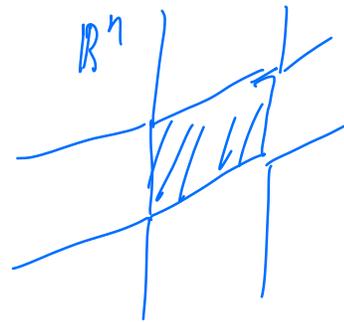
$$= \underline{x} + t_3 \underline{y} \in L$$

$$\Rightarrow \lambda \underline{z}_1 + (1-\lambda) \underline{z}_2 \in L.$$

## Ex. (Hyperplanes)

In  $\mathbb{R}^n$ , given  $\underline{a} \in \mathbb{R}^n$ ,  $b \in \mathbb{R}$ , then a hyperplane  $H(\underline{a}, b) = H$  is given by

$$H = \{ \underline{x} \in \mathbb{R}^n \mid \underline{a}^T \underline{x} = b \}$$



$H$  is convex (for any  $a, b$ )

Proof: Let  $\underline{x}, \underline{y} \in H$  (i.e.  $\underline{a}^T \underline{x} = b$   
 $\underline{a}^T \underline{y} = b$ )

Let  $\lambda \in (0, 1)$ :  $\lambda \underline{x} + (1-\lambda) \underline{y} \stackrel{?}{\in} H$

i.e. is  $\underline{a}^T (\lambda \underline{x} + (1-\lambda) \underline{y}) = b$  ???

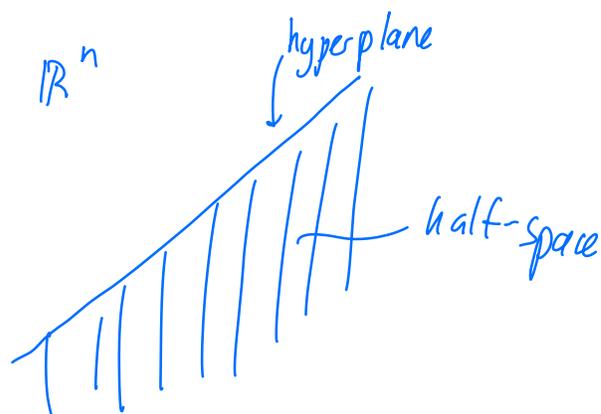
$$\begin{aligned} \underline{a}^T (\lambda \underline{x} + (1-\lambda) \underline{y}) &= \lambda \underline{a}^T \underline{x} + (1-\lambda) \underline{a}^T \underline{y} \\ &= \lambda b + (1-\lambda) b = b \end{aligned}$$

$\Rightarrow \lambda \underline{x} + (1-\lambda) \underline{y} \in H.$

(so  $H$  is convex)

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Ex (Half-spaces)



Given  $\underline{a} \in \mathbb{R}^n$ ,  $b \in \mathbb{R}$ , then  $H^-(\underline{a}, b) = H^-$  is a half-space defined as:

$$H^- = \{ \underline{x} \in \mathbb{R}^n \mid \underline{a}^\top \underline{x} \leq b \}$$

$H^-$  is convex: Let  $\underline{x}, \underline{y} \in H^-$ ,  $\lambda \in (0, 1)$ .

$$\text{I.e. } \underline{a}^\top \underline{x} \leq b$$

$$\underline{a}^\top \underline{y} \leq b$$

$$\text{if } \lambda = -1: \lambda \underline{a}^\top \underline{x} = -\underline{a}^\top \underline{x} \geq -b = \lambda b$$

$$\text{Then: } \underline{a}^\top (\lambda \underline{x} + (1-\lambda) \underline{y}) = \lambda \underline{a}^\top \underline{x} + (1-\lambda) \underline{a}^\top \underline{y}$$

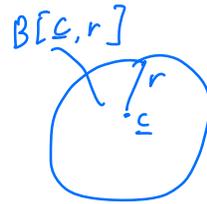
$$\leq \lambda b + (1-\lambda)b = b$$

$\lambda \in (0, 1)$

$\Rightarrow H^-$  is convex

Ex. (Euclidean balls in  $\mathbb{R}^n$ )

$$B[\underline{c}, r] = \{ \underline{x} \in \mathbb{R}^n \mid \|\underline{x} - \underline{c}\|_2 \leq r \}$$



$B[\underline{c}, r]$  is convex: Let  $\underline{x}, \underline{y} \in B[\underline{c}, r]$ ,  $\lambda \in (0, 1)$ .

$$\text{Is } \lambda \underline{x} + (1-\lambda) \underline{y} \in B[\underline{c}, r] ?$$

$$\|[\lambda \underline{x} + (1-\lambda) \underline{y}] - \underline{c}\|_2$$

$$= \| [\lambda x + (1-\lambda)y] - (\lambda c + (1-\lambda)c) \|_2$$

$$= \| \lambda [x - c] + (1-\lambda) [y - c] \|_2$$

$$\leq |\lambda| \|x - c\|_2 + |1-\lambda| \|y - c\|_2$$

triangle inequality

$$= \lambda \|x - c\|_2 + (1-\lambda) \|y - c\|_2 \quad (\lambda \in (0,1))$$

$$\leq \lambda r + (1-\lambda)r = r$$

$$\Rightarrow \lambda x + (1-\lambda)y \in B[c, r]$$

Recall:  $\underline{x} \in \mathbb{R}^n$ ,  $\|\underline{x}\|_p^p = \sum_{j=1}^n |x_j|^p$  ( $l^p$ -norm)

Fact:  $\|\underline{x}\|_p$  is a (proper) norm iff

$B[\underline{0}, r]$  under the distance  $\|\cdot\|_p$  is a convex set.

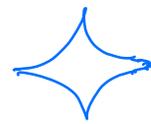
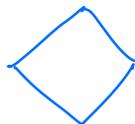
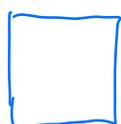
$l^2$

$l^\infty$

$l^1$

$l^\pi$

$l^{1/2}$





# Convex set examples

L12-S02

(Oops)

## Convexity-preserving operations

("Properties of convex sets")

Theorem: Intersections of convex sets are convex.Proof (two sets)Suppose  $C_1, C_2$  are convex. $C = C_1 \cap C_2$  - Is  $C$  convex?Let  $\underline{x}, \underline{y} \in C$ ,  $\lambda \in (0, 1)$

Then  $x, y \in C_1, C_2$

$\Rightarrow \lambda x + (1-\lambda)y \in C_1 \quad (x, y \in C_1, \lambda \in (0,1))$

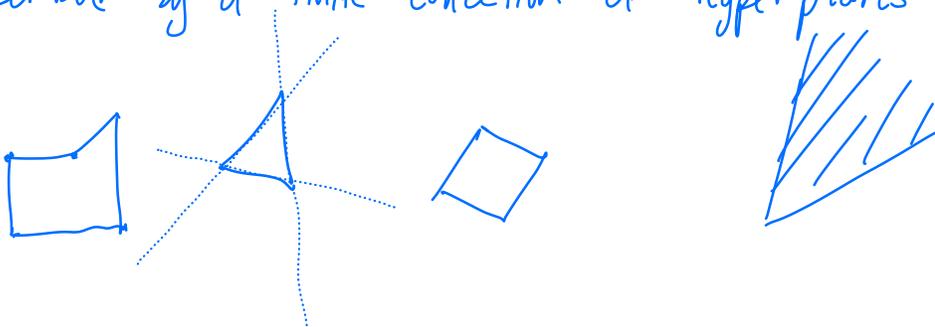
$\Rightarrow \lambda x + (1-\lambda)y \in C_2 \quad (x, y \in C_2, \lambda \in (0,1))$

$\Rightarrow \lambda x + (1-\lambda)y \in C_1 \cap C_2. \quad \square$

Unions of convex sets are not (necessarily) convex.

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A polytope in  $\mathbb{R}^n$  is a set  $S$  whose boundary is described by a finite collection of hyperplanes



Ex: A set  $C$  is a polytope that is convex  
("convex polytope")

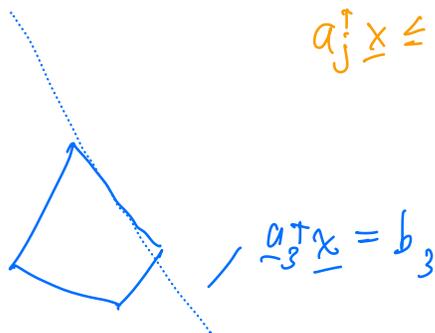
if it's described, given  $\underline{A} \in \mathbb{R}^{m \times n}$ ,  $\underline{b} \in \mathbb{R}^m$ , by

$$C = \{ \underline{x} \in \mathbb{R}^n \mid \underline{A} \underline{x} \leq \underline{b} \},$$

where  $\underline{A}\underline{x} \leq \underline{b}$  means that all componentwise inequalities are true.

$C$  is convex because

$$C = \bigcap_{j=1}^m \underbrace{H(\underline{a}_j, b_j)}_{\substack{\text{half-space} \\ \underline{a}_j^T \underline{x} \leq b_j}}, \text{ where: } \underline{A} = \begin{pmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_m^T \end{pmatrix}$$
$$\underline{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$



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More convexity-preserving operations:

Notation:  $S, T \subset \mathbb{R}^n$

$$S + T = \{ \underline{x} + \underline{y} \in \mathbb{R}^n \mid \underline{x} \in S, \underline{y} \in T \}$$

↑  
"Minkowski"/"set" addition

Proposition: If  $C_1, C_2, \dots, C_k$  are convex sets in  $\mathbb{R}^n$ ,

and  $a_1, a_2, \dots, a_m \in \mathbb{R}$ , then  
 $\sum_{i=1}^m a_i C_i$  is convex

$$(a \cdot C = \{ax \mid x \in C\})$$

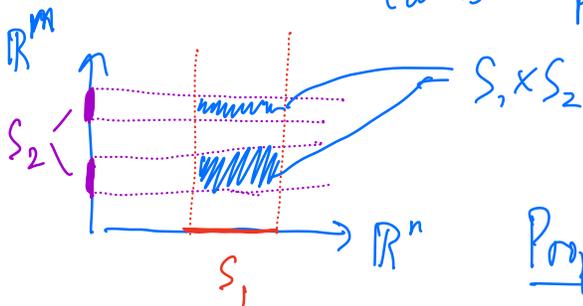
I.e., set addition is convexity-preserving.

Notation: ("Cartesian product")

Given  $S_1 \subset \mathbb{R}^n$ ,  $S_2 \subset \mathbb{R}^m$

Then:  $S_1 \times S_2 = \{(\underline{x}, \underline{y}) \in \mathbb{R}^{m+n} \mid \underline{x} \in S_1, \underline{y} \in S_2\}$

↑  
Cartesian product



Proposition: The Cartesian product

is convexity preserving.

If  $C_1, C_2$  in  $\mathbb{R}^n, \mathbb{R}^m$ ,  
 respectively are convex, then  
 $C_1 \times C_2$  is convex.

Notation:  $S \subset \mathbb{R}^n$ ,  $\underline{A} \in \mathbb{R}^{m \times n}$

$$\underline{A}(S) = \{ \underline{A}\underline{x} \mid \underline{x} \in S \} \subset \mathbb{R}^m$$

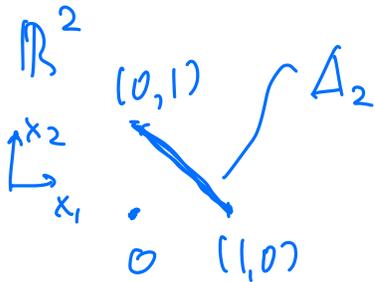
 (arbitrary) linear operation on a set  $S$ .

Proposition: IF  $C \subset \mathbb{R}^n$  is convex, and  $\underline{A} \in \mathbb{R}^{m \times n}$ ,  
then  $\underline{A}(C)$  is convex.

# Convex combinations NS

L12-S04

Recall: in  $\mathbb{R}^n$ ,  $\Delta_n = \left\{ \underline{\lambda} \in \mathbb{R}^n \mid \lambda_i \geq 0, \lambda_i \leq 1 \ \forall i=1..n, \sum_{i=1}^n \lambda_i = 1 \right\}$



Def: Let  $\underline{x}_1, \dots, \underline{x}_k$  be vectors, let  $\underline{\lambda} \in \Delta_k$ . Then  $\sum_{i=1}^k \lambda_i \underline{x}_i$  is a convex combination of  $\{\underline{x}_i\}_{i=1}^k$ .

(Note:  $k=2$  looks familiar...  $\underline{\lambda} \in \Delta_2 \Rightarrow \underline{\lambda} = (\lambda_1, 1-\lambda_1)$ )

Theorem: Let  $C$  be a convex set, and let  $\underline{x}_1, \dots, \underline{x}_k \in C$ .

Then all convex combinations of  $\{\underline{x}_i\}_{i=1}^k$  lie in  $C$ :

$$\forall \underline{\lambda} \in \Delta_k, \text{ then } \sum_{i=1}^k \lambda_i \underline{x}_i \in C.$$

Proof:  $k=2$ : true by definition of convex sets.

Induction: assume it's true for  $r < k$ .

Goal: prove it's true for  $r+1$ .

$$\text{Assume } \underline{\lambda} \in \Delta_{r+1} \Rightarrow \sum_{i=1}^{r+1} \lambda_i \underline{x}_i \stackrel{?}{\in} C$$

$$\sum_{i=1}^{r+1} \lambda_i \underline{x}_i = \lambda_{r+1} \underline{x}_{r+1} + \sum_{i=1}^r \lambda_i \underline{x}_i$$

$$= \lambda_{r+1} \underline{x}_{r+1} + (1 - \lambda_{r+1}) \underbrace{\sum_{i=1}^r \frac{\lambda_i \underline{x}_i}{1 - \lambda_{r+1}}}_{\text{"V"}}$$

$$\text{"V"} = \sum_{i=1}^r \underbrace{\frac{\lambda_i}{1 - \lambda_{r+1}}}_{\mu_i} \underline{x}_i =: \sum_{i=1}^r \mu_i \underline{x}_i$$

because  $\underline{\lambda} \in \Delta_{r+1}$

$$\sum_{i=1}^r \mu_i = \sum_{i=1}^r \frac{\lambda_i}{1 - \lambda_{r+1}} = 1$$

Also:  $\mu_i \geq 0 \Rightarrow \underline{\mu} \in \Delta_r$

$\Rightarrow \underline{v} \in C$  (by inductive hypothesis)

$$\Rightarrow \underline{x} = \lambda_{r+1} \underline{x}_{r+1} + (1 - \lambda_{r+1}) \underline{v}$$

↑                      ↗  
elements of  $C$

$\Rightarrow \underline{x} \in C$  ( $C$  is convex).  $\square$

# The convex hull

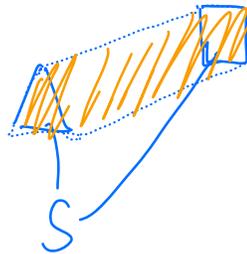
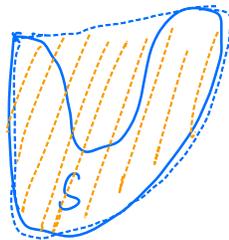
L12-S05

Def: Let  $S \subset \mathbb{R}^n$ . Then the convex hull of  $S$ ,  $\text{conv}(S)$ , is the set of all convex combinations from elements of  $S$ .

$$\text{conv}(S) = \left\{ \sum_{i=1}^k \lambda_i \underline{x}_i \mid \begin{array}{l} \underline{x}_i \in S \quad \forall i=1..k, \\ \underline{\lambda} \in \Delta_k, \\ k \in \mathbb{N} \end{array} \right\}$$

Geometrically: convex hulls are sets constructed by placing a rubber band around  $S$

Ex:  : convex hull



Fact:  $\text{conv}(S)$  is convex.

Proposition:  $\text{conv}(S)$  is the smallest convex set containing  $S$ .

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Note: the definition (algebraic) of  $\text{conv}(S)$  is unwieldy:  $\sum_{i=1}^k \lambda_i x_i$  for  $k$  arbitrary.

Turns out:  $k=n+1$  is enough ( $x_i \in \mathbb{R}^n$ )

# Carathéodory's Theorem

L12-S06

Theorem: Let  $S \subset \mathbb{R}^n$ . Let  $\underline{x} \in \text{conv}(S)$  be arbitrary.

Then:  $\exists \underline{x}_1, \dots, \underline{x}_{n+1}, \underline{d} \in \Delta_{n+1}$  ( $\underline{x}_i \in S$ ) s.t.

$$\underline{x} = \sum_{i=1}^{n+1} d_i \underline{x}_i$$

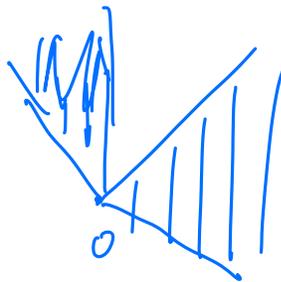
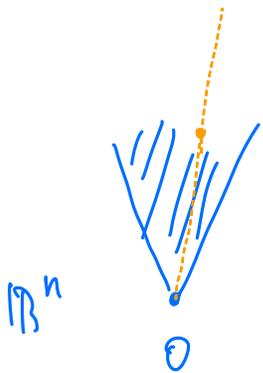
Note: Such a representation for  $\underline{x}$  is called a barycentric representation.

# Convex cones

L12-S07

Def: A set  $S$  is a cone if  $\forall \underline{x} \in S, \lambda \geq 0$ , then

$$\lambda \underline{x} \in S$$



(cones need not be convex)

Recall:  $\mathbb{R}_+^k = \{ \underline{x} \in \mathbb{R}^k \mid x_i \geq 0 \quad \forall i=1..k \}$

Definition: (Conic combination) Let  $\underline{x}_1, \dots, \underline{x}_k$  be vectors, and let  $\underline{\lambda} \in \mathbb{R}_+^k$ . Then  $\sum_{i=1}^k \lambda_i \underline{x}_i$  is a conic combination.

Proposition: A set  $S$  is a convex cone iff all conic combinations lie in  $S$ .

Ex: Recall that  $\{ \underline{x} \in \mathbb{R}^n \mid \underline{A} \underline{x} \leq \underline{b} \}$  is a convex polytope.  
( $\underline{A}, \underline{b}$  arbitrary)

If  $\underline{b} = \underline{0}$ , then this set is a convex cone.

Ex: Consider polynomials of degree  $k$  on  $\mathbb{R}$ .  
The set of all non-negative polynomials is a convex cone.

# Examples

L12-S08

# The conic representation theorem

L12-S09