

Convex sets

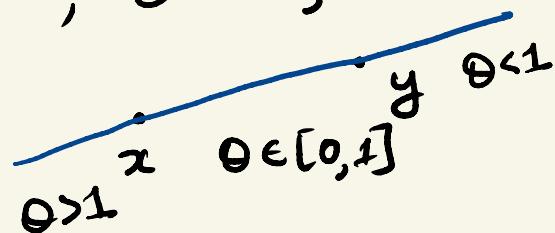
- Affine sets
- Convex sets
- Examples.

Affine sets.

A line

through the distinct points $x, y \in \mathbb{R}^n$

$$S = \{ z \mid z = \theta x + (1-\theta)y , \theta \in \mathbb{R} \}$$



Affine set contains all line

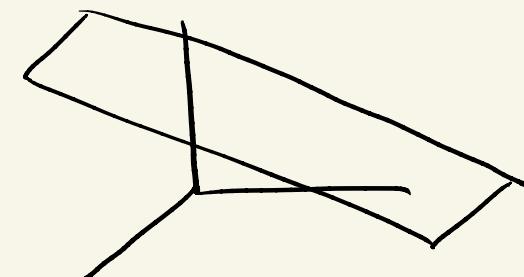
through any distinct point in the set.

$$x, y \in S \Leftrightarrow \theta x + (1-\theta)y = z \in S, \forall \theta \in \mathbb{R}$$

Example: The solution set of linear

equations: $S = \{ x \mid Ax = b \}$

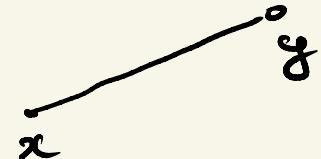
$x, y \in S, \theta x + (1-\theta)y = z$ satisfies $Az = b$.



Convex combination.

line segment between any two points $x, y \in \mathbb{R}^n$ is:

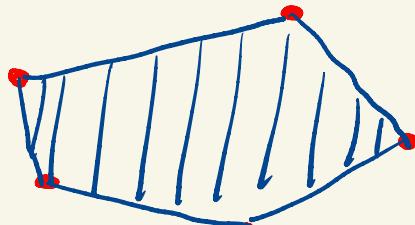
$$z = \theta x + (1-\theta)y, 0 \leq \theta \leq 1$$



x is a convex combination of x_1, \dots, x_n if

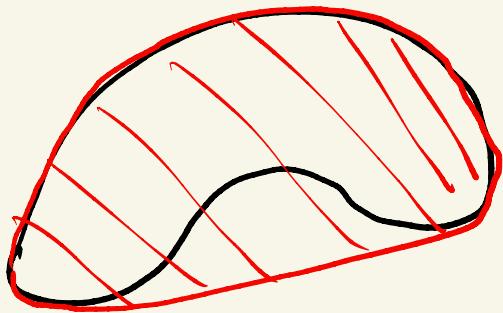
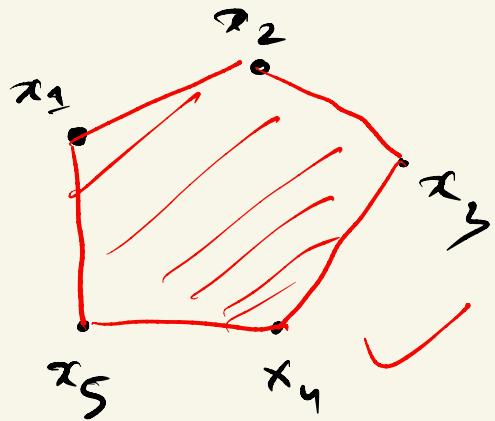
$$x = \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n = \sum_{i=1}^n \theta_i x_i \text{ and}$$

$$\sum_{i=1}^n \theta_i = 1 \text{ with } \theta_i \geq 0$$



The convex hull of S contains all convex combinations of points S :

$$\text{co } S = \left\{ z = \sum_{i=1}^n \theta_i x_i \mid x_i \in S, \sum_{i=1}^n \theta_i = 1, \theta_i \geq 0 \right\}$$

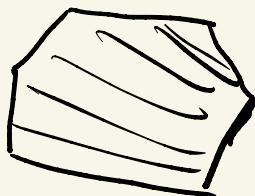


Convex set.

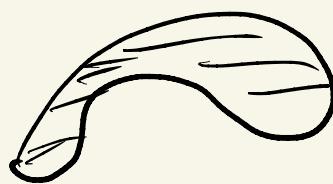
$S \subset \mathbb{R}^n$ is a convex if it contains all convex combinations of points in S .

$$x, y \in S \Leftrightarrow 0 \leq \theta \leq 1, \quad \theta x + (1-\theta)y \in S.$$

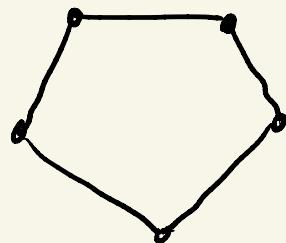
Example:



convex



not convex



not convex

Linear set

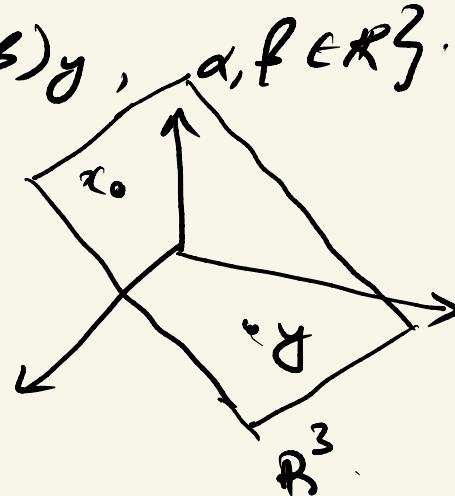
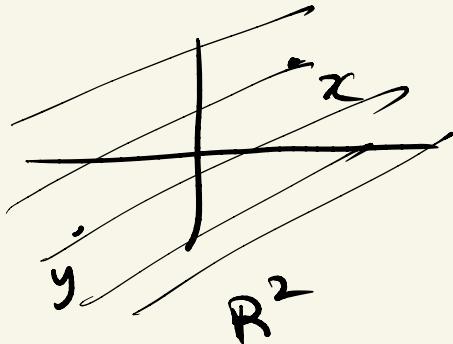
S is a subspace if

- $0 \in S$
- closed under addition
- closed under scalar multiplication.

S is the subspace containing two distinct points

$$x, y \in \mathbb{R}^n$$

$$S = \{ z \mid z = \alpha x + (1-\alpha)y, \alpha, \beta \in \mathbb{R} \}$$

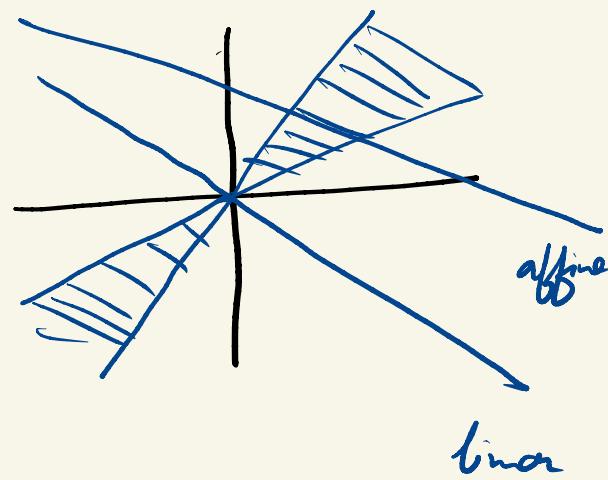


A linear set contains all lines through distinct points in the set.

$$x \in S, y \in S \Leftrightarrow dx + fy = z \in S, \forall d, f \in \mathbb{R}.$$

Linear sets are affine but affine sets are not always linear

Example: The range and null space of a matrix is a linear set.

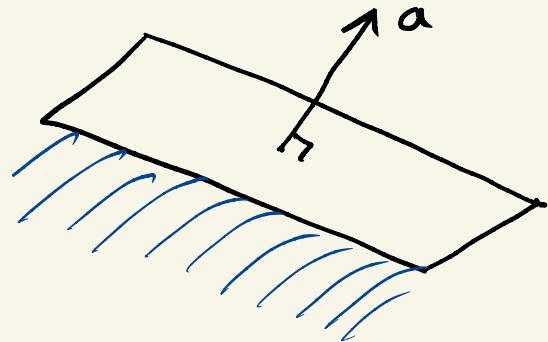


Half space and hyper planes.

hyper plane: set of the form $\{x \mid a^T x = b\}$.

half space: set of the form

$$\{x \mid a^T x \leq b\}.$$



- $a \neq 0$ is a normal vector.

- hyperplane are affine and convex.

$$\begin{aligned} \text{pick } x, y \in S, a^T [\theta x + (1-\theta)y] \\ &= \theta b + (1-\theta)b \\ &= b \quad \text{if } \theta \in [0, 1]. \\ \Rightarrow \theta x + (1-\theta)y &\in S. \end{aligned}$$

- half space are convex but are not affine.
- Example: The non-negative orthant

$$R_+^n = \{x : x_i \geq 0 \text{ for } i=1, \dots, n\}.$$

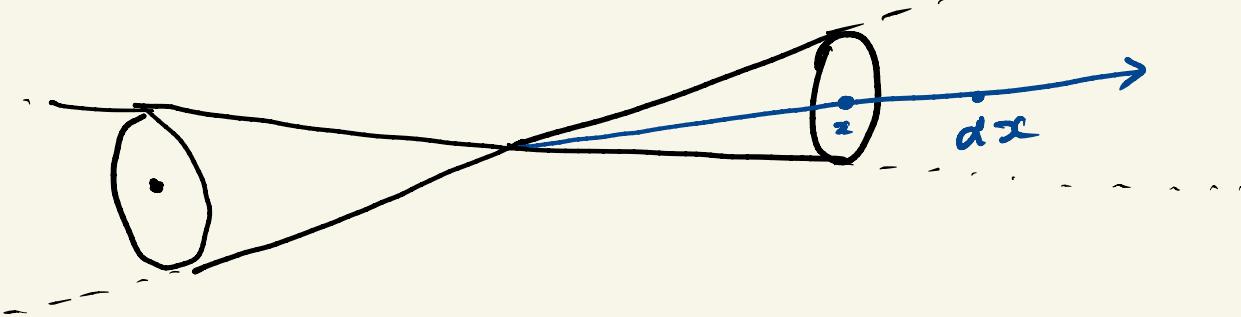
is intersection of halfspace

$$H_i = \{x \in R^n / e_i^T x \geq 0\}.$$

- The non-negative orthant is a cone

Cone

A set $S \subseteq \mathbb{R}^n$ is a cone if $x \in S \Leftrightarrow \alpha x \in S \quad \forall \alpha \geq 0$



A convex cone is a cone that is convex:

$$x, y \in S \Leftrightarrow \underbrace{\theta_1 x + \theta_2 y}_{\text{conic combination}} \in S, \quad \forall \theta_1, \theta_2 \geq 0.$$

Example: \mathbb{R}_+^n - non-negative orthant

$$L_+^n = \left\{ \begin{bmatrix} x \\ t \end{bmatrix} \mid \|x\| \leq t, x \in \mathbb{R}^n, t \in \mathbb{R}_+ \right\}$$

