

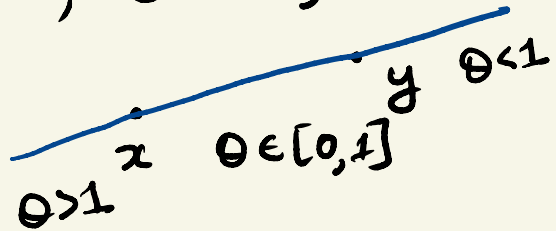
Convex sets

- Affine sets
- Convex sets
- Examples.

Affine sets.

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

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

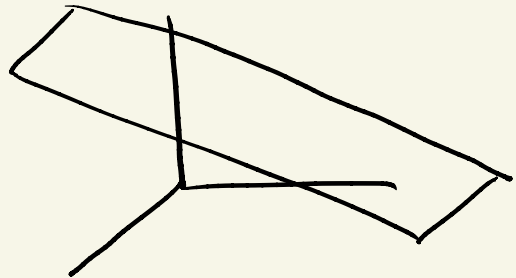


Affine set contains all line through any distinct point in the set.

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

Example: The solution set of linear equations: $S = \{x \mid Ax = b\}$.

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



Convex combination.

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

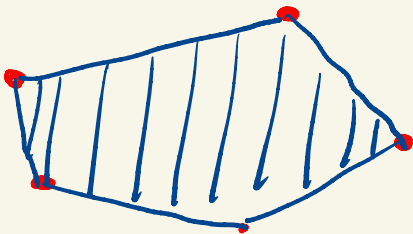
$$z = \theta x + (1-\theta)y, 0 \leq \theta \leq 1$$



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

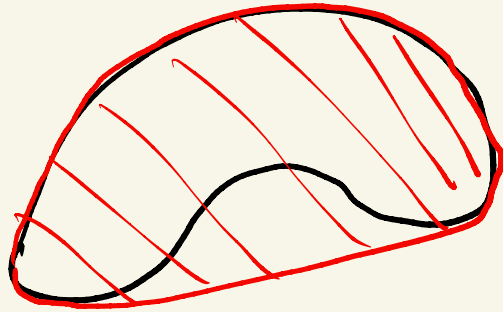
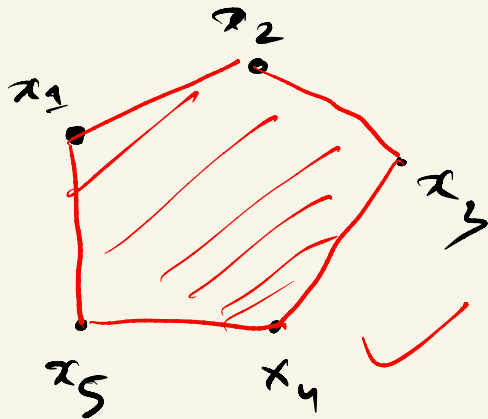
$$x = \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n = \sum_{i=1}^n \theta_i x_i \quad \text{and}$$

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



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

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

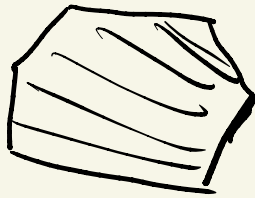


Convex set:

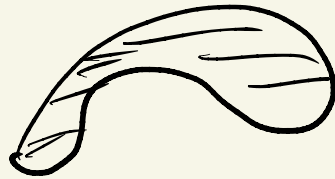
$S \subseteq \mathbb{R}^n$ is a convex if it contains all convex combinations of points in S .

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

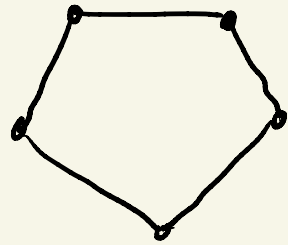
Example:



convex



not-convex



not-convex.

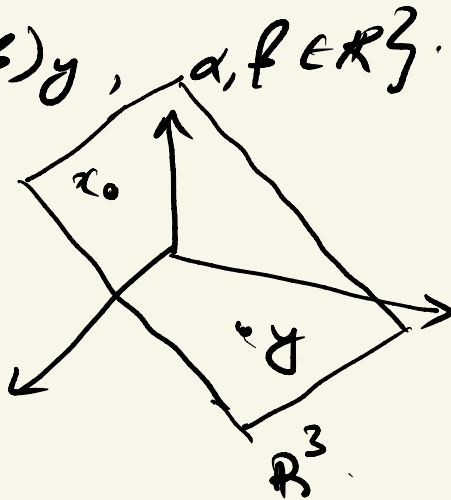
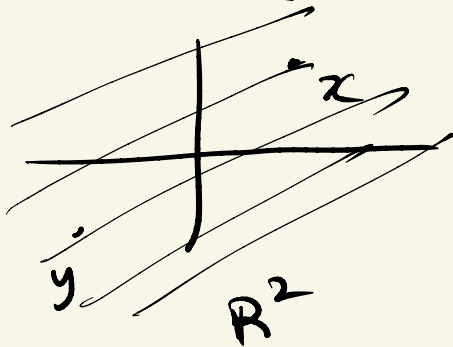
Linear set.

S is a subspace if

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

S is the subspace containing two distinct points $x, y \in \mathbb{R}^n$

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

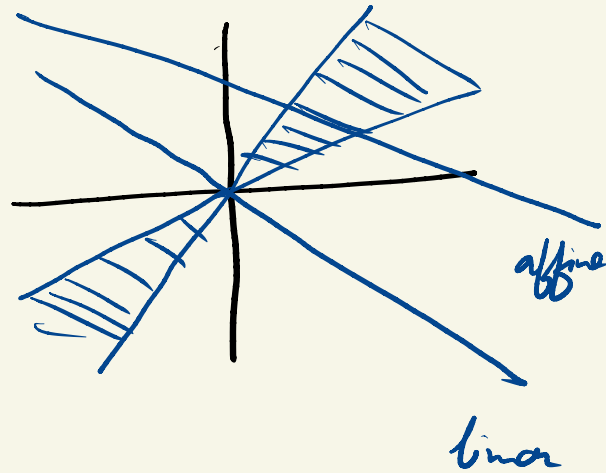


A linear set contains all lines through distinct points in the set:

$$x \in S, y \in S \Leftrightarrow \alpha x + \beta y = z \in S, \quad \forall \alpha, \beta \in \mathbb{R}$$

Linear sets are affine but affine sets are not always linear

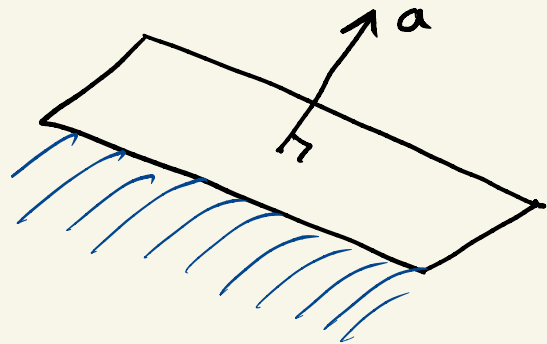
Example: The range and null space of a matrix is a linear set.



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.

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

- half space are convex but are not affine.

- Example: The non-negative orthant

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

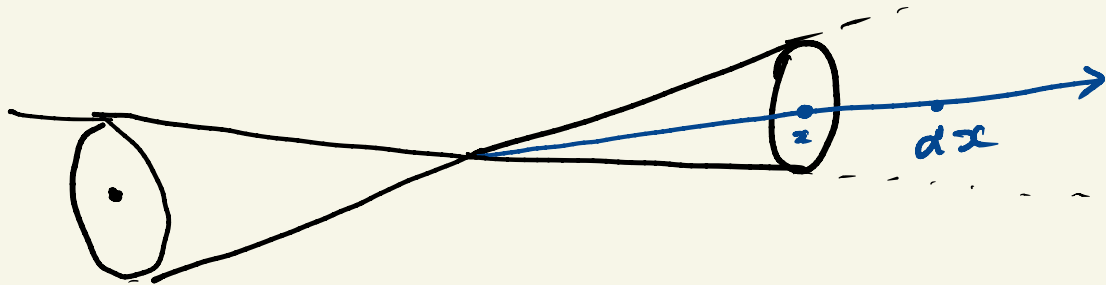
is intersection of half space

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

- The non-negative orthant is a cone

Cone

A set $S \subset \mathbb{R}^n$ is a cone if $x \in S \iff dx \in S \quad \forall d \geq 0$



A convex cone is a cone that is convex:

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

Example:

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

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

