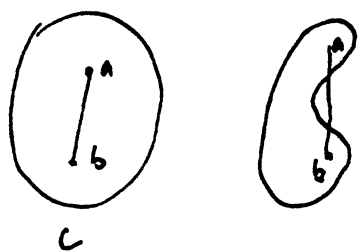


# Topics in Convex Optimisation

Def A set  $C \subseteq \mathbb{R}^n$  is convex if  $\forall a, b \in C \ \forall \lambda \in [0, 1] \ \lambda a + (1-\lambda)b \in C$



• Halfspace:  $\{x \in \mathbb{R}^n : \langle a, x \rangle \leq b\}$  where  $a \in \mathbb{R}^n \setminus \{0\}$ ,  $b \in \mathbb{R}$

• Unit balls:  $\{x \in \mathbb{R}^n : \|x\| \leq 1\}$  where  $\|\cdot\|$  norm on  $\mathbb{R}^n$

Prop (operations that preserve convexity)

• Let  $C \subseteq \mathbb{R}^n$  convex and  $A: \mathbb{R}^n \rightarrow \mathbb{R}^n$  linear, then  $A(C)$  is convex.

• Let  $C_1, C_2 \subseteq \mathbb{R}^n$  convex. then  $C_1 \cap C_2$  is convex.

$$C = \{a \in \mathbb{R}^{n+1} : a_0 + a_1 x + \dots + a_n x^n \geq 0 \ \forall x \in \mathbb{R}\}$$

$$= \bigcap_{x \in \mathbb{R}} \underbrace{\{a \in \mathbb{R}^{n+1} : a_0 + a_1 x + \dots + a_n x^n \geq 0\}}_{H_x \text{ halfspace (definition)}}$$

Def (Dimension of a convex set)

Let  $C \subseteq \mathbb{R}^n$  be convex. the dimension of  $C$  is the dim of the smallest affine space containing  $C$ .  
 $C$  is called full dimensional in  $\mathbb{R}^n$  if  $\dim C = n$ .

Def (Convex hull)

Let  $S \subseteq \mathbb{R}^n$ . The convex hull of  $S$ , denoted  $\text{conv}(S)$  is the smallest convex set that contains  $S$ .

$$\text{conv}(S) = \bigcap_{\substack{C \text{ convex} \\ S \subseteq C}} C = \{x \in \mathbb{R}^n : \exists k \in \mathbb{N}, s_1, \dots, s_k \in S \ x = \sum \lambda_i s_i \ \lambda_1, \dots, \lambda_k \in [0, 1] \ \sum \lambda_i = 1\}$$

Exercise (Carathéodory's Theorem)

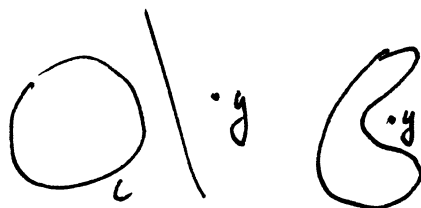
Let  $S \subseteq \mathbb{R}^n$ . Show that any element of  $\text{conv}(S)$  can be expressed as a convex combination of at most  $n+1$  points on  $S$ .

Theorem (Separating hyperplane theorem)

Let  $C \subseteq \mathbb{R}^n$ ,  $y \notin C$ .  $C$  convex.

then there exist  $a \in \mathbb{R}^n \setminus \{0\}$   $b \in \mathbb{R}$

$$\text{s.t. } \begin{cases} \langle a, x \rangle \leq b \ \forall x \in C \\ \langle a, y \rangle \geq b \end{cases}$$



Proof: We'll assume  $C$  is closed. Let  $p_C(y) = \arg \min_x \{ \|y - x\|, x \in C \}$

$$\langle y - p_C(y), x - p_C(y) \rangle \leq 0 \ \forall x \in C$$



$$a = y - p_C(y) \neq 0 \quad \text{since } y \notin C$$

$$b = \langle a, p_C(y) \rangle$$

This choice of  $a, b$  satisfies:

$$\langle a, x \rangle \leq b \quad \langle a, y \rangle \geq b \quad \forall x \in C$$

$$(\langle a, x \rangle - b = \langle y - p_C(y), x \rangle - \langle y - p_C(y), p_C(y) \rangle = \langle y - p_C(y), x - p_C(y) \rangle \leq 0)$$

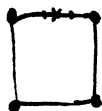
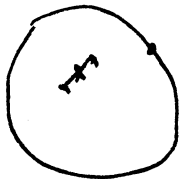
Def (Faces and extreme points)

Let  $C \subseteq \mathbb{R}^n$  be convex. A set  $F \subseteq C$  is called a face if:

(i)  $F$  is convex

(ii)  $\forall x \in F \quad \forall a, b \in C, \lambda \in (0, 1)$  s.t.  $x = \lambda a + (1-\lambda)b \Rightarrow a, b \in F$

If  $F$  is a singleton  $F = \{x_0\}$   $x_0$  is called an extreme point



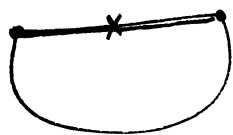
Theorem (Minkowski)

Let  $C$  be closed and bounded convex set. Let  $\text{ext}(C)$  be the extreme points of  $C$ . Then  $C = \text{conv}(\text{ext}(C))$ .

Lemma Let  $C \subseteq \mathbb{R}^n$  be full-dimensional closed convex set

(i) Let  $F \subseteq G \subseteq C$  s.t.  $G$  is a face of  $C$  and  $F$  a face of  $G$ , then  $F$  is a face of  $C$

(ii) Let  $x_0 \in C \setminus \underbrace{\text{int}(C)}_{\text{interior of } C}$ . Then there exists face  $F$  of  $C$  with  $\dim F < \dim C$  s.t.  $x_0 \in F$



Proof of (ii):  $x_0 \in \text{int}(C)$  Separating hyperplane theorem:  $\exists a \neq 0, b \in \mathbb{R}$   
s.t.  $\langle a, x_0 \rangle = b \quad \langle a, x \rangle \leq b \quad \forall x \in \text{int}(C)$

$F = C \cap \{x : \langle a, x \rangle = b\}$  a face that contains  $x_0$ ,  $\dim \leq n-1 < n = \dim(C)$   $\square$

Proof of theorem

We want to show  $C = \text{conv}(\text{ext}(C))$ . The inclusion  $C \supseteq \text{conv}(\text{ext}(C))$  is trivial.

We need to show  $C \subseteq \text{conv}(\text{ext}(C))$ .

