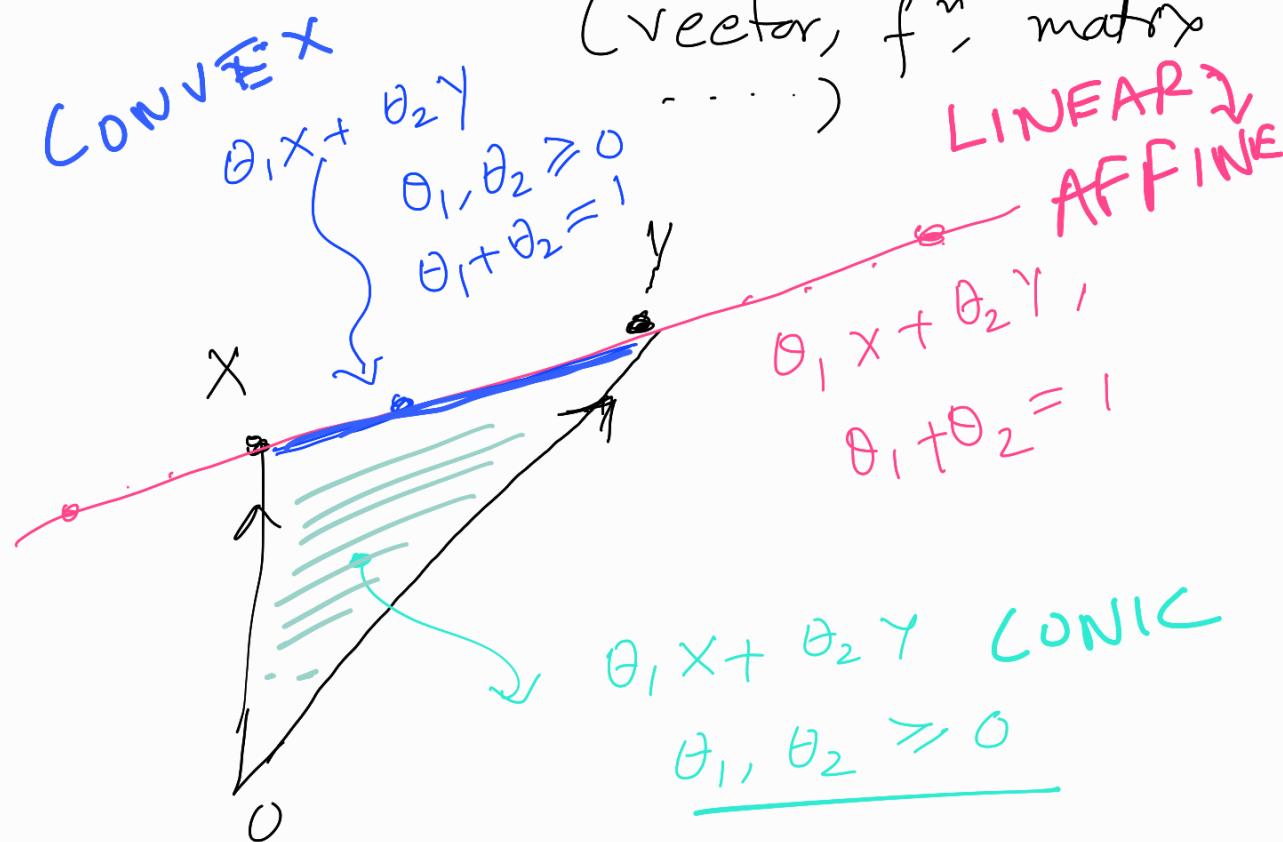


Combination of mathematical objects

(vector, function, matrix)



Affine hull :

The affine hull of a set C is the set of all affine combinations of points in C ,

$$\text{aff } C = \left\{ \sum \theta_i x_i \mid \sum \theta_i = 1 \right\}$$

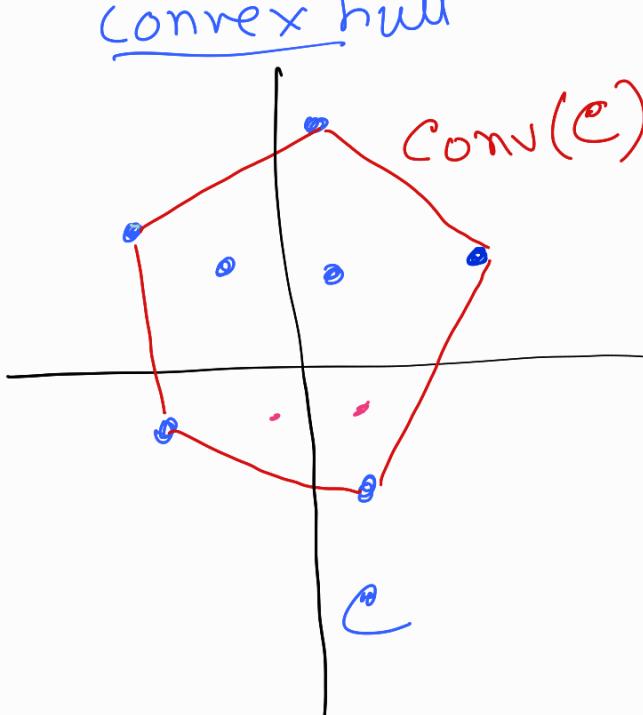
Convex hull of $C \subseteq \mathbb{R}^n$ is defined as

$$\text{Conv}(C) = \left\{ \sum \theta_i x_i \mid \underline{\sum \theta_i = 1, \theta_i \geq 0} \right\}$$

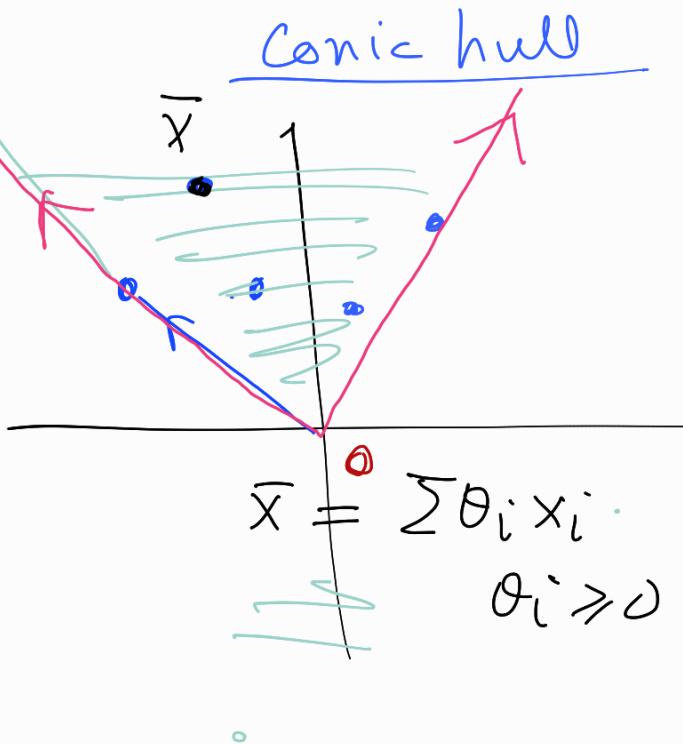
Conic hull : of a set C is defined as

$$\left\{ \sum \theta_i x_i \mid \theta_i \geq 0 \right\}$$

convex hull

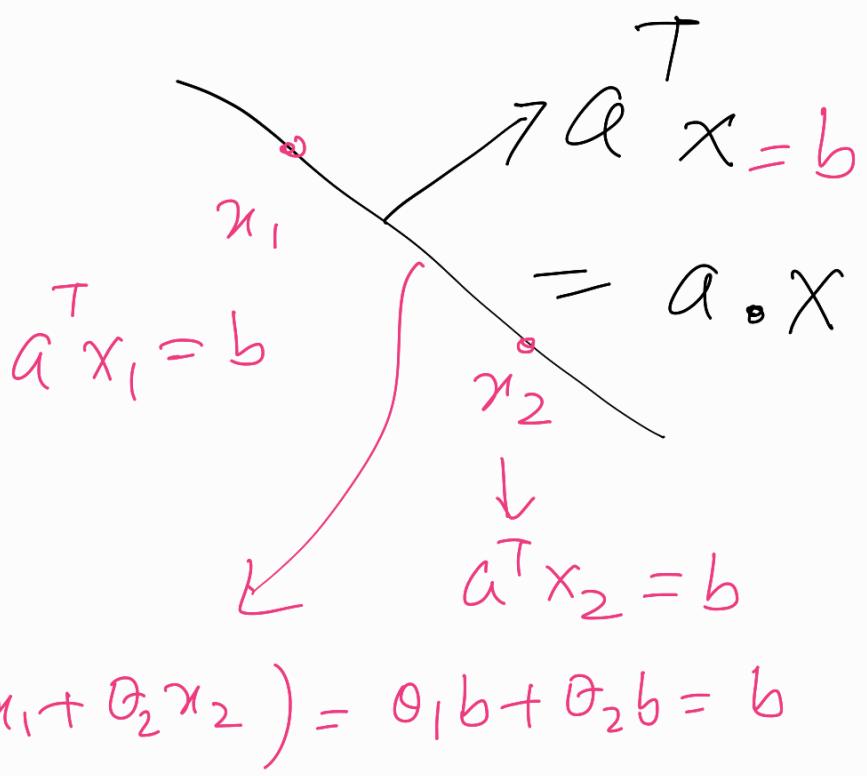


Convex set



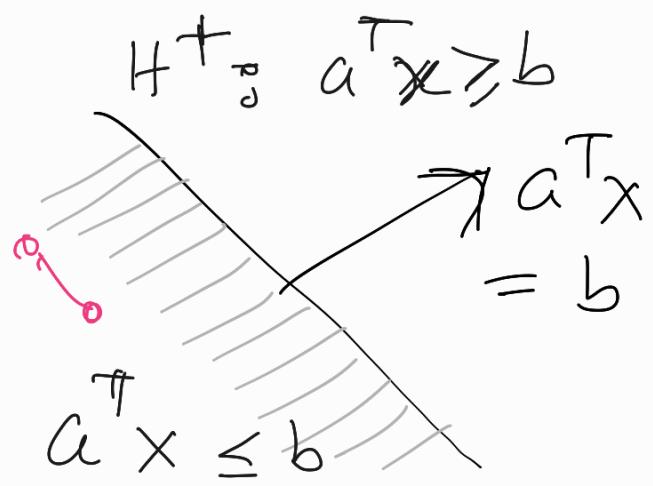
Examples of convex set

1) Hyperplane

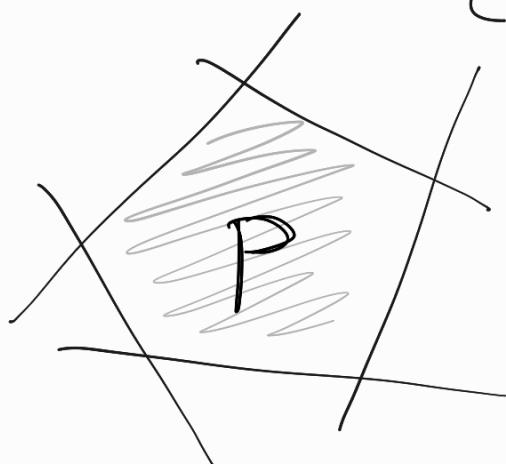


$$a^T (\theta_1 x_1 + \theta_2 x_2) = \theta_1 b + \theta_2 b = b$$

2) Half space



3) Polyhedron $P = \{x \mid a_i^T x \leq b_i, i=1 \dots m\}$

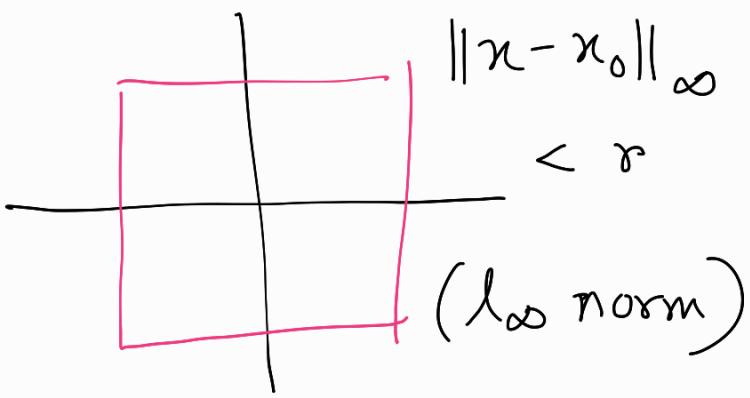
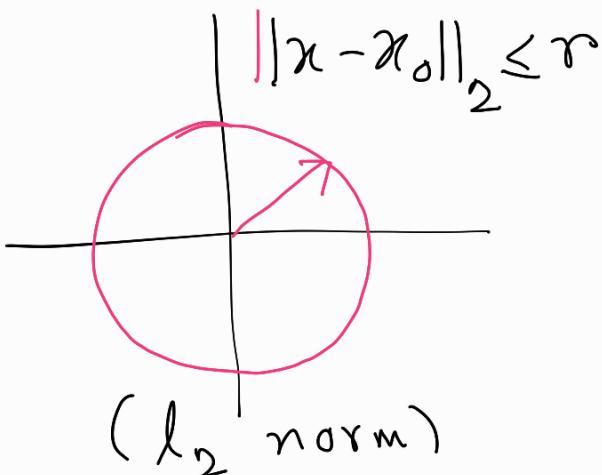


↓
Intersection
of half spaces

Note: Intersection of convex sets
is also convex

4) Norm ball : $\{x \mid \|x - x_0\|_p \leq r\}$

$$r > 0, p \geq 1$$

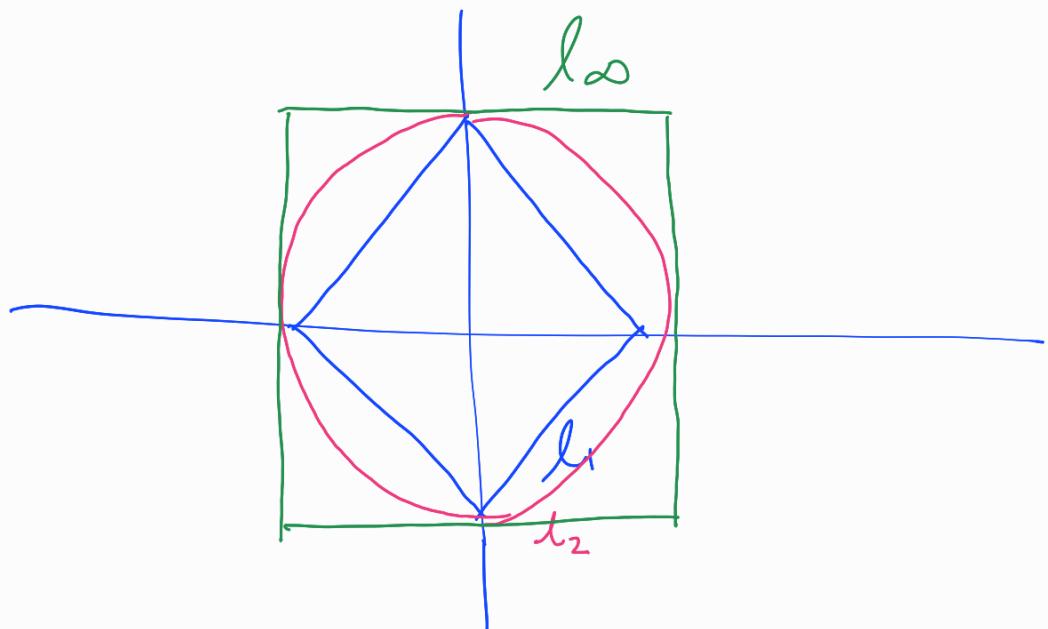


$$l_1 \text{ norm} : \|x - y\|_1 = \sum |x_i - y_i|$$

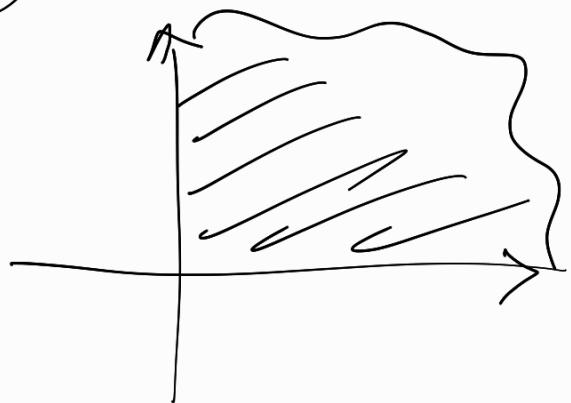
$$l_2 \text{ norm} : \|x - y\|_2 = \sqrt{\sum (x_i - y_i)^2}$$

$$l_p \text{ norm} : \|x - y\|_p = \sqrt[p]{\sum (x_i - y_i)^p}$$

$$l_\infty \text{ norm} : \|x - y\|_\infty = \max \left\{ |x_i - y_i|, i=1\dots \right\}$$



5) Positive orthant \mathbb{R}^n_+



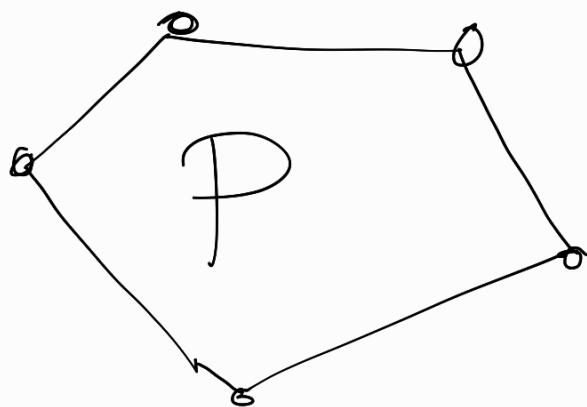
6) \mathbb{R}^n is convex

of $x_1 \dots x_n$

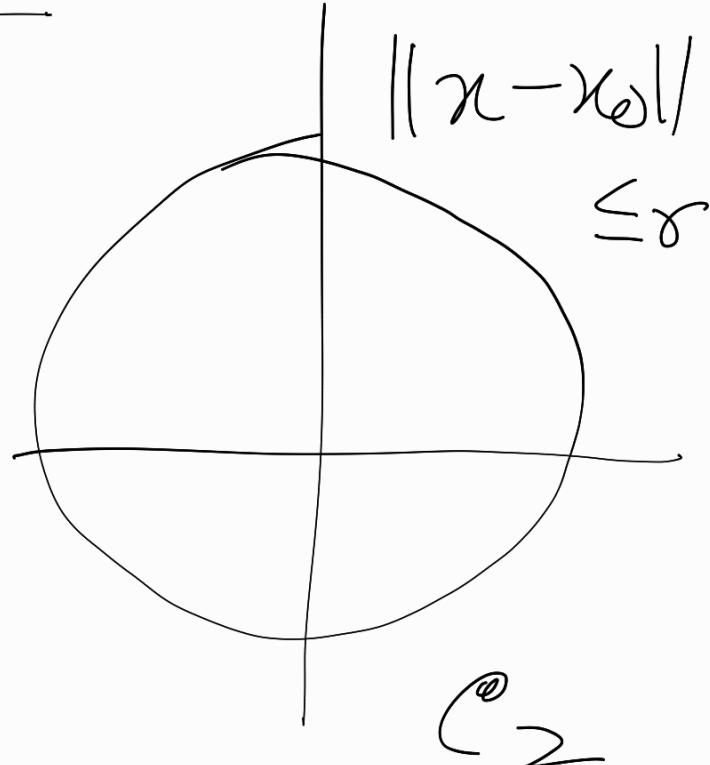
Try to show convex hull is
convex and smallest set
that contains $\{x_1 \dots x_n\}$

→ Separated?

Convex sets → Supported

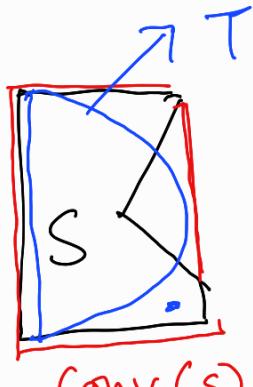


C_1



C_2

Theorem : $\text{Conv}(S)$ is the smallest convex set that contains S .



By definition it is convex.

To show it is smallest assume T is convex,
 $(T \supseteq S)$ contains S but is smaller
 i.e., $\text{Conv}(S) \supseteq T$

Let $\bar{x} \notin T$ but $\bar{x} \in \text{Conv}(S)$

$$\bar{x} \in \text{Conv}(S) \Rightarrow \bar{x} = \sum \theta_i x_i, \sum \theta_i = 1, \theta_i \geq 0 \\ x_i \in S \quad \forall i$$

$$\Rightarrow x_i \in T, \quad \forall i$$

$$\Rightarrow \bar{x} \in T \quad (\because T \text{ is convex})$$

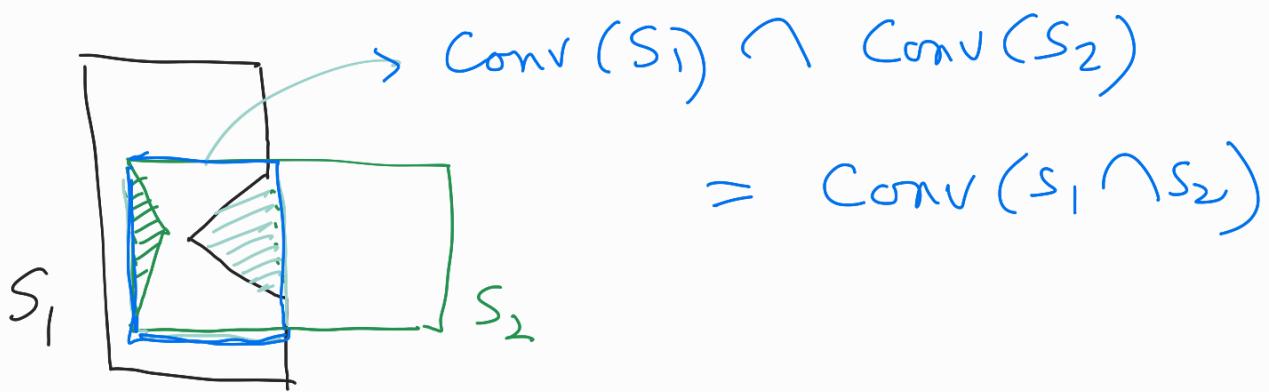
CONTRADICTION

Example : Show $\text{Conv}(S_1 \cap S_2) \subseteq$

$$\text{Conv}(S_1) \cap \text{Conv}(S_2)$$

Show $x \in \text{Conv}(S_1 \cap S_2) \Rightarrow$

$$x \in \text{Conv}(S_1) \cap \text{Conv}(S_2)$$



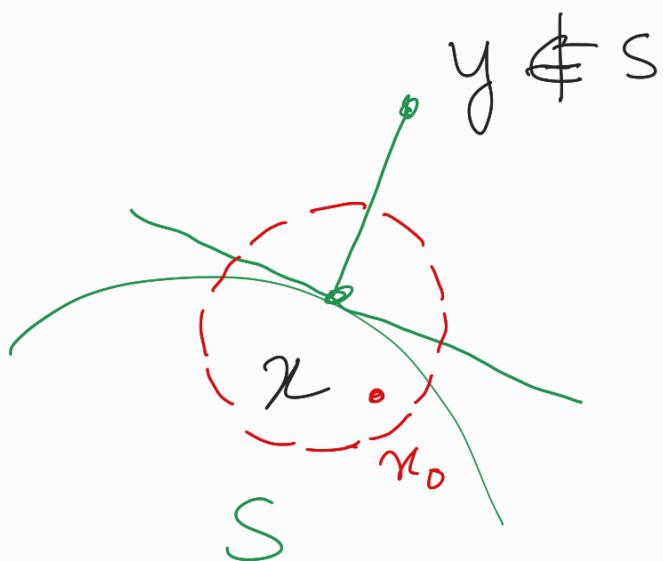
$$\text{Let } x \in \text{Conv}(S_1 \cap S_2) \Rightarrow x = \sum x_i \theta_i \\ x_i \in S_1 \cap S_2$$

$$\Rightarrow x_i \in S_1 \text{ & } x_i \in S_2$$

$$\Rightarrow x \in \text{Conv}(S_1) \text{ & } x \in \text{Conv}(S_2) \\ \therefore \text{Conv}(S_1) \ni S_1 \text{ etc.}$$

$$\Rightarrow x \in \underbrace{\text{Conv}(S_1) \cap \text{Conv}(S_2)}$$

PROJECTIONS



$$x = \Pi_S y$$

y is projection of
y on S convex

What is x?

Ans : $\min_{x \in S} \|y - x\|$

$$\equiv \min \|y - x\|$$

s.t. $x \in S \cap \{x \in S \mid \|y - x\| \leq \|y - x_0\|\}$

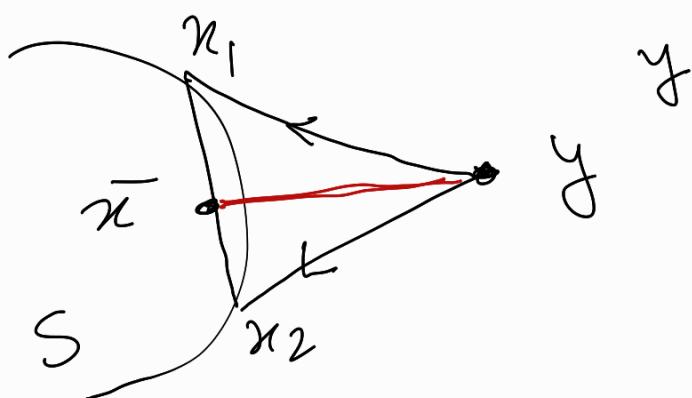
distance of x_0 and
y is more, $x_0 \notin S$

bounded, closed and convex
 \equiv compact

because it
is norm ball

According to Weierstrass theorem a continuous function has an extremum on a compact set.

Is x unique? Suppose x_1 and x_2 are equidistant from y .



$$x_1, x_2 \in S \Rightarrow \lambda x_1 + (1-\lambda) x_2 \in S$$

Take $\lambda = \frac{1}{2}$ so, $\bar{x} = \frac{1}{2} x_1 + \frac{1}{2} x_2 \in S$

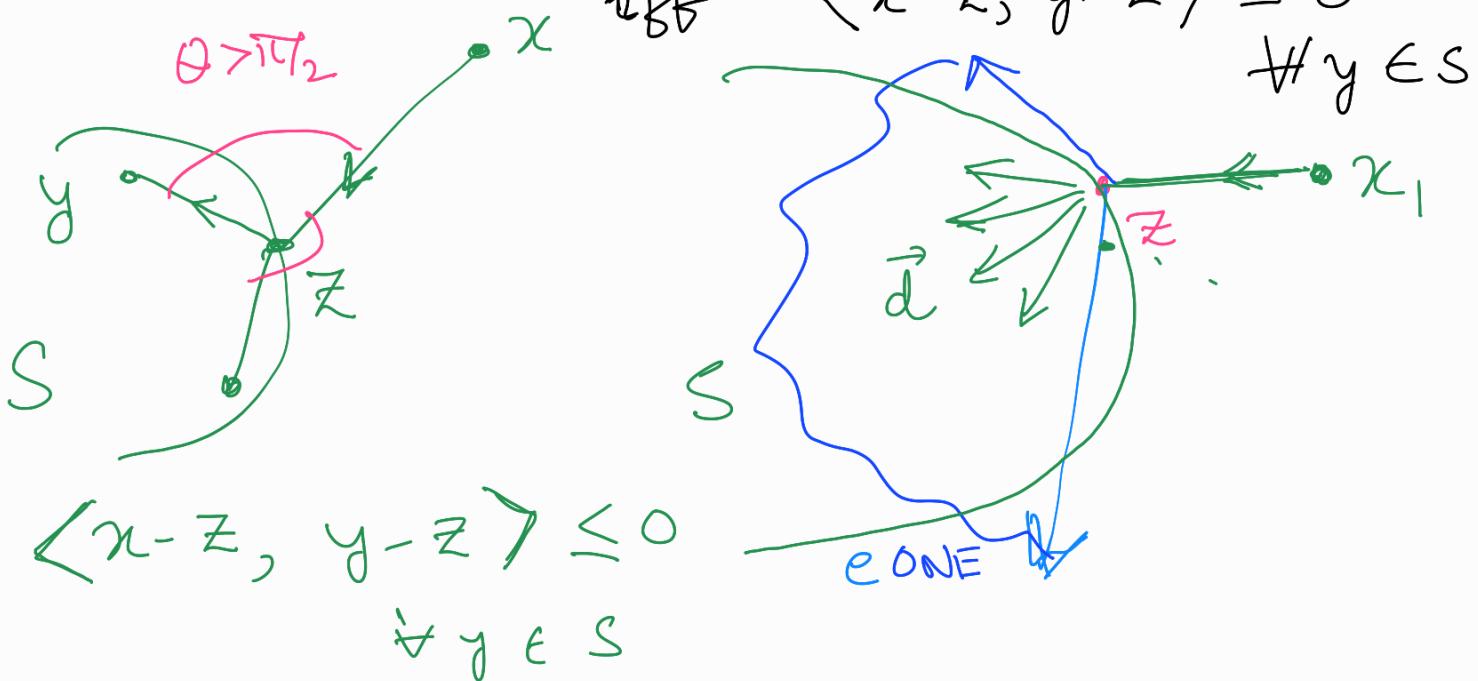
Check that $\|y - \bar{x}\|$

$$\begin{aligned} &= \|y - \frac{1}{2}(x_1 + x_2)\| \\ &\leq \frac{1}{2} \|y - x_1\| + \frac{1}{2} \|y - x_2\| \\ &= \|y - x_1\| \end{aligned}$$

This leads to contradiction. Therefore there are no two points x_1 and x_2 which are both projections. So, projection point is always unique.

Theorem : S convex, non empty, closed
 Then $z = \Pi_S(x)$, for some $x \notin S$

iff $\langle x-z, y-z \rangle \leq 0$



" \Rightarrow "

Suppose $z = \Pi_S(x)$ is the projection
 of x on S (convex)

For any $y \in S$, let

$$\begin{aligned} g(t) &= \|x-z - t(y-z)\|^2 \\ &= \|x-z\|^2 - 2t \langle x-z, y-z \rangle + t^2 \|y-z\|^2 \end{aligned}$$

$$\text{Put } t=0, \quad g(0) = \|x-z\|^2 < g(t) \quad \forall t \in (0,1]$$

$$\text{So, } 2 \langle x-z, y-z \rangle < t \|y-z\|^2 \quad t \in (0,1] \quad \langle x-z, y-z \rangle \leq 0$$

Therefore $\langle x-z, y-z \rangle \leq 0$

Now, suppose,

$$\langle x-z, y-z \rangle \leq 0 \quad \forall y \in S$$

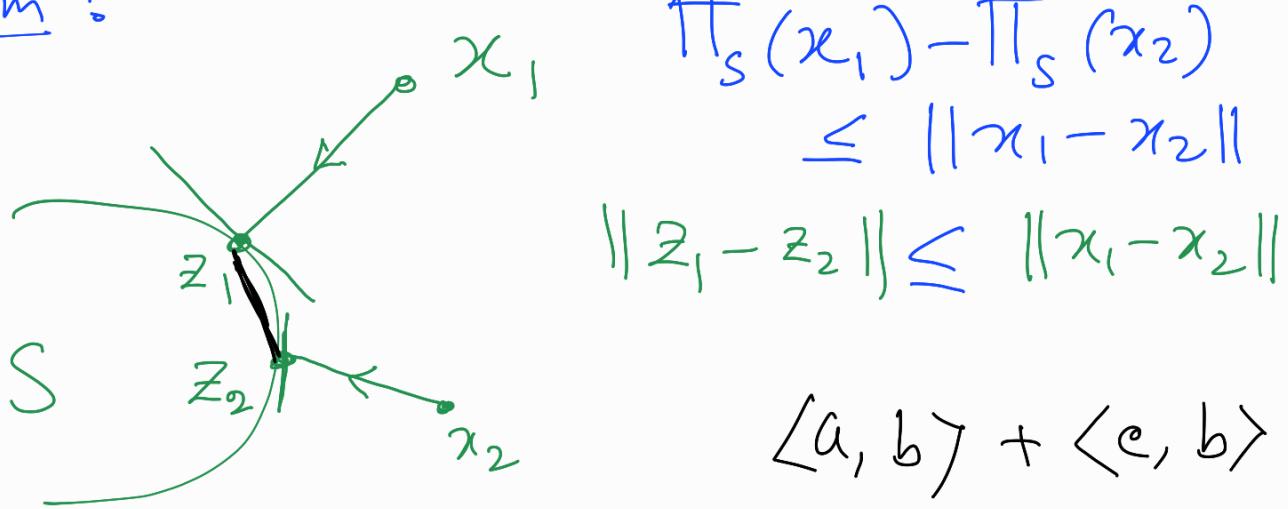
Since $g(0) < g(t) \quad \forall t \in (0, 1]$

we have $g(0) < g(1)$, for $y \neq z$

$$\Rightarrow \|x-z\| < \|x-y\|, \quad \forall y \in S \\ \text{and } y \neq z$$

So, z must be projection point.

Theorem :



Proof : $\langle x_1 - z_1, z_1 - z_2 \rangle \leq 0$

and $\langle x_2 - z_2, z_1 - z_2 \rangle \leq 0$

Add both

$$\|z_1 - z_2\|^2 + \langle z_1 - z_2, x_1 - x_2 \rangle \leq 0$$

$$\Rightarrow \|z_1 - z_2\|^2 \leq \langle z_1 - z_2, x_2 - x_1 \rangle$$

$$= \|z_1 - z_2\| \cdot \|x_2 - x_1\|$$

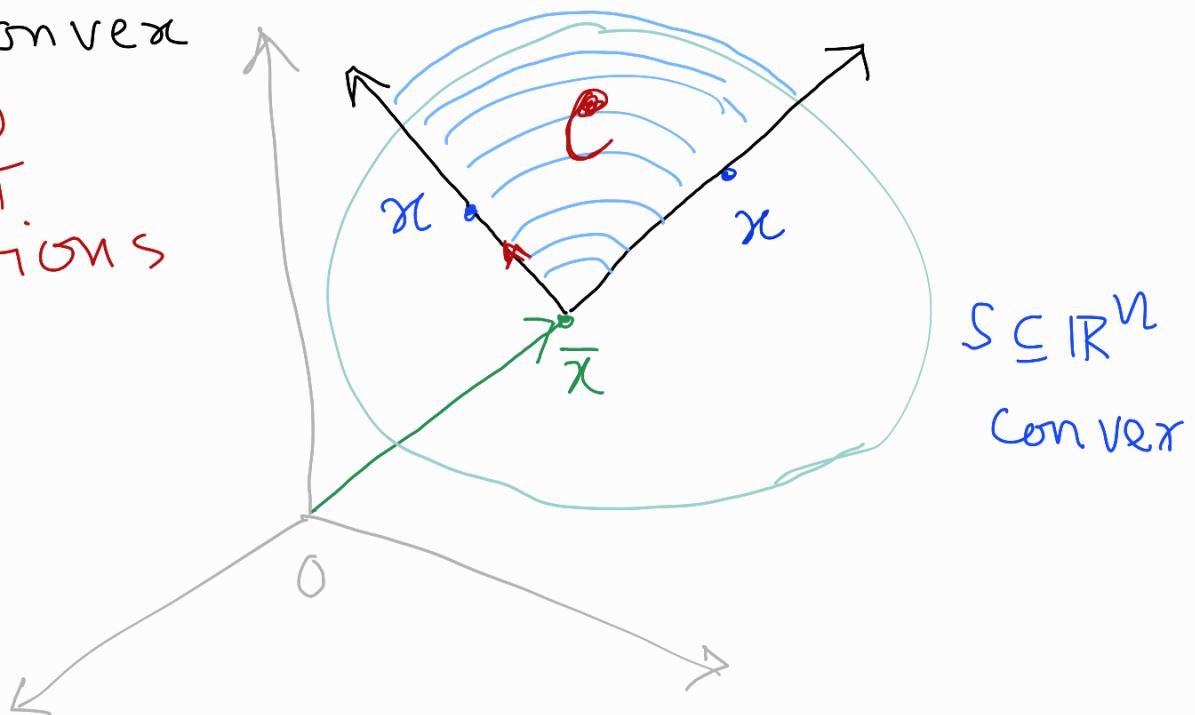
$$\Rightarrow \|z_1 - z_2\| < \|x_2 - x_1\| \quad \underline{\text{QED.}}$$

Homework (convex) Image Set

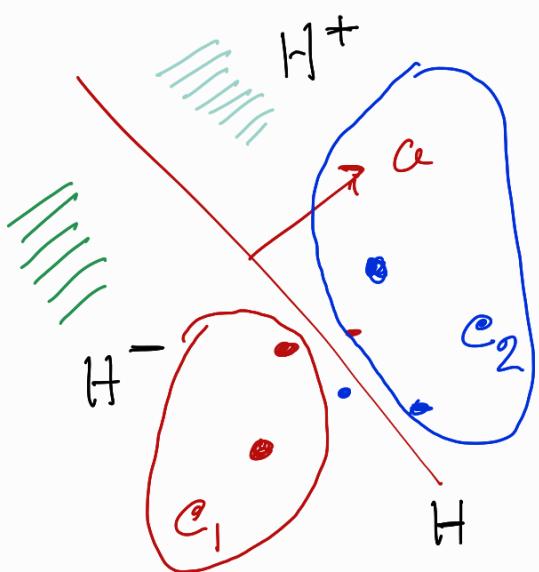
1) Let $S \subseteq \mathbb{R}^n$, $A_{m \times n}$ and $a \in \mathbb{R}$
 show $AS = \{y \mid y = AX, X \in S\} \subseteq \mathbb{R}^m$ is convex

2) Let $S \subseteq \mathbb{R}^n$, $S \neq \emptyset$ and $\bar{x} \in S$,
 show $C = \{y \mid y = \lambda(x - \bar{x}), \lambda \geq 0, x \in S\}$
 is convex

Cone of
directions



Separation of Hyperplanes



Hyperplane is defined as

$$H = \{x \in \mathbb{R}^n \mid \langle x, a \rangle = \alpha\}$$

$$a \neq 0$$

$$a \in \mathbb{R}^n$$

$$\alpha \in \mathbb{R}$$

$$H^- = \{x \in \mathbb{R}^n \mid \langle x, a \rangle \leq \alpha\}$$

*Closed
Half spaces*

$$H^+ = \{x \in \mathbb{R}^n \mid \langle x, a \rangle \geq \alpha\}$$

$$H^{--} = \{x \in \mathbb{R}^n \mid \langle x, a \rangle < \alpha\}$$

*Open
Half space*

$$H^{++} = \{x \in \mathbb{R}^n \mid \langle x, a \rangle > \alpha\}$$

Let C_1 and C_2 be two non-empty sets.

- ① H is a separating hyperplane of C_1 & C_2
if $C_1 \subseteq H^+$ and $C_2 \subseteq H^-$
- ② . . . strictly separating . . .
if $C_1 \subseteq H^{++}$ and $C_2 \subseteq H^{--}$

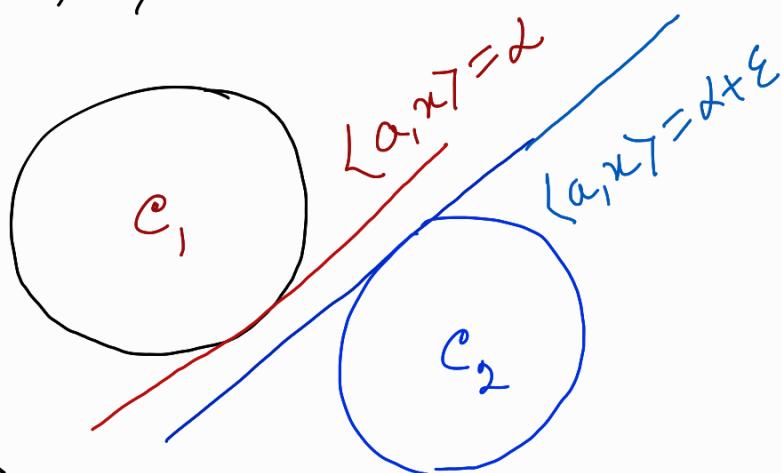
(3) H is called properly separating hyperplane if H separates C_1 and C_2 and those are not both contained in H



(4) H is called strongly separating hyperplane of C_1 and C_2 if.

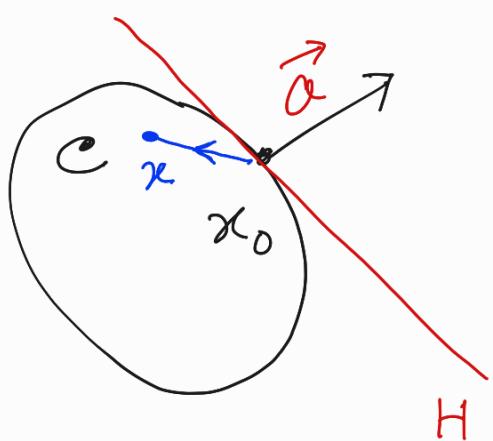
$$\langle a, x \rangle = \alpha \text{ for } x \in C_1 \text{ and}$$

$$\langle a, x \rangle = \alpha + \epsilon \text{ for } x \in C_2, \epsilon > 0$$



Theorem

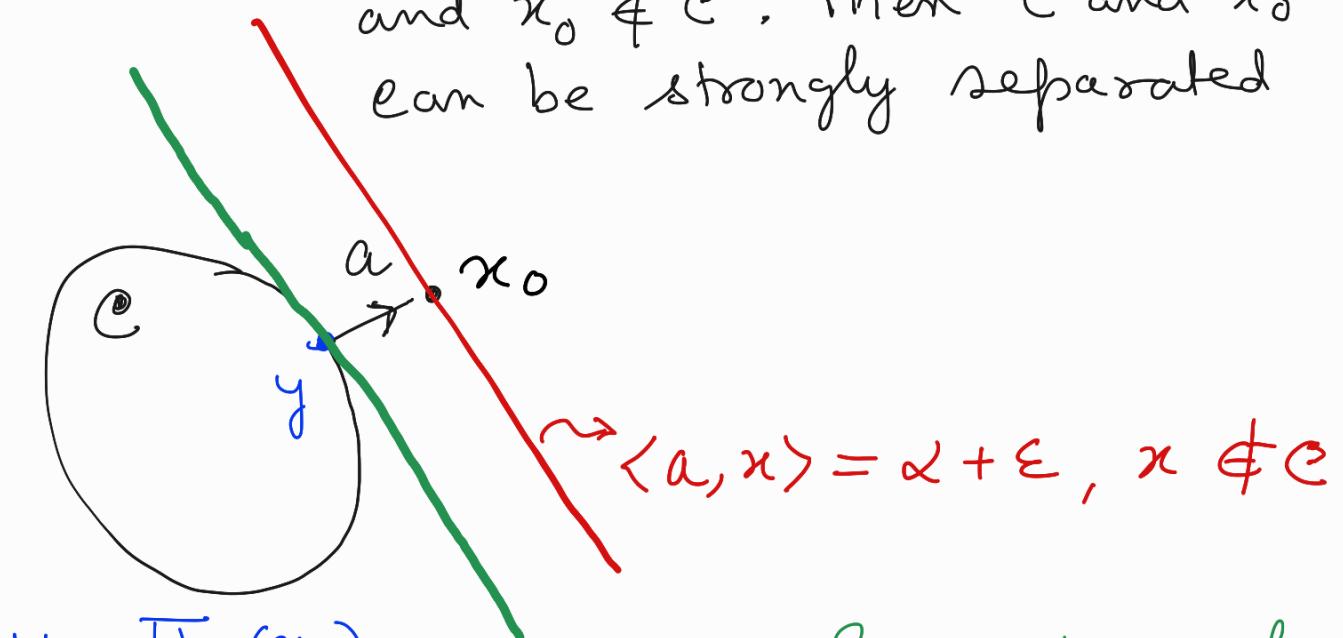
Supporting hyperplane : If $C \subseteq \mathbb{R}^n$ is a closed convex set and $x_0 \in \partial C$ (boundary point), then there exists a hyperplane to C at x_0 s.t. $\langle a, x \rangle \leq \langle a, x_0 \rangle, \forall x \in C$



Separation Theorems

(Separation between a point and set)

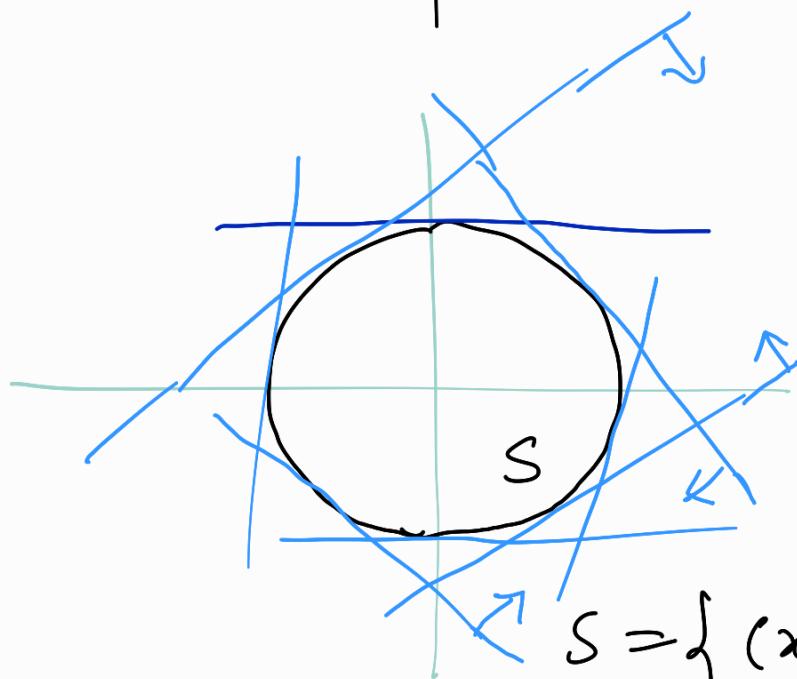
Theorem : Let $C \subseteq \mathbb{R}^n$ be closed convex and $x_0 \notin C$. Then C and x_0 can be strongly separated



Supporting hyperplane of
 C at point $y = \pi_C(x_0)$
 $\langle a, x \rangle = \alpha, x \in C$

Proof :

Note: A closed convex set is the intersection of closed halfspaces



$$S = \{(x, y) |$$

$$x^2 + y^2 \leq r^2\}$$

$$S = \bigcap H_i, H_i = \{x \mid \langle a, x \rangle \leq c\}$$

$x \in S$

Separating hyperplane theorem

Let C_1 and C_2 be non-empty convex sets, if $C_1 \cap C_2 = \emptyset$ then there exists a hyperplane that separates them.

$$\text{Proof } C = C_1 - C_2$$

$$= \left\{ x = x_1 - x_2 \mid x_1 \in C_1, x_2 \in C_2 \right\}$$

As, $C_1 \cap C_2 = \emptyset$, so, $0 \notin C$

So, by separation of set and point not in set, we can say C and 0 can be separated, ie, $\exists a \neq 0$,

$$\langle a, x \rangle \geq 0 \quad \forall x \in C$$

$$\Rightarrow \langle a, x_1 \rangle \geq \langle a, x_2 \rangle, \text{ for } x_1 \in C_1 \\ x_2 \in C_2$$

