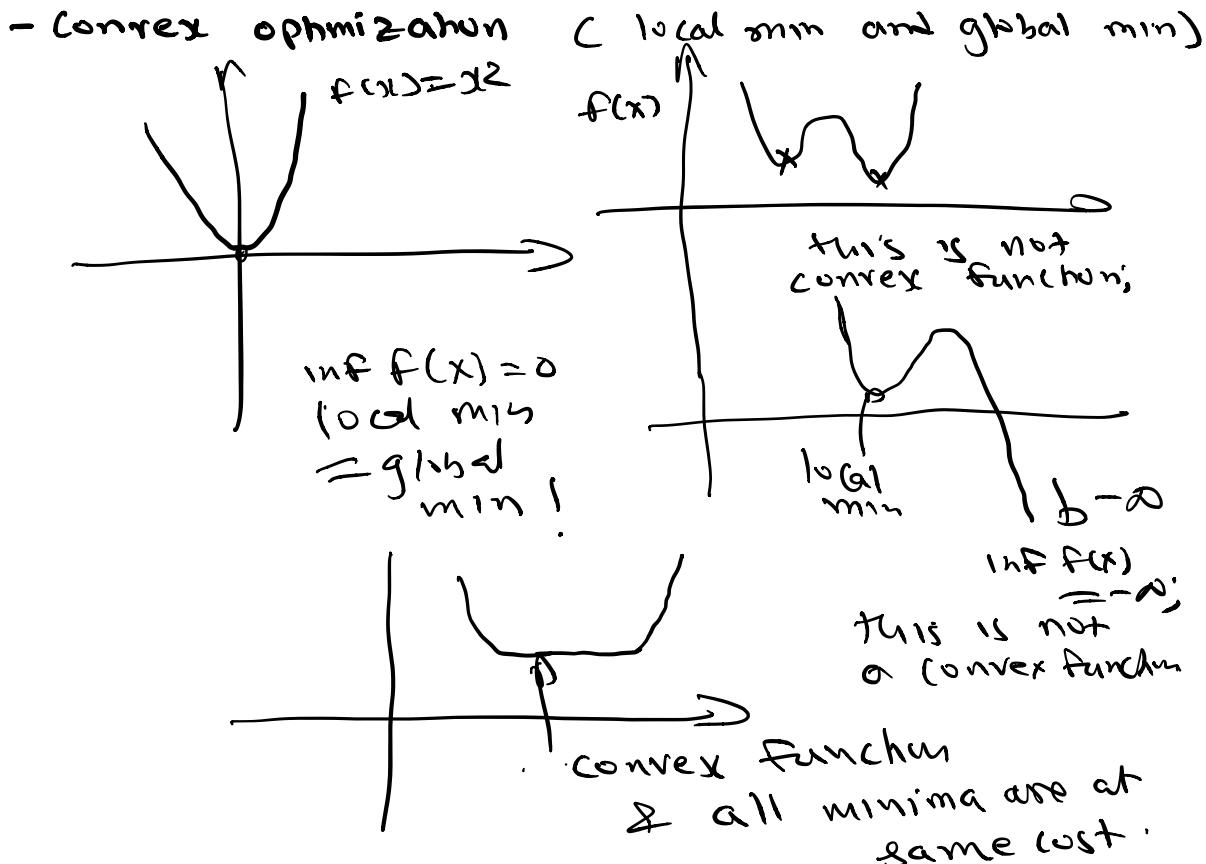


## Lecture -21: Convex Sets and Functions



→ Social welfare maximization problem (convex programming)  
Convex sets & functions.

First define convex sets:

Then convex functions:

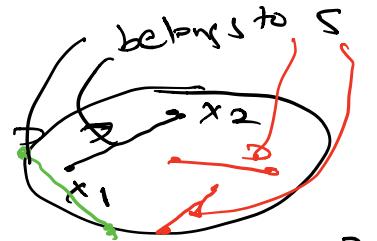
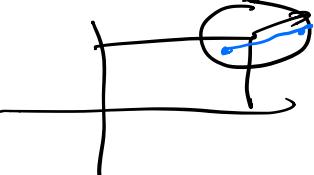
Finally Optimizing functions on  
convex sets which is convex programing.

Defn: A set  $S$  in  $\mathbb{R}^n$  is said to be convex.  
if the line segment joining two points of  
the set, belongs to the set  $S$ .

$$(x_1 - a)^2 + (x_2 - b)^2 \leq r^2$$

↳

also a  
convex set.  
(closed set)



$$S \subset \mathbb{R}^2$$

Convex set.

$$(x_1 - a)^2 + (x_2 - b)^2 \leq r^2$$

Q) Is this convex. Yes?

convex set

$\nsubseteq$  closed open ✓

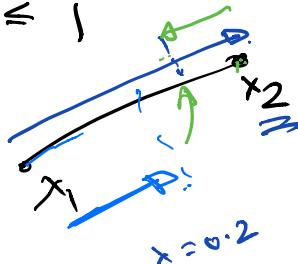
Convex set : let  $x_1 \in S$ ,  $x_2 \in S$ , arbitrary points

$$L = \{x : x = \lambda x_1 + (1-\lambda)x_2 \mid 0 \leq \lambda \leq 1\}$$

should  $\subset S$ .

$$\lambda x_1 + (1-\lambda)x_2$$

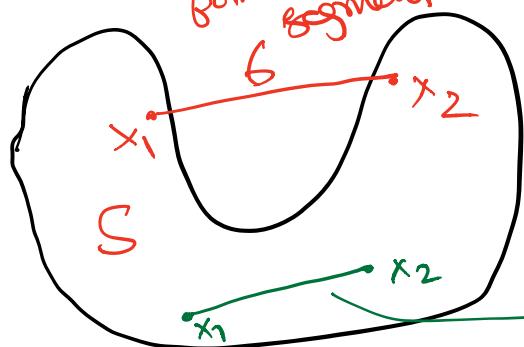
$$= x_2 - \lambda(x_2 - x_1) \rightarrow \lambda \in [0, 1]$$



$$\lambda x_2 + (1-\lambda)x_1 \quad \checkmark \quad \lambda \in [0, 1]$$

↳ alternate description

points on the line segment are not in S



GS ✗

↳ ... created from points

every element

in set  $S$  should be in set  $L$

Affine combinations. ✓

Examples of convex sets:

$$1. \quad S = \{ (x_1, x_2, x_3) : 3x_1 + 2x_2 - x_3 = 4 \}$$

Plane

$$(a_1, a_2, a_3) \in S, \quad (b_1, b_2, b_3) \in S.$$

$$\lambda a + (1-\lambda) b \quad \lambda \in [0, 1]$$

Check that it belongs to  $S$ .

$$\lambda \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} + (1-\lambda) \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

$$x_1 = \lambda a_1 + (1-\lambda) b_1$$

$$x_2 = \lambda a_2 + (1-\lambda) b_2$$

$$x_3 = \lambda a_3 + (1-\lambda) b_3.$$

check  $3x_1 + 2x_2 - x_3 = 4$  ? if yes then  
 $S$  convex

$$3(\lambda a_1 + (1-\lambda) b_1) + 2(\lambda a_2 + (1-\lambda) b_2) + 3(\lambda a_3 + (1-\lambda) b_3)$$

$$\lambda(3a_1 + 2a_2 + 3a_3) + (1-\lambda)(3b_1 + 2b_2 + 3b_3)$$

$$= \lambda 4 + (1-\lambda) 4 = 4$$

$$\lambda \vec{a} + (1-\lambda) \vec{b} \quad \lambda \in [0, 1]$$

$\in$  Plane given,  
and hence

plane  $\rightarrow$  convex set.

$$\underline{\mathbb{R}^n}$$

Hyperplane:

$$S = \{x : p^t x = \alpha\}$$

$$p = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

$$\alpha = 4;$$

$$p_1 x_1 + p_2 x_2 + \dots + p_n x_n = \alpha \text{ a constant.}$$

$$p^t \bar{x} = \alpha$$

$$p^t x = \alpha \Rightarrow p^t \bar{x}$$

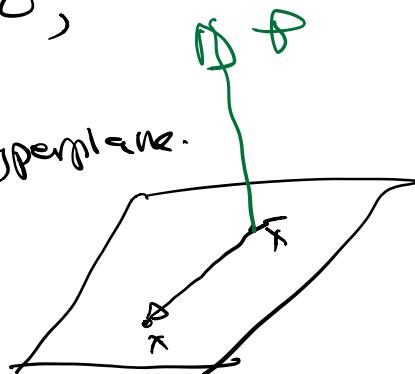
$$p^t(x - \bar{x}) = 0;$$

$$p$$

gradient or normal to the hyperplane.

$$f(x) = p^t x$$

$$\nabla f(x) = p$$



Check:  $S = \{x : p^t x = \alpha\}$  is a convex set.

$$p^t \hat{x} = \alpha \quad \hat{x} \in S$$

$$p^t \bar{x} = \alpha \quad \bar{x} \in S$$

$$x = \lambda \hat{x} + (1-\lambda) \bar{x} \quad \lambda \in [0, 1]$$

$$\begin{aligned} p^t x &= p^t [\lambda \hat{x} + (1-\lambda) \bar{x}] \\ &= \lambda p^t \hat{x} + (1-\lambda) p^t \bar{x} = \alpha \end{aligned}$$

$$\begin{array}{l} p^t \hat{x} = \alpha \\ p^t \bar{x} = \alpha \end{array}$$

$\rho^T x \leq \alpha$  [on the hyperplane]  
 hyperplane is a convex set.

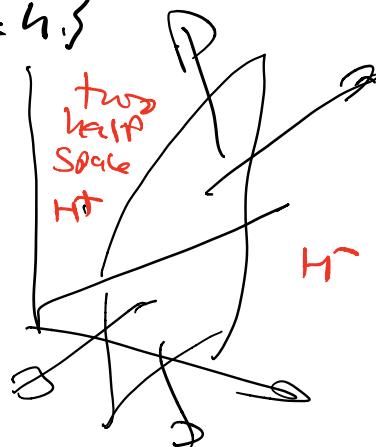
$$S = \{ (x_1, x_2, x_3) : 3x_1 + 2x_2 - x_3 \leq 4 \}$$

half space.

$$H^+ S = \{ x : \rho^T x \leq \alpha \}$$

$$H^- S = \{ x : \rho^T x \geq \alpha \}$$

$$\underline{H^+ \cup H^- = \mathbb{R}^n}$$



a) Is the half space convex?

Yes

$$\rho^T \hat{x} \leq \alpha; \quad \hat{x} \in H^+$$

$$\rho^T \bar{x} \leq \alpha; \quad \bar{x} \in H^-$$

$$x = \lambda \hat{x} + (1-\lambda) \bar{x} \quad \lambda \in (0,1]$$

$$\begin{aligned} \rho^T x &= \lambda \rho^T \hat{x} + (1-\lambda) \rho^T \bar{x} \\ &\leq \lambda \alpha + (1-\lambda) \alpha \\ &= \alpha \end{aligned}$$

$$\rho^T x \leq \alpha$$

Half space is convex.

$$3 \leq 5$$

$$0.2 \times 3 \leq 0.5$$

$$-0.2 \times 3 \geq -0.2 \times 5$$

$$3. S = \{(x_1, x_2, x_3) : x_1 + 2x_2 - x_3 \leq 5 \\ 3x_1 - 2x_2 + x_3 \leq 6\}$$

Intersection of two half spaces.  
This is also convex.

$$\begin{array}{ll} \hat{x} \in S & \bar{x} \in S \\ p_1^T \hat{x} \leq \alpha_1 & p_1^T \bar{x} \leq \alpha_1 \\ p_2^T \hat{x} \leq \alpha_2 & p_2^T \bar{x} \leq \alpha_2. \end{array}$$

$$x : \lambda \hat{x} + (1-\lambda) \bar{x} \\ p_1^T x \in H_1^- \text{ (first half space)} \\ p_2^T x \in H_2^- \text{ (second half space)} \\ \therefore x \in S \}$$

and hence intersection  
of half spaces is a  
convex set.

$$S = \{x : Ax \leq b\}. \text{ also convex}$$

$$Ax \leq b = \left\{ \begin{array}{l} a_1^T x \leq b_1 \\ a_2^T x \leq b_2 \\ \vdots \\ a_m^T x \leq b_m \end{array} \right\}$$

$a_i^T + \text{first row of } A_j$

Intersection of  $m$ -half spaces  
polyhedral set is always convex.

4.  $S = \{x_1, x_2 : x_2 \geq |x_1|\}$

convex cone

$\text{S} - x_1^2 + x_2^2 \leq 9$   
closed disc

