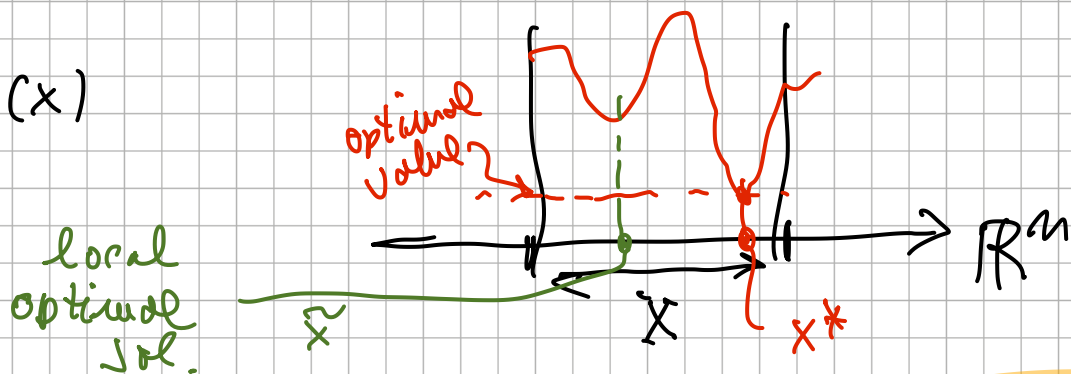


MATHEMATICAL OPT. PROBLEM

$X \subseteq \mathbb{R}^n$ set of feasible sol.s

$f: X \rightarrow \mathbb{R}$ objective function

$\min_{x \in X} f(x)$



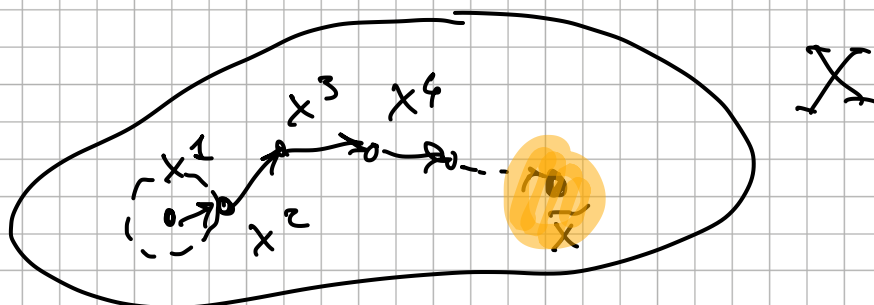
x^* optimal sol.

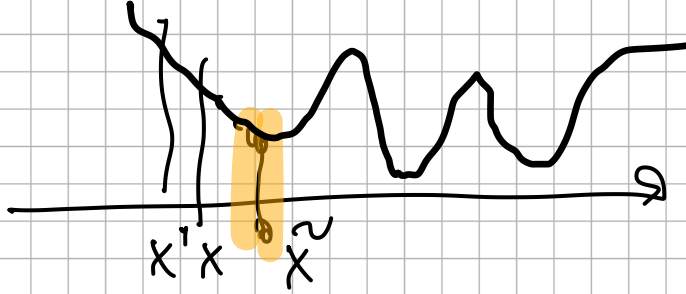
$\max \psi(x) = - \min_{x \in X} -\psi(x)$

3 cases:

- $X = \emptyset$ PROBL. INFEASIBLE "min = $+\infty$ "
- f is not bounded from below on X "min = $-\infty$ "
- $x^* \in X : f(x^*) \leq f(x), \forall x \in X$
"GLOBAL optimal sol."

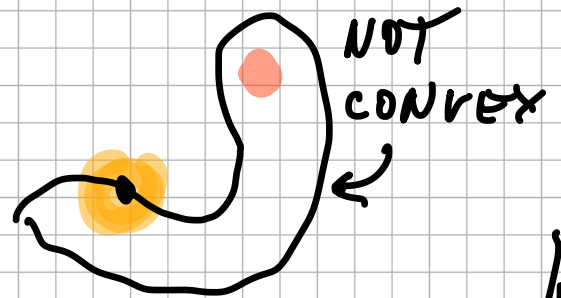
\tilde{x} is locally optimal $\iff \exists \epsilon > 0 :$
 $f(\tilde{x}) \leq f(x), \forall x \in X : \|x - \tilde{x}\| \leq \epsilon$



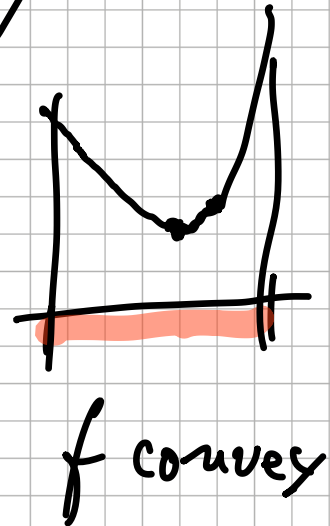
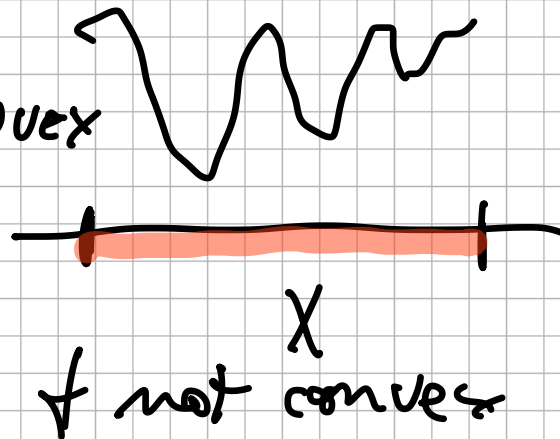


2 PROPERTIES :

• X CONVEX

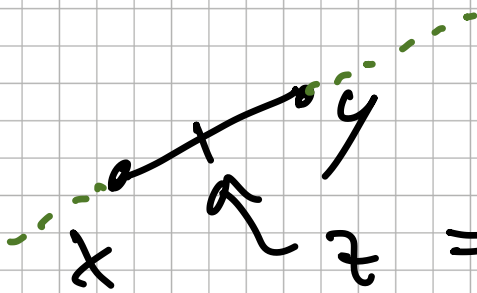


• f CONVEX



Def :

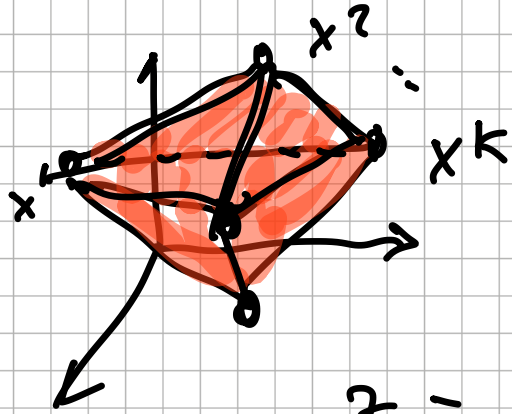
convex comb. of two points $x, y \in \mathbb{R}^n$



$$z = \lambda x + (1-\lambda)y$$

for $\lambda \in [0, 1]$

$\lambda \notin \{0, 1\} \Rightarrow$ STRICT conv. comb.



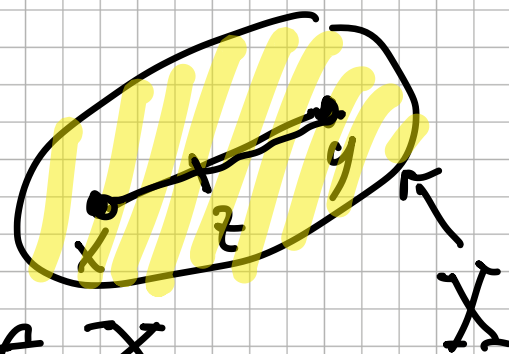
$$z = \sum_{i=1}^k \lambda_i x^i$$

$$\sum_{i=1}^k \lambda_i = 1, \quad \lambda_i \geq 0$$

CONVEX SET

$\forall x, y \in X, \forall \lambda \in [0, 1]:$

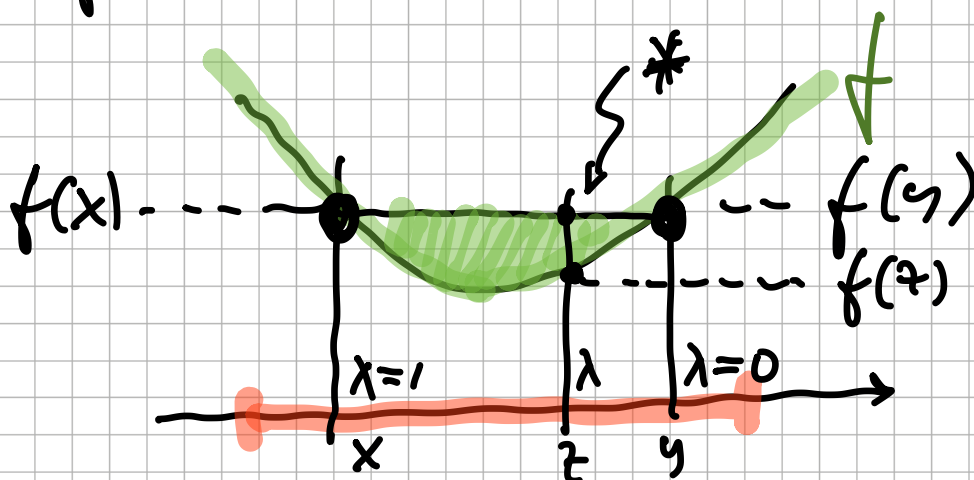
$$z = \lambda x + (1 - \lambda)y \in X$$



- $\bigcap_{i=1}^k X_i = X$ is convex if all X_i 's are convex

CONVEX FUNCTION

$f: X \rightarrow \mathbb{R}$ with X convex

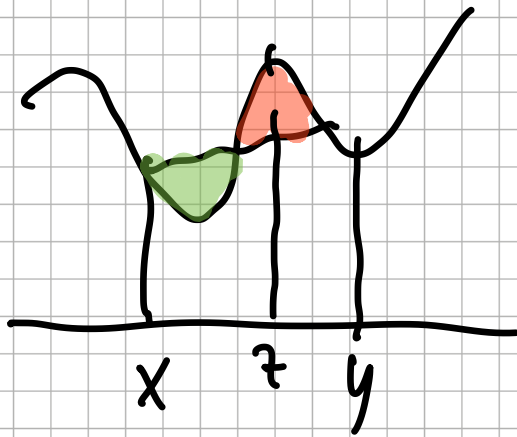


$$z = \lambda x + (1-\lambda)y \quad \text{for } \lambda \in [0,1]$$

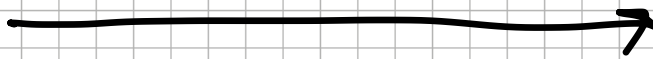
$$(*) = \lambda f(x) + (1-\lambda)f(y)$$

$$f(z) \leq (*) = \lambda f(x) + (1-\lambda)f(y)$$

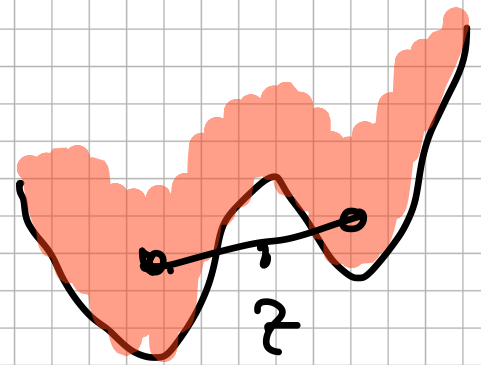
$$f(\underbrace{\lambda x + (1-\lambda)y}_z) \leq \lambda f(x) + (1-\lambda)f(y)$$



NOT convex!



f is convex



f not convex

- $$X = \{x \in \mathbb{R}^n : g_i(x) \leq 0, i=1, \dots, m\}$$

$$g_i : \mathbb{R}^n \rightarrow \mathbb{R} \text{ convex } \forall i$$

$\Rightarrow X$ is convex

Proof: $X = \bigcap_{i=1}^m X_i$ with

$$X_i = \{x \in \mathbb{R}^n : g_i(x) \leq 0\}$$

Take any X_i and any two points $x, y \in X_i$ and any $\lambda \in [0, 1]$:

$$z = \lambda x + (1-\lambda)y$$

by convexity of g_i .

$$g_i(z) = g_i(\lambda x + (1-\lambda)y) \leq$$

$$\lambda g_i(x) + (1-\lambda)g_i(y) \leq 0$$

$\underbrace{\lambda}_{\geq 0} \underbrace{g_i(x)}_{\leq 0} + \underbrace{(1-\lambda)}_{\geq 0} \underbrace{g_i(y)}_{\leq 0} \leq 0$

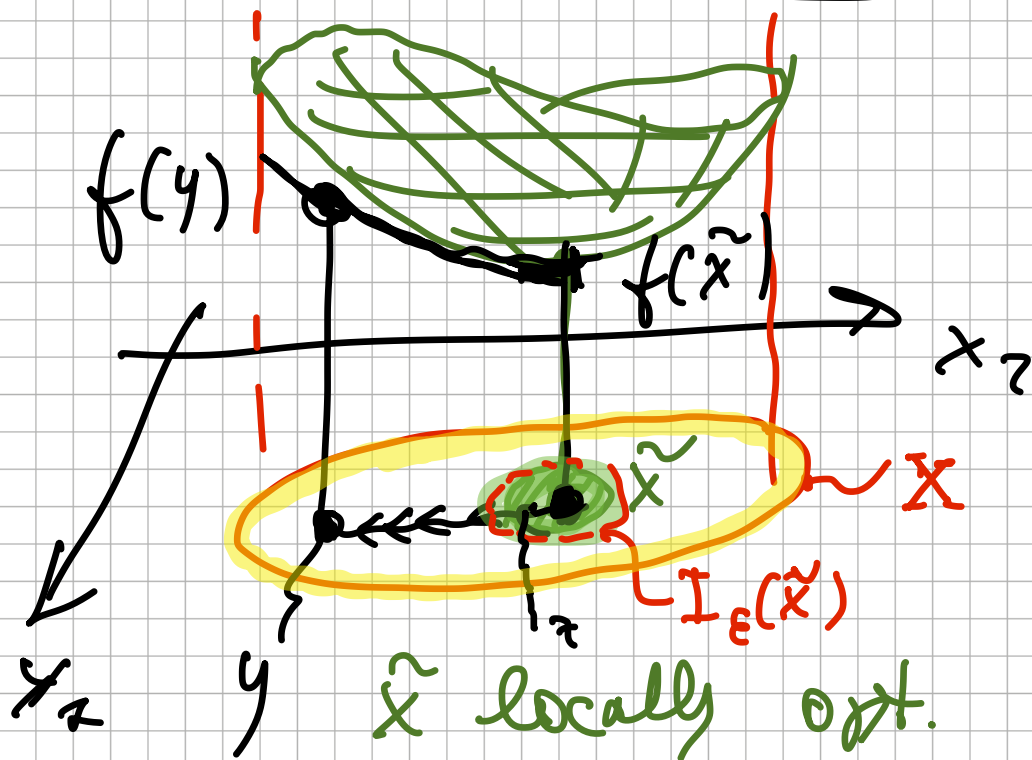
$\Rightarrow \exists \tilde{x} \in X; \Rightarrow X; \text{convex} \quad \square$

THEOREM $\min \{ f(x) : x \in X \}$

with f, X are convex.

Every local optimal sol.

is also globally optimal.



Proof : Let $\tilde{x} \in X$ local optimal sol. Then, $\exists \epsilon > 0$:
 $f(\tilde{x}) \leq f(z), \forall z \in I_\epsilon(\tilde{x})$

Take ANY $y \in \tilde{X}$ and
consider

$$z = \lambda \tilde{x} + (1-\lambda)y$$

with $\lambda \leq 1 \Rightarrow z \in I_{\tilde{x}}(\tilde{X})$

$$\boxed{f(\tilde{x})} \leq f(z) =$$

\uparrow
 $z \in I_{\tilde{x}}(\tilde{X})$

f convex

$$= f(\lambda \tilde{x} + (1-\lambda)y) \geq$$
$$\boxed{\lambda f(\tilde{x}) + (1-\lambda)f(y)} \Rightarrow$$

$$\boxed{f(\tilde{x})} \underbrace{(1-\lambda)}_{\neq 0} \leq \underbrace{(1-\lambda)}_{\neq 0} \boxed{f(y)}$$

□

SPECIAL CASE:

- all $g_i(x)$'s and f are LINEAR \Rightarrow

$$\min f(x)$$

$$g_i(x) \leq 0$$

$$\Rightarrow \left\{ \begin{array}{l} \min c^T x \\ -a_i^T x + b_i \leq 0 \\ -x_i \leq 0, \\ i = 1, \dots, n \end{array} \right.$$

$$c^T x = [c_1, \dots, c_n] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} =$$

$$= \sum_{i=1}^n c_i x_i$$

$$\Rightarrow \min \left\{ c^T x : \begin{array}{l} Ax \geq b \\ x \geq 0 \end{array} \right\}$$

LINEAR OPT. PROBL.