$min f_0(x) \leftarrow objective$

$$\times \in \mathbb{R}^{N}$$
 $\chi_{n \times 1}$

St.
$$f_i(x) \leq 0$$
 $i=1,2,...m$

constraint

or
$$f_i(x) = 0$$

min
$$(x-1)^2$$

2 >0

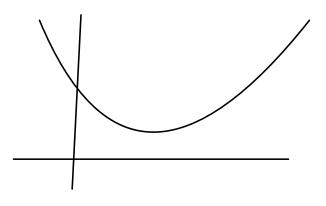
$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad a = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

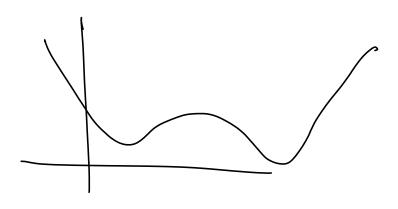
= min
$$(x_1 - 1)^2 + (x_{\bar{2}}2)^2$$

 $x_1 \geqslant 0$, $x_2 \geqslant 0$

$$\underline{X} \cdot \begin{bmatrix} \chi_1^* \\ \chi_2^* \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Convex Optimization





- easy to optimize over mathematical properties
- guaranteed to converge (solution -> optimum)

predictable algorithms

Eg. Least Squares problem

$$x^* = \underset{A \in \mathbb{R}^m}{\text{arg min } } \|Ax - b\|_2^2$$

$$(Ax - b)^T (Ax - b)$$

$$b \in \mathbb{R}^m$$

$$A \in \mathbb{R}^{m \times m}$$

$$(Ax) \in \mathbb{R}^m$$

$$Ax - b \in \mathbb{R}^m$$

$$Ax - b \in \mathbb{R}^m$$

Special case: n=m and A-1 exists

$$x^* = A^{-1}b$$
 so that $||Ax^* - b|| = 0$

$$n \neq m$$
 then $x^* = (A^TA)^{-1}A^Tb$ when $(A^TA)^{-1}$ exists closed form (formula)

-well studied

min
$$c^T x$$
 $x \in \mathbb{R}^m$
 $Ax \leq b \longrightarrow b \in \mathbb{R}^m$

$$A : \begin{bmatrix} -a_1^T - \\ -a_2^T - \end{bmatrix}$$

$$a_i^T x \leq b_i^* \quad i=1,2...m$$

$$\begin{bmatrix} A \end{bmatrix}_{ij} : a_{ij} \\ \lambda_{ij} = a_{ij} \\ \lambda_{ij} = a_{ij} \\ \lambda_{ij} = a_{ij} \\ \lambda_{ij} = b_{ij} \\ \lambda_{ij}$$

$$a_1 x \leq b_1$$
 $a_2 x \leq b_2$
 \vdots
 $a_m x \leq b_m$

Complexity? # floating point operations (eg. \times , \div , + , -) xeR" m>n $\sim O(mn^2)$ $\sqrt{O(N^3)}$ Koughly (size n) = max (# variables, # constraints) # operations \leq (const) $\times n^3$ does not depend on problem structure / condition number

Vs. General optimization problems 0 (en)

 $e^{\text{N}} \gg \text{N}^3 \approx \text{N}^5 \dots$ for large N

e.g. $e^{10} \sim 22026$