

Proof of Th. 3.1:

" \Rightarrow ": Let $x \in \mathbb{R}^n$: $Ax \leq b$

Suppose there is $y \geq 0$ with $y^T A = 0$
 $y^T \cdot b < 0$

$$\Rightarrow 0 > y^T b \geq y^T \cdot (A \cdot x) = (y^T \cdot A) \cdot x = 0 \quad \text{!}$$

" \Leftarrow ": Suppose that $A \cdot x \leq b$ has no solution x .

Prove by induction on # columns of A ($=n$) that there is $y \geq 0$ with $y^T A = 0^T$ and $y^T \cdot b < 0$.

$n=1$: clear.

$n-1 \rightarrow n$: $Ax \leq b$ has no solution

$\Rightarrow A'x' \leq b'$ has no solution

Ind. $\Rightarrow \exists y' \geq 0$: $y'^T A' = 0^T$ and $y'^T \cdot b' < 0$

Notice that $A' = U \cdot A$ for some matrix $U \geq 0$

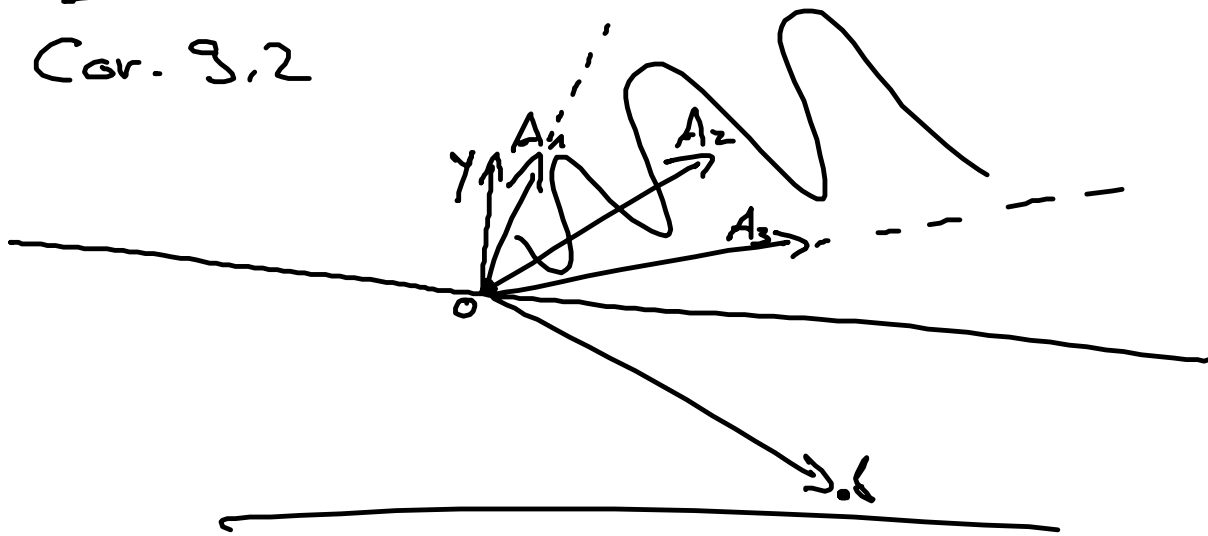
$$b' = U \cdot b$$

Set $y^T := y'^T \cdot U \geq 0$

$$\begin{aligned} \Rightarrow y^T \cdot A &= (y'^T \cdot U) \cdot A = y'^T \cdot (U \cdot A) \\ &= y'^T \cdot A' = 0^T \end{aligned}$$

$$\begin{aligned} y^T \cdot b &= (y'^T \cdot U) \cdot b = y'^T \cdot (U \cdot b) \\ &= y'^T \cdot b' < 0. \end{aligned} \quad \square$$

Cor. 3.2



Proof of Cor. 3.2 (Farkas' Lemma) using LP duality:

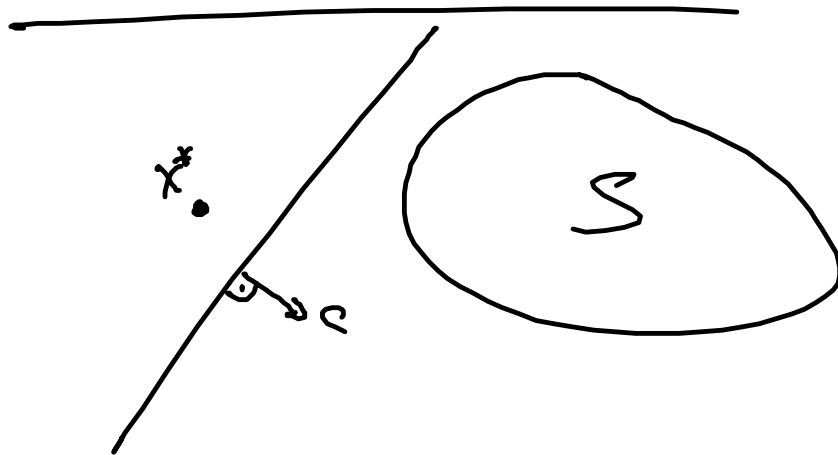
Assume that there is no $x \geq 0 : Ax = b$:

Consider infeasible LP:

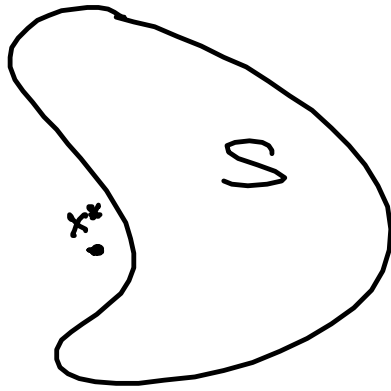
$$\begin{aligned} \max & \quad 0^T \cdot x \\ \text{s.t.} & \quad A \cdot x = b \\ & \quad x \geq 0 \end{aligned}$$

$$\begin{aligned} \min & \quad y^T \cdot b \\ \text{s.t.} & \quad y^T \cdot A \geq 0^T \end{aligned}$$

Dual LP is feasible (e.g. $y=0$) and hence unbounded $\rightarrow \exists y : y^T b < 0, y^T A \geq 0^T$ \square



Theorem 3.3 does not hold for nonconvex S :



Proof of Theorem 3.3:

Let $w \in S$ arbitrary and

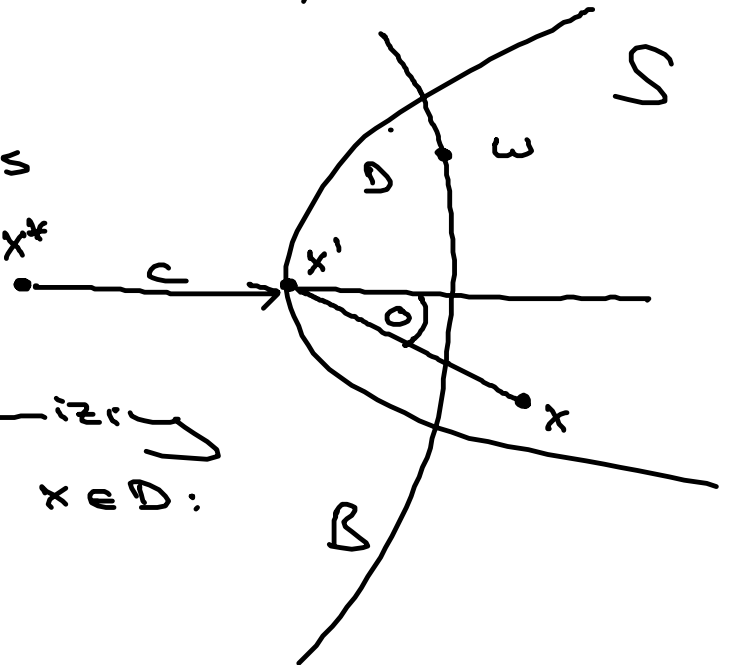
$$B = \{x \mid \|x - x^*\|_2 \leq \|w - x^*\|_2\}$$

Set $D := S \cap B$

Notice that D is bounded, closed and non-empty.

$\Rightarrow \exists x' \in D$ minimizing $\|x - x^*\|_2$ over all $x \in D$:

($f(x) := \|x - x^*\|_2$ continuous)



$$\|x' - x^*\|_2 \leq \|x - x^*\|_2 \quad \forall x \in D$$

even $\forall x \in S$

$$\text{Set } c := x' - x^*$$

Let $x \in S$ and $\lambda \in (0, 1] \Rightarrow x' + \lambda \cdot (x - x') \in S$

$$\begin{aligned} \underline{\|x' - x^*\|_2^2} &\leq \|x' + \lambda \cdot (x - x') - x^*\|_2^2 \\ &= \underline{\|x' - x^*\|_2^2} + 2\lambda \cdot (x' - x^*)^T \cdot (x - x') + \lambda^2 \|x - x'\|_2^2 \end{aligned}$$

$$\Rightarrow 2\lambda \cdot (x' - x^*)^T \cdot (x - x') + \lambda^2 \cdot \|x - x'\|^2 \geq 0$$

$$\text{Let } \lambda \text{ go to } 0 : \underbrace{(x' - x^*)^T}_{c^T} \cdot (x - x') \geq 0$$

$$\Rightarrow \underbrace{(x' - x^*)^T}_{c^T} \cdot x \geq \underbrace{(x' - x^*)^T}_{c^T} \cdot x'$$

$$= c^T \cdot x^* + c^T \cdot \underbrace{(x' - x^*)}_{c}$$

$$> c^T \cdot x^*$$

□

From Theorem 3.3 one can easily obtain Farkas' Lemma (see exercises).

Let $x \neq 0$ with $Ax \geq 0$

$$\text{Write } x = \underbrace{\frac{1}{2} \cdot 0}_{\in \mathcal{P}} + \underbrace{\frac{1}{2} \cdot (2x)}_{\in \mathcal{P}}$$

