

# AA Lecture 19

3/23/21

## Linear Programming Duality

Standard form:  
(primal)

$$\begin{aligned} v^* = \min_{x \geq 0} & c^T x \\ \text{s.t. } & Ax = b \end{aligned}$$

$$\begin{aligned} c &\in \mathbb{R}^n \\ b &\in \mathbb{R}^m \\ A &\in \mathbb{R}^{m \times n} \end{aligned}$$

1) if want to prove  $v^* \leq v$   
just show some  $x \in F$ ,  $c^T x \leq v$ .

2)  $v^* \geq \dots ?$

Def. dual of s.f.LP:

$$\begin{aligned} u^* = \max_y & b^T y \\ \text{s.t. } & A^T y \leq c \end{aligned}$$

$$y \in \mathbb{R}^m.$$

Then [weak duality]:  $u^* \leq v^*$ .

pf: fix  $x \in F$ :  $Ax = b$  and  $x \geq 0$ .

$$\text{fix any } y \in \mathbb{R}^m \quad (y^T A)x = y^T b$$

$\nwarrow$  a vector  $\in \mathbb{R}^n$ .

Suppose  $y$  s.t.  $y^T A \leq c^T$  (each coord.)

$$\Rightarrow \max_{y^T b} = y^T A^T x \leq c^T x$$

$$\Rightarrow \max_{\substack{y^T b \\ y^T A \leq c^T}} \leq \min_{\substack{c^T x \\ x: Ax=b \\ x \geq 0}} = v^*$$

☒

### Dual of the Dual:

Primal:

$$v^* = \min_{x \geq 0} c^T x$$

$$\text{s.t. } Ax = b$$

Dual:

$$v^* = \max_{y} b^T y$$

$$\text{s.t. } A^T y \leq c$$

Dual is standard form:

$$-v^* = \min_{A^T y \leq c} -b^T y = \begin{cases} \min_{y^+, y^-} -b^T(y^+ - y^-) \\ A^T(y^+ - y^-) + \delta = c \\ y^+, y^-, \delta \geq 0 \end{cases}$$

$y^+, y^- \in \mathbb{R}^m$

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

Dual of the dual:

unknowns:  $z \in \mathbb{R}^n$

objective func:  $\max_{z} c^T z$

$A^T$	$-A^T$	$I_n$
$y^+$	$y^-$	$\delta$

Constraints:

$$\begin{pmatrix} A \\ -A \\ I \end{pmatrix} \cdot z \leq \begin{pmatrix} -b \\ +b \\ 0 \end{pmatrix}$$

$$A_i^T z = b_i \Leftrightarrow \begin{cases} A_i^T z \leq -b_i \\ -A_i^T z \leq +b_i \\ z_i \leq 0 \end{cases} \quad A_i = i^{th} \text{ row of } A$$

$$z' = -z.$$

Dual of (-Dual):

$$\begin{aligned} & \max c^T(-z') \\ \text{s.t. } & A z' = b \\ & z' \geq 0 \end{aligned}$$

Dual of the Dual:

$$\begin{aligned} & -\max -c^T z' = \min c^T z' \\ \text{s.t. } & A z' = b \\ & z' \geq 0. \end{aligned}$$

= Primal.

Strong duality:

Theorem:  $w^* = v^*$ .

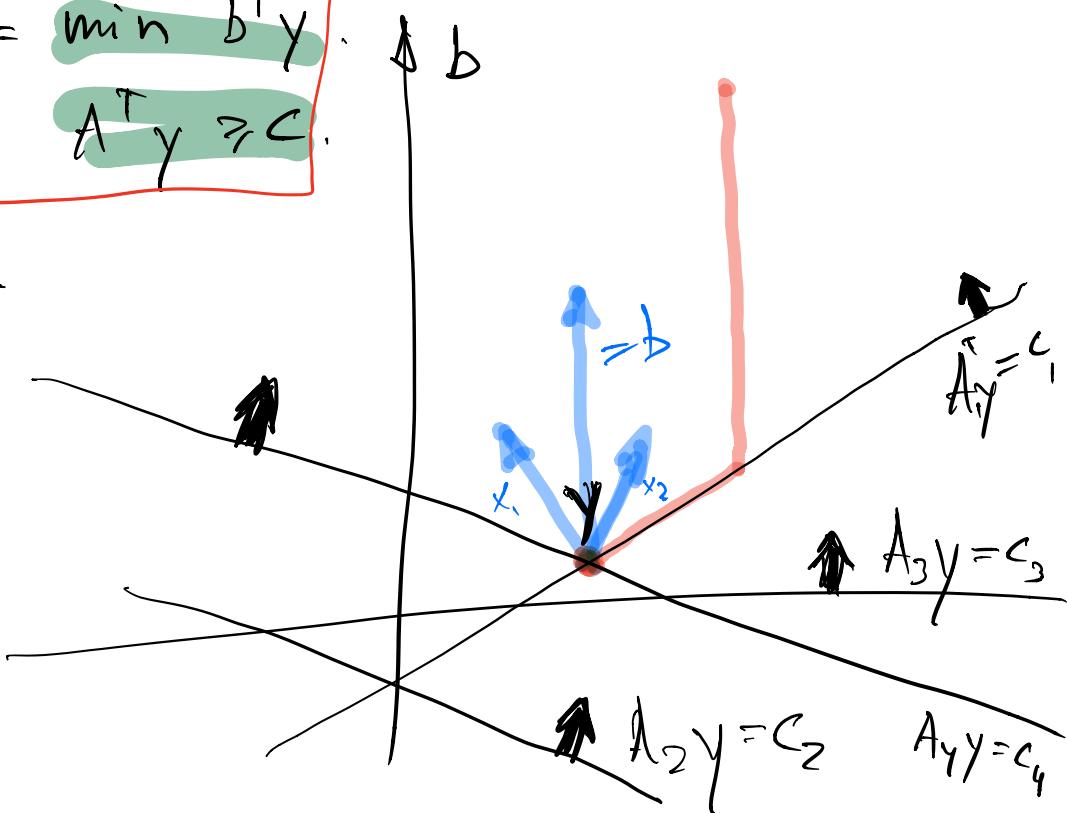
$$\boxed{\begin{aligned} v^* &= \min c^T x \\ Ax &= b, x \geq 0 \\ w^* &= \max b^T y \\ A^T y &\leq c. \end{aligned}}$$

Intuition:

Consider dual: with  $y \mapsto -y$   
 $c \mapsto -c$

$$\boxed{\begin{aligned} -u^* = \min_{y \in \mathbb{R}^m} & b^T y \\ \text{s.t. } & A^T y \geq c \end{aligned}}$$

$$y \in \mathbb{R}^m.$$



What does it mean that ball is resting?

$\Rightarrow$  a set  $S$  of constraints/half-spaces:

1)  $i \in S \Rightarrow A_i y = c_i$  (tight).

2)  $\exists$  coefficients  $x_1 \dots x_n \in \mathbb{R}_+$

$$\sum x_i A_i = b \Leftrightarrow Ax = b.$$

2') if  $i \notin S$ ,  $x_i = 0$ .

Value of primal:  $-c^T x$

dual:  $-b^T y$ .

$$\sum x_i A_i y = y^T A x \stackrel{\textcircled{1}}{=} \sum x_i c_i = c^T x.$$

$$\stackrel{\textcircled{2}}{=} y^T b = b^T y.$$

OK

Note: for every  $i \in [n]$ : must have

$x_i = 0$  or  
 $A_i y = c_i$  (or both).

(complementary slackness).

Ellipsoid Algorithm: [Khachiyan '79].

Solves feasibility problem.

$$\begin{aligned} \text{LP: } & \min_{x^*} c^T x \\ & Ax \geq b. \end{aligned}$$

$$Q_v = \{x : Ax \geq b, c^T x \leq v\}.$$

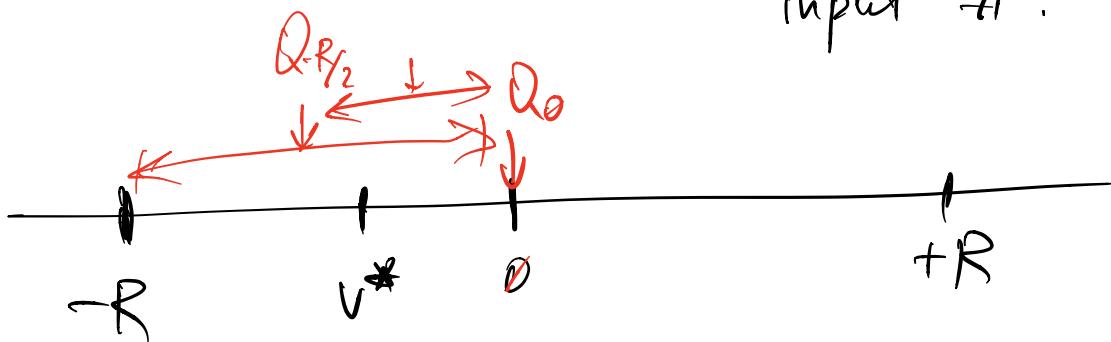
FP: find  $x \in Q_v$  if exists.

Remark: can solve LP using FP with poly-time slope-down.

ff: binary search on V.

Let  $R = \text{upper bound on value of } |v^*|$ .

$$\lg_2 R \leq \text{poly}(n, m) \cdot B \xrightarrow{\text{# bits to represent each input #.}}$$



BS: fix  $\ell = -R, r = +R$

$$m = \frac{\ell + r}{2}$$

check if  $Q_m$  is feasible

if yes  $\Rightarrow$  recurse on  $\ell = \ell$   
 $r = m$ .

if no  $\Rightarrow$  recurse on  $\ell = m$   
 $r = r$ .

stop when  $r - \ell$  is sufficiently small,  $\propto \frac{1}{R^{O(1)}}$ .

(then opt. sol.  $v^* \in \mathcal{P}(x, v)$ )

# BS iterations =  $\underline{\mathcal{O}(\log R)} = \underline{\mathcal{O}(\text{poly}(n, m)B)}$

Solving  $Q_v = \left\{ x; \begin{array}{l} Ax \geq b \\ C^T x \leq v \end{array} \right\}$ .

