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# Agenda

- Duality Review
  - Motivation example
  - Weak/Strong duality, duality for LPs
  - Farkas Lemma
- Taking the dual example
- Dual of  $\ell_1$  norm example

## Logistics

- HW 5 out, due Friday March 25 at 9pm
- HW 6 due Friday April 1 at 9pm
- Midterm 2 on Tuesday April 12

## Duality motivation / obtaining lower bounds

- We want to get a lower bound for

$$\begin{array}{ll}\min_x & x_1 + 3x_2 \\ \text{s.t.} & x_1 + x_2 \geq 2 \\ & x_2 \geq 1 \\ & x_1 - x_2 \geq 3\end{array}$$

- One idea: add the constraints, match cost

$$\begin{aligned} & y_1(x_1 + x_2 \geq 2) \\ & + y_2(x_2 \geq 1) \\ & + y_3(x_1 - x_2 \geq 3) \\ \xrightarrow{\substack{\text{match} \\ \text{these}}} \quad & (y_1 + y_3)x_1 + (y_1 + y_2 - y_3)x_2 \geq \underbrace{2y_1 + y_2 + 3y_3}_{\substack{\text{think: dual objective}}} \\ \xrightarrow{\text{Cost!}} \quad & x_1 + 3x_2 \end{aligned}$$

$$\Rightarrow \underbrace{y_1 + y_3 = 1, \quad y_1 + y_2 - y_3 = 3}_{\substack{\text{think: dual constraints}}}, \quad y \geq 0$$

$$y = (1, 2, 0) \rightarrow \text{bound} = 4$$

$$y = (0, 4, 1) \rightarrow \text{bound} = 7$$

## Farkas Lemma

Given  $A$  and  $b$ , one of the following 2 statements true

1.  $\exists x$  s.t.  $Ax=b, x \geq 0$
2.  $\exists y$  s.t.  $A^T y \geq 0, b^T y < 0$

- 1 and 2 can't both be true : simple

Assume  $x \geq 0, Ax=b, y^T A \geq 0$

$$\Rightarrow y^T b = \underbrace{y^T A x}_{\geq 0} \geq 0 \Rightarrow y^T b \geq 0$$

- 1 and 2 can't both be false : harder, need duality

Primal	Dual
$\min O$	$\max -b^T y$
s.t. $Ax=b$	s.t. $A^T y \geq 0$
$x \geq 0$	

Recall		Primal	Dual
		$p^* = \infty$	$d^* = \infty$
		primal inf	dual vnb
		$p^*$ finite	$d^*$ finite
			opt. vals equal
		$p^* = -\infty$	$d^* = -\infty$
		exception	primal unbd, dual inf.

dual:  $y \geq 0$  always feas.

$$\Rightarrow d^* = 0 \text{ or}$$

$$d^* = \infty$$

$$d^* = 0: \Rightarrow 1. \text{ is true}$$

$d^* = \infty$  iff 2:

$\exists y$  s.t.  $A^T y \geq 0, b^T y < 0$

2. holds

# Dual of std LP

$$\begin{array}{ll} \min_x & C^T x \\ \text{s.t. } & Ax = b \\ & x \geq 0 \end{array}$$

$$\begin{array}{ll} \max_y & -b^T y \\ \text{s.t. } & A^T y + c \geq 0 \end{array}$$

Inequality form

$$\begin{array}{ll} \min_x & C^T x \\ \text{s.t. } & Ax \leq b \end{array}$$

$$\begin{array}{ll} \max_y & -b^T y \\ \text{s.t. } & A^T y + c = 0 \\ & y \geq 0 \end{array}$$

Example

$$\begin{array}{ll} \min_x & 3x_1 + 4x_2 \\ \text{s.t. } & x_1 + x_2 \geq 5 \\ & 2x_1 + x_2 \geq 6 \\ & x \geq 0 \end{array}$$

Pattern match!

$$A = \begin{pmatrix} -1 & -1 \\ -2 & -1 \end{pmatrix} \quad A^T = \begin{pmatrix} -1 & -2 \\ -1 & -1 \end{pmatrix}$$

$$\begin{array}{ll} \min_x & 3x_1 + 4x_2 \\ \text{s.t. } & -x_1 - x_2 \leq -5 \\ & -2x_1 - x_2 \leq -6 \\ & x \geq 0 \end{array}$$

$$\begin{array}{ll} \max_y & 5y_1 + 6y_2 \\ \text{s.t. } & -y_1 - 2y_2 + 3 = 0 \\ & -y_1 - y_2 + 4 = 0 \\ & y \geq 0 \end{array}$$

Other way: derive the dual

$$\begin{array}{ll} \min_x & C^T x \\ \text{s.t. } & Ax \geq b \end{array}$$

formulate Lagrangian:  $L(x, y) = C^T x + y^T (Ax - b)$   
 $\text{s.t. } y \geq 0$

$$\text{dual fn: } g(y) = \min_x C^T x + y^T (Ax - b) \quad (\text{dom}(g) : y \geq 0)$$

$$\begin{aligned} &= \min_x (C + A^T y)^T x - b^T y \\ &= \begin{cases} -\infty & \text{if } C + A^T y \neq 0 \\ -b^T y & \text{if } C + A^T y = 0 \end{cases} \end{aligned}$$

Dual prob:

$$\max_y g(y) \quad \rightarrow$$

$$\max_y -b^T y \quad \text{s.t. } y \geq 0, C + A^T y = 0$$

# Dual of 1 norm problem

$$\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$$

s.t.  $\|\mathbf{Ax} + \mathbf{b}\|_1 \leq 1$

a. formulate this LP inequality form

$$\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$$

s.t.  $s_1 + \dots + s_m \leq 1$

$$s \geq \mathbf{Ax} + \mathbf{b}$$

$$s \geq -\mathbf{Ax} - \mathbf{b}$$

$$\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$$

s.t.  $s_1 + \dots + s_m \leq 1$

$$\mathbf{Ax} - \mathbf{s} \leq \mathbf{b}$$

$$-\mathbf{Ax} - \mathbf{s} \leq \mathbf{b}$$

b. derive the dual LP and show that it's equivalent to

$$\max_{\mathbf{z}} \mathbf{b}^T \mathbf{z} - \|\mathbf{z}\|_\infty$$

s.t.  $\mathbf{A}^T \mathbf{z} + \mathbf{c} = 0$

Recall

$$\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x} \quad \xleftarrow{\text{dual}} \quad \max_{\mathbf{y}} -\mathbf{b}^T \mathbf{y}$$

s.t.  $\mathbf{A} \mathbf{x} \leq \mathbf{b}$

$\mathbf{y} \geq 0$

$$\begin{pmatrix} \mathbf{x} \\ \mathbf{s} \end{pmatrix} = \tilde{\mathbf{x}}$$

$$\min_{\tilde{\mathbf{x}}} \tilde{\mathbf{c}}^T \tilde{\mathbf{x}}$$

s.t.  $\tilde{\mathbf{A}} \tilde{\mathbf{x}} \leq \tilde{\mathbf{b}}$

$$\tilde{\mathbf{A}} = \begin{pmatrix} \mathbf{A} & -\mathbf{I} \\ -\mathbf{A} & -\mathbf{I} \\ \mathbf{0} & \mathbf{I}^T \end{pmatrix}, \quad \tilde{\mathbf{b}} = \begin{pmatrix} -\mathbf{b} \\ \mathbf{b} \\ \mathbf{1} \end{pmatrix}$$

$$\tilde{\mathbf{c}} = \begin{pmatrix} \mathbf{c} \\ \mathbf{0} \end{pmatrix}$$

dual:

$$\max_{\tilde{\mathbf{y}}} -\tilde{\mathbf{b}}^T \tilde{\mathbf{y}}$$

s.t.  $\tilde{\mathbf{y}} \geq 0$

$$\tilde{\mathbf{A}}^T \tilde{\mathbf{y}} + \tilde{\mathbf{c}} = 0$$

$$\tilde{\mathbf{y}} = \begin{pmatrix} \mathbf{y} \\ \mathbf{z} \\ \mathbf{w} \end{pmatrix}$$

$$\max_{\mathbf{y}} \mathbf{b}^T \mathbf{y} - \mathbf{b}^T \mathbf{z} - \mathbf{w}$$

s.t.  $\mathbf{y}, \mathbf{z}, \mathbf{w} \geq 0$

$$(\mathbf{S}) + \begin{pmatrix} \mathbf{A}^T & -\mathbf{A}^T & \mathbf{0} \\ -\mathbf{I} & -\mathbf{I} & \mathbf{0} \end{pmatrix} \begin{pmatrix} \mathbf{y} \\ \mathbf{z} \\ \mathbf{w} \end{pmatrix} = 0$$

$$\max_{y, z, w} b^T y - b^T z - w$$

s.t.  $y, z, w \geq 0$

$$\begin{pmatrix} c \\ 0 \end{pmatrix} + \begin{pmatrix} A^T & -A^T & 0 \\ -I & -I & I \end{pmatrix} \begin{pmatrix} y \\ z \\ w \end{pmatrix} = 0 \quad \left. \begin{array}{l} A^T y - A^T z + c = 0 \\ -y - z + Iw = 0 \end{array} \right\}$$

$$\max_{y, z, w} b^T(y - z) - w$$

s.t.  $y, z, w \geq 0$

$$A^T(y - z) + c = 0$$

$$y + z = Iw$$

Let  $\tilde{z} = y - z$

$\rightarrow \boxed{\|\tilde{z}\|_\infty \leq w}$  because the max absolute difference between 2 non-neg. vectors can't be larger than  $w$

$\rightarrow$  also use the fact that  $y_i + z_i = w$

$$\begin{matrix} w \\ \geq 0 \end{matrix} \quad \begin{matrix} \tilde{z}_i \\ \geq 0 \end{matrix} \Rightarrow y_i \leq w, z_i \leq w \quad \forall i$$

$$\Rightarrow |y_i - z_i| \leq w \quad \forall i$$

$$\Rightarrow \|\tilde{z}\|_\infty \leq w$$

$\rightarrow$  In this case, we will get  $\boxed{\|\tilde{z}\|_\infty = w}$  since at least one element of  $y$  or  $z$  is zero. If this weren't true, then  $y, z, w$  wouldn't be optimal because we could decrease  $y$  and  $z$  by the same amount to improve the objective

$$\max_{\tilde{z}} b^T \tilde{z} - \|\tilde{z}\|_\infty$$

s.t.  $A^T \tilde{z} + c = 0$  ✓

c. give a direct argument that whenever  $x$  is primal feasible and  $z$  is dual feasible,  $c^T x \geq b^T z - \|z\|_\infty$

2 arguments

1. weak duality

2. Recall primal and dual

$$\min_x c^T x \\ \text{s.t. } \|Ax + b\|_1 \leq 1$$

$$\max_z b^T z - \|z\|_\infty \\ \text{s.t. } A^T z + c = 0$$

Use Cauchy-Schwarz  $U^T V \leq \|U\|_1 \|V\|_\infty$

$$(Ax + b)^T z \leq \underbrace{\|Ax + b\|_1}_{\leq 1} \|z\|_\infty$$

$$\Rightarrow (Ax + b)^T z \leq \|z\|_\infty$$

$$b^T z - \|z\|_\infty \leq -x^T A^T z = c^T x$$