

LECTURE 10: LP DUALITY

PRIMAL

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

WHERE $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $c \in \mathbb{R}^n$

- n VARIABLES
- m CONSTRAINTS

DUAL

$$\begin{aligned} q^* = \sup \quad & b^T y \\ \text{s.t.} \quad & A^T y \leq c \\ & y \in \mathbb{R}^m \end{aligned}$$

- 
- m VARIABLES
 - n CONSTRAINTS

LAGRANGIAN DUALITY

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

$$\left[\begin{aligned} \min f(x) \\ \text{s.t. } g_i(x) &\leq 0, i=1, \dots, m \\ x &\in \mathbb{R}^n \end{aligned} \right]$$

$Ax=b$ $x \geq 0$

- LAGRANGIAN RELAX CONSTRAINTS " $Ax=b$ "
- INTRODUCE LAGRANGIAN DUAL FUNCTION

$$q(y) = \min_{x \geq 0} [c^T x + y^T (b - Ax)]$$

$$= \min_{x \geq 0} [(c - A^T y)^T x + y^T b]$$

$$= b^T y + \underbrace{\min_{x \geq 0} [(c - A^T y)^T x]}_{-\infty \text{ UNLESS } c - A^T y \geq 0}$$

$$= \begin{cases} b^T y & \text{IF } c - A^T y \geq 0 \\ -\infty & \text{OTHERWISE} \end{cases}$$

LAGRANGIAN DUAL PROBLEM:

$$q^* = \sup_{y \text{ FREE}} q(y) = \boxed{\begin{aligned} \sup b^T y \\ \text{s.t. } A^T y &\leq c \\ y &\in \mathbb{R}^m \end{aligned}}$$

DUAL PROBLEMS FOR NON-STANDARD FORM LPS

PRIMAL

OBJECTIVE: min

VARIABLES : ≥ 0 [CANONICAL]
 ≤ 0 [NON-CANONICAL]
 FREE

CONSTRAINTS : $\geq$ [CANONICAL]
 $\leq$
 $=$

DUAL

OBJECTIVE: max

CONSTRAINTS: $\leq$
 $\geq$
 $=$

VARIABLES : ≥ 0
 ≤ 0
 FREE

EX.

PRIMAL

$$\min \quad 2x_1 + 3x_2 + x_3$$

$$\text{s.t.} \quad x_1 + x_2 + x_3 \leq 1 \quad (y_1)$$

$$2x_1 \quad + 4x_3 = 4 \quad (y_2)$$

$$x_1 \geq 0$$

$$x_2 \leq 0$$

$$x_3 \text{ FREE}$$

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 0 & 4 \end{bmatrix} \quad A^T = \begin{bmatrix} 1 & 2 \\ 1 & 0 \\ 1 & 4 \end{bmatrix}$$

$$b = \begin{bmatrix} 1 \\ 4 \end{bmatrix} \quad c = \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}$$

DUAL

$$\max \quad y_1 + 4y_2$$

$$y_1 + 2y_2 \leq 2 \quad [x_1]$$

$$y_1 \geq 3 \quad [x_2]$$

$$y_1 + 4y_2 = 1 \quad [x_3]$$

$$y_1 \leq 0 \quad [\text{NON-CANONICAL}]$$

$$y_2 \text{ FREE}$$

PRIMAL/DUAL PAIR

$$\begin{array}{ll} (P) & z^* = \inf c^T x \\ & \text{s.t. } Ax = b \\ & x \geq 0 \end{array}$$

$$\begin{array}{ll} (D) & q^* = \sup b^T y \\ & \text{s.t. } A^T y \leq c \\ & y \in \mathbb{R}^m \end{array}$$

- ALL THEOREMS WILL ASSUME THIS PAIR
- ANALOGOUS RESULTS FOR ALL LP DUAL PAIRS

WEAK DUALITY THEOREM

IF x IS A FEASIBLE POINT IN (P)
AND y IS A FEASIBLE POINT IN (D) THEN

$$c^T x \geq b^T y$$

PROOF:

$$\begin{array}{ccc} c^T x \geq y^T A x = y^T b = \underline{\underline{b^T y}} \\ \Rightarrow \uparrow & & \uparrow \\ \left[\begin{array}{l} A^T y \leq c \\ x \geq 0 \end{array} \right] & & [Ax = b] \end{array}$$

- IT ALSO FOLLOWS THAT

$$z^* \geq q^*$$

COROLLARY:

- IF THE OPTIMAL OBJECTIVE VALUE OF THE PRIMAL PROBLEM (P) IS $-\infty$. THEN THE DUAL PROBLEM (D) IS INFEASIBLE
- IF THE OPT. OBJ. VALUE OF THE DUAL (D) IS $+\infty$ THEN THE PRIMAL PROBLEM (P) IS INFEASIBLE

COROLLARY:

IF x IS FEASIBLE IN (P) AND y IS FEASIBLE
IN THE DUAL PROBLEM (D) AND

$$C^T x = b^T y.$$

THEN x IS OPTIMAL IN (P) AND
 y IS OPTIMAL IN (D) .

- USEFUL FOR OPTIMALITY GUARANTEES.

STRONG DUALITY THEOREM:

IF BOTH (P) AND (D) ARE FEASIBLE, THEN

1. THERE EXISTS x^* OPTIMAL IN (P) AND y^* OPTIMAL IN (D)

2. $c^T x^* = b^T y^*$ (MEANING $z^* = q^*$)

POSSIBILITIES IN P/D

D

	FINITE OPT	UNBOUNDED	INFEASIBLE
FINITE OPT	X		
UNBOUNDED			X
INFEASIBLE		X	X

↑
CAN HAPPEN

SIMPLEX METHOD

- BFS , $x = \begin{bmatrix} x_B \\ x_N \end{bmatrix}$, $A = [B, N]$, $c = \begin{bmatrix} c_B \\ c_N \end{bmatrix}$

- FOR A SPECIFIC BFS ($A = [B, N]$)

$$x_B = B^{-1}b$$

$$x_N = 0$$

- SIMPLEX METHOD ITERATIVELY UPDATES (B, N) UNTIL OPTIMALITY.

- WHEN OPTIMAL $\tilde{c}_N^T = c_N^T - c_B^T B^{-1} N \geq 0$

- INTRODUCE $y^B = (c_B^T B^{-1})^T$

$$- y^* = (C_B^T B^{-1})^T \quad \begin{array}{c} \text{OBJ. VAL} \\ \downarrow \\ \text{IN } (D) \end{array}$$

$$- \text{BY DEFINITION } \underline{b^T y^*} = C_B^T \underbrace{(B^{-1}b)}_{x_B} = C_B^T x_B = C_B^T x_B + \underbrace{C_N^T x_N}_{=0} = \underline{C^T x^*}$$

SO THEY HAVE SAME OBJECTIVE VALUE!

~ BUT IS y^* FEASIBLE IN (D) ?

$$\left. \begin{array}{l} C_B^T - (y^*)^T B = 0 \\ \underbrace{C_N^T - (y^*)^T N}_{\tilde{C}_N^T} \geq 0 \end{array} \right\} \Rightarrow C^T - (y^*)^T A \geq 0$$

$$\Rightarrow \underline{A^T y^* \leq C}$$

FEASIBLE IN (D)

y^* OPTIMAL
IN (D)

REMARKS

- IF YOU SOLVE (P) WITH SIMPLEX METHOD.

YOU GET AN OPTIMAL SOLUTION TO THE DUAL PROBLEM (D)

FOR FREE. $y^* = (C_B^T B^{-1})^T$

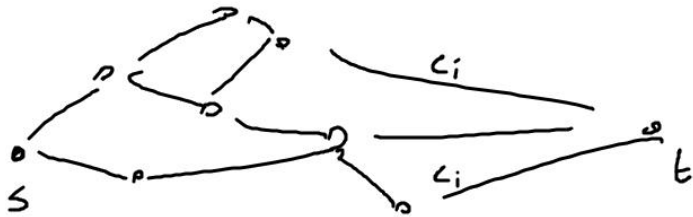
- DIRECTLY PROVIDE OPTIMALITY CERTIFICATE TO YOUR BOSS!

WHAT IS THE DUAL PROBLEM?

EXAMPLE:

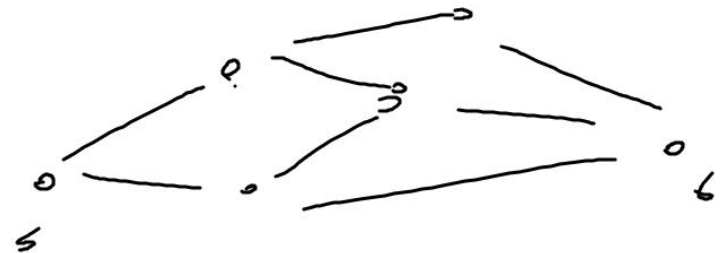
MINIMUM CUT PROBLEM

- A GRAPH WITH EDGE CAPACITIES.
- HOW MUCH CAPACITY DO YOU NEED TO CUT AWAY IN ORDER FOR THERE TO BE NO WAY FROM s TO t



MAXIMUM FLOW PROBLEM

- FIND MAXIMUM FLOW BETWEEN s AND t



COMPLEMENTARITY SLACKNESS THEOREM

LET x BE FEASIBLE IN (P) AND y FEASIBLE IN (D). THEN

$$\left. \begin{array}{l} x \text{ OPTIMAL IN (P)} \\ y \text{ OPTIMAL IN (D)} \end{array} \right\} \Leftrightarrow x_j (c_j - A_{\cdot j}^T y) = 0, \quad j=1, \dots, n$$

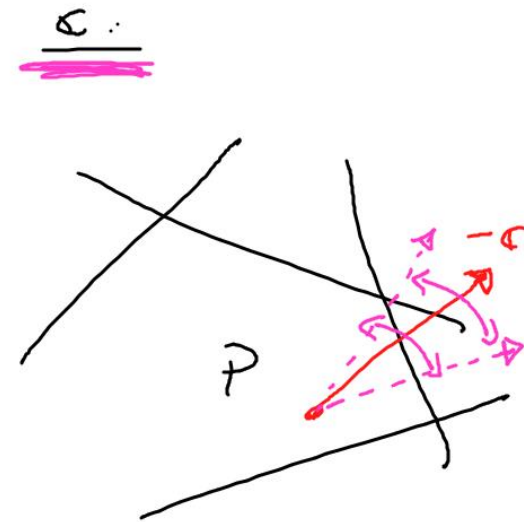
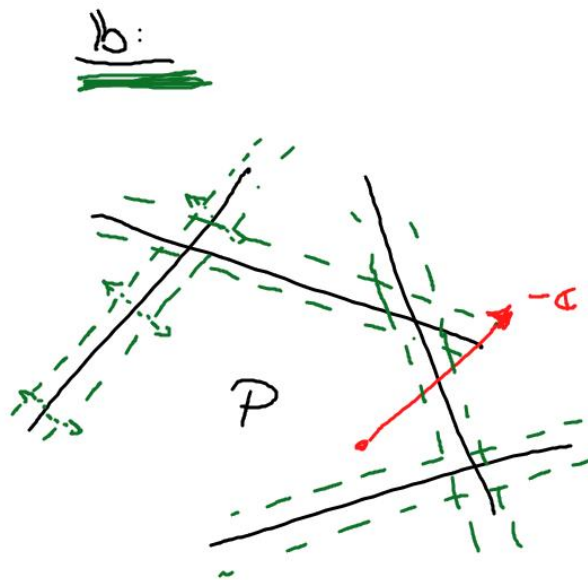
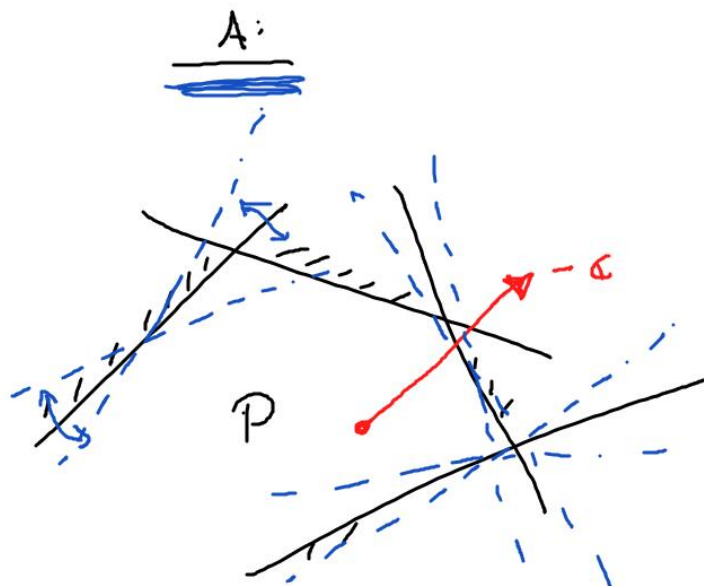
WHERE $A_{\cdot j}$ IS j TH COLUMN OF A

- IF THERE IS "SLACK" IN THE DUAL CONSTRAINT THEN THE CORRESPONDING PRIMAL VARIABLE MUST BE 0
- IF THE PRIMAL VARIABLE IS POSITIVE, THEN THE CORR. DUAL CONSTRAINT MUST FULFILLED WITH EQUALITY.

PERTURBATIONS

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

A, b, c



PERTURBATIONS IN b

$$v(b) = \min_{\substack{x \\ \text{s.t. } Ax = b \\ x \geq 0}} c^T x \quad (P)$$

$$v(b') = \min_{\substack{x \\ \text{s.t. } Ax = b' \in [b + \Delta b] \\ x \geq 0}} c^T x$$

- CONSIDER PERTURBING b TO $b + \Delta b = b'$, $v(b')$?

- SUPPOSE FOR OPTIMAL SOLUTION TO (P), WE HAVE * BFS $x = \begin{bmatrix} x_B \\ x_N \end{bmatrix}$
AND THAT $x_B = B^{-1}b \geq 0$ [NON-DEGENERATE]

- THEN IF Δb IS SMALL $x'_B = B^{-1}b' = B^{-1}b + B^{-1}\Delta b = x_B + B^{-1}\Delta b \geq 0$

S. STILL FEASIBLE

- AND IT WILL STILL BE OPTIMAL

$$\underline{v(b')} = c_B^T x'_B = \underbrace{c_B^T B^{-1}}_{y^{*T}} b' = \underline{(y^*)^T b'}$$

⇒

SHADOW PRICE THEOREM

IF, FOR A GIVEN b , THE OPTIMAL BFS OF (P) IS NON-DEGENERATE, THEN

$$\nabla V(b) = y^*$$

- IT MEANS THAT CHANGING THE RIGHT HAND SIDE OF (P) WILL AFFECT IN A CHANGE IN OBJECTIVE VALUE WHICH IS y^* .
- SO CHANGING i TH RIGHT-HAND-SIDE WILL AFFECT THE OPT. OBJECTIVE VALUE WITH y_i^* .

PERTURBATIONS IN c

$$\begin{array}{ll} \min & (c + \Delta c)^T x \\ \text{s.t.} & Ax = b \\ & x \geq 0 \end{array}$$

- LET $A = [B, N]$ BE OPTIMAL BFS WHEN $\Delta c = 0$.

- OPTIMALITY CONDITION

$$\tilde{c}_N^T = c_N^T - c_B^T B^{-1} N \geq 0$$

- WHEN $\Delta c \neq 0$.

$$\tilde{c}_N^T = (c_N + \Delta c_N)^T - (c_B + \Delta c_B)^T B^{-1} N \geq 0 \quad ?$$

$\uparrow$
EASY TO CHECK
IF OPTIMALITY STILL HOLDS

$\uparrow$
MORE COMPUTATIONS
TO BE DONE

