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# **ECE 307- Techniques for Engineering Decisions**

## **Lecture 4. Duality Concepts in Linear Programming**

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# DUALITY

- **Definition:** A *LP* is in *symmetric form* if all the variables are restricted to be *nonnegative* and all the constraints are inequalities of the type:

<i>objective type</i>	<i>corresponding inequality type</i>
<i>max</i>	$\leq$
<i>min</i>	$\geq$

# DUALITY DEFINITIONS

□ We first define the *primal* and *dual* problems

$$\begin{array}{ll} \max & Z = \underline{c}^T \underline{x} \\ \text{s.t.} & \underline{Ax} \leq \underline{b} \\ & \underline{x} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \max \\ \text{s.t.} \end{array}} \right\} (P)$$

$$\begin{array}{ll} \min & W = \underline{b}^T \underline{y} \\ \text{s.t.} & \underline{A}^T \underline{y} \geq \underline{c} \\ & \underline{y} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} (D)$$

# DUALITY DEFINITIONS

- The problems  $(P)$  and  $(D)$  are called the *symmetric dual LP* problems; we restate them as

$$\left. \begin{array}{l} \max Z = c_1 x_1 + c_2 x_2 + \dots + c_n x_n \\ s.t. \\ a_{11} x_1 + a_{12} x_2 + \dots + a_{1n} x_n \leq b_1 \\ a_{21} x_1 + a_{22} x_2 + \dots + a_{2n} x_n \leq b_2 \\ \vdots \\ a_{m1} x_1 + a_{m2} x_2 + \dots + a_{mn} x_n \leq b_m \\ x_1 \geq 0, \quad x_2 \geq 0, \quad \dots, \quad x_n \geq 0 \end{array} \right\} (P)$$

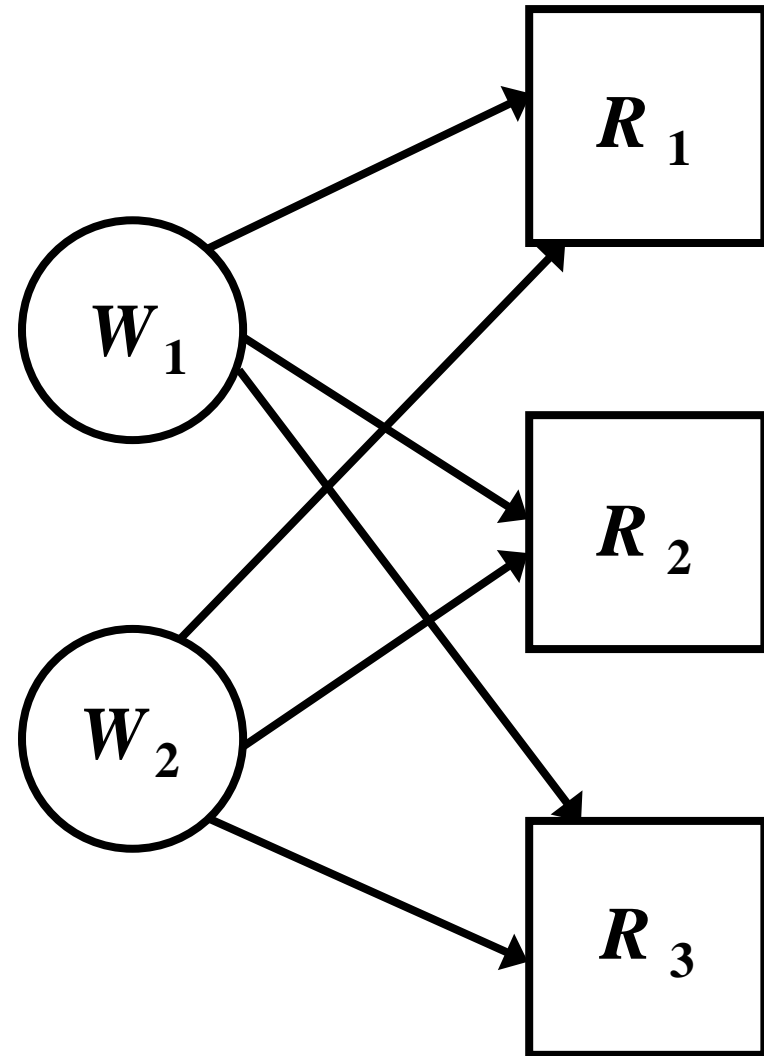
# DUALITY DEFINITIONS

$$\left. \begin{array}{l} \min W = b_1 y_1 + b_2 y_2 + \dots + b_m y_m \\ \\ s.t. \\ \\ a_{11} y_1 + a_{21} y_2 + \dots + a_{m1} y_m \geq c_1 \\ \\ a_{12} y_1 + a_{22} y_2 + \dots + a_{m2} y_m \geq c_2 \\ \\ \vdots \\ \\ a_{1n} y_1 + a_{2n} y_2 + \dots + a_{mn} y_m \geq c_n \\ \\ y_1 \geq 0, \quad y_2 \geq 0, \quad \dots, \quad y_m \geq 0 \end{array} \right\} (D)$$

# EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM

*shipment cost coefficients*

<i>warehouses</i>	<i>retail stores</i>		
	$R_1$	$R_2$	$R_3$
$W_1$	2	4	3
$W_2$	5	3	4



# EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM

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□ We are given that the *supplies* stored in warehouses

$W_1$  and  $W_2$  satisfy

$$\text{supply at } W_1 \leq 300$$

$$\text{supply at } W_2 \leq 600$$

□ We are also given the *demands needed* to be met at

the retail stores  $R_1$ ,  $R_2$ , and  $R_3$  :

$$\text{demand at } R_1 \geq 200$$

$$\text{demand at } R_2 \geq 300$$

$$\text{demand at } R_3 \geq 400$$

# EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM

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□ The problem is to determine the *least-cost shipping schedule*

□ We define the decision variable

$x_{ij}$  = *quantity shipped from  $W_i$  to  $R_j$   $i = 1, 2$ ,  $j = 1, 2, 3$*

□ The shipping costs may be viewed as

$c_{ij}$  = *element  $i, j$  of the transportation cost matrix*



# FORMULATION STATEMENT

$$\min Z = \sum_{i=1}^2 \sum_{j=1}^3 c_{ij} x_{ij} = 2x_{11} + 4x_{12} + 3x_{13} + 5x_{21} + 3x_{22} + 4x_{23}$$

*s.t.*

$$x_{11} + x_{12} + x_{13} \leq 300$$

$$x_{21} + x_{22} + x_{23} \leq 600$$

$$x_{11} + x_{21} \geq 200$$

$$x_{12} + x_{22} \geq 300$$

$$x_{13} + x_{23} \geq 400$$

$$x_{ij} \geq 0 \quad i = 1, 2, j = 1, 2, 3$$

# DUAL PROBLEM SETUP USING SYMMETRIC FORM

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$$\min Z = \sum_{i=1}^2 \sum_{j=1}^3 c_{ij} x_{ij}$$

*s.t.*

$$y_1 \leftrightarrow -x_{11} - x_{12} - x_{13} \geq -300$$

$$y_2 \leftrightarrow -x_{21} - x_{22} - x_{23} \geq -600$$

$$y_3 \leftrightarrow x_{11} + x_{21} \geq 200$$

$$y_4 \leftrightarrow x_{12} + x_{22} \geq 300$$

$$y_5 \leftrightarrow x_{13} + x_{23} \geq 400$$

$$x_{ij} \geq 0 \quad i=1,2 \quad j=1,2,3$$

# DUAL PROBLEM SETUP

$$\max W = -300y_1 - 600y_2 + 200y_3 + 300y_4 + 400y_5$$

*s.t.*

$$-y_1 + y_3 \leq c_{11} = 2$$

$$-y_1 + y_4 \leq c_{12} = 4$$

$$-y_1 + y_5 \leq c_{13} = 3$$

$$-y_2 + y_3 \leq c_{21} = 5$$

$$-y_2 + y_4 \leq c_{22} = 3$$

$$-y_2 + y_5 \leq c_{23} = 4$$

$$y_i \geq 0 \quad i = 1, 2, \dots, 5$$

# THE *DUAL PROBLEM* INTERPRETATION

□ The moving company proposes to the manufacturer to:

buy all the 300 units at  $W_1$  at  $y_1 / unit$

buy all the 600 units at  $W_2$  at  $y_2 / unit$

sell all the 200 units at  $R_1$  at  $y_3 / unit$

sell all the 300 units at  $R_2$  at  $y_4 / unit$

sell all the 400 units at  $R_3$  at  $y_5 / unit$

□ To convince the manufacturer to get the business, the mover ensures that the delivery fees cannot exceed the transportation costs the manufacturer would incur (the dual constraints)

# THE *DUAL PROBLEM* INTERPRETATION

$$-y_1 \quad + y_3 \leq c_{11} = 2$$

$$-y_1 \quad + y_4 \leq c_{12} = 4$$

$$-y_1 \quad + y_5 \leq c_{13} = 3$$

$$-y_2 + y_3 \leq c_{21} = 5$$

$$-y_2 + y_4 \leq c_{22} = 3$$

$$-y_2 + y_5 \leq c_{23} = 4$$

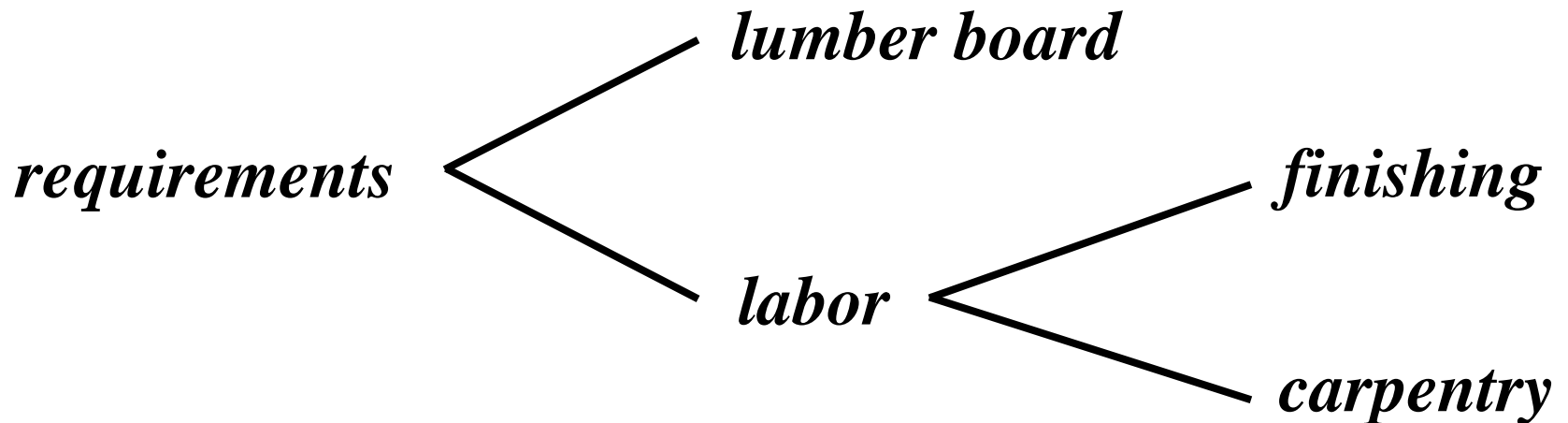
□ The mover wishes to maximize profits, i.e.,  
*revenues – costs*  $\Rightarrow$  *dual cost objective function*

$$\max W = -300 y_1 - 600 y_2 + 200 y_3 + 300 y_4 + 400 y_5$$

# EXAMPLE 2: FURNITURE PRODUCTS

## ❑ Resource requirements

<i>item</i>	<i>sales price (\$)</i>
<i>desks</i>	<b>60</b>
<i>tables</i>	<b>30</b>
<i>chairs</i>	<b>20</b>



# EXAMPLE 2: FURNITURE PRODUCTS

- ❑ The Dakota Furniture Company manufacturing:

<i>resource</i>	<i>desk</i>	<i>table</i>	<i>chair</i>	<i>available</i>
<i>lumber board (ft )</i>	8	6	1	48
<i>finishing (h)</i>	4	2	1.5	20
<i>carpentry (h)</i>	2	1.5	0.5	8

- ❑ We assume that the demand for desks, tables and chairs is unlimited and the available resources are already purchased
- ❑ The decision problem is to maximize *total revenues*

# PRIMAL AND DUAL PROBLEM FORMULATION

□ We define decision variables

$x_1 = \text{number of desks produced}$

$x_2 = \text{number of tables produced}$

$x_3 = \text{number of chairs produced}$

□ The Dakota problem is

$$\max Z = 60x_1 + 30x_2 + 20x_3$$

*s.t.*

$$y_1 \leftrightarrow 8x_1 + 6x_2 + x_3 \leq 48 \quad \text{lumber}$$

$$y_2 \leftrightarrow 4x_1 + 2x_2 + 1.5x_3 \leq 20 \quad \text{finishing}$$

$$y_3 \leftrightarrow 2x_1 + 1.5x_2 + 0.5x_3 \leq 8 \quad \text{carpentry}$$

$$x_1, x_2, x_3 \geq 0$$



# PRIMAL AND DUAL PROBLEM FORMULATION

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□ The dual problem is

$$\min \quad W = 48y_1 + 20y_2 + 8y_3$$

*s.t.*

$$8y_1 + 4y_2 + 2y_3 \geq 60 \quad \text{desk}$$

$$6y_1 + 2y_2 + 1.5y_3 \geq 30 \quad \text{table}$$

$$y_1 + 1.5y_2 + 0.5y_3 \geq 20 \quad \text{chair}$$

$$y_1, y_2, y_3 \geq 0$$

# PRIMAL AND DUAL PROBLEM FORMULATION

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$$\max \quad Z = 60x_1 + 30x_2 + 20x_3$$

$$y_1 \leftrightarrow 8x_1 + 6x_2 + x_3 \leq 48 \quad \text{lumber}$$

$$y_2 \leftrightarrow 4x_1 + 2x_2 + 1.5x_3 \leq 20 \quad \text{finishing}$$

$$y_3 \leftrightarrow 2x_1 + 1.5x_2 + 0.5x_3 \leq 8 \quad \text{carpentry}$$

$$x_1, x_2, x_3 \geq 0$$

$$\max \quad W = 48y_1 + 20y_2 + 8y_3$$

$$48y_1 + 20y_2 + 8y_3 \geq 60 \quad \text{desk}$$

$$6y_1 + 2y_2 + 1.5y_3 \geq 30 \quad \text{table}$$

$$y_1 + 1.5y_2 + 0.5y_3 \geq 20 \quad \text{chair}$$

$$y_1, y_2, y_3 \geq 0$$

# INTERPRETATION OF THE DUAL PROBLEM

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- ❑ An entrepreneur wishes to purchase all of Dakota's resources
- ❑ He needs, therefore, to determine the prices to pay for each unit of each resource
  - $y_1 = \text{price paid for 1 lumber board ft}$
  - $y_2 = \text{price paid for 1 h of finishing labor}$
  - $y_3 = \text{price paid for 1 h of carpentry labor}$
- ❑ We solve the Dakota dual problem to determine  $y_1, y_2$  and  $y_3$

# INTERPRETATION OF THE DUAL PROBLEM

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- ❑ To induce Dakota to sell the raw resources, the resource prices must be set sufficiently high
- ❑ For example, the entrepreneur must offer Dakota at least \$ 60 for a combination of resources that consists of 8 *ft* of lumber board, 4 *h* of finishing and 2 *h* of carpentry, since Dakota could use this combination to sell a desk for \$ 60: this requirement implies the following dual constraint:

$$8y_1 + 4y_2 + 2y_3 \geq 60$$

# INTERPRETATION OF DUAL PROBLEM

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- ❑ In the same way, we obtain the two additional constraints for a table and for a chair
- ❑ The  $i^{th}$  primal variable is associated with the  $i^{th}$  constraint in the dual problem statement
- ❑ The  $j^{th}$  dual variable is associated with the  $j^{th}$  constraint in the primal problem statement

# EXAMPLE 3: DIET PROBLEM

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- ☐ A new diet requires that all food eaten come from one of the four “basic food groups”:
  - ☐ chocolate cake
  - ☐ ice cream
  - ☐ soda
  - ☐ cheesecake
- ☐ The four foods available for consumption are
  - ☐ brownie
  - ☐ cola
  - ☐ chocolate ice cream
  - ☐ pineapple cheesecake

# EXAMPLE 3: DIET PROBLEM

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- ❑ Minimum requirements for each day are:
  - 500 *cal*
  - 6 *oz* chocolate
  - 10 *oz* sugar
  - 8 *oz* fat
- ❑ The objective is to minimize the diet costs

# EXAMPLE 3: DIET PROBLEM

<i>food</i>	<i>calories</i>	<i>chocolate</i> (oz)	<i>sugar</i> (oz)	<i>fat</i> (oz)	<i>costs</i> (cents)
<i>brownie</i>	<b>400</b>	<b>3</b>	<b>2</b>	<b>2</b>	<b>50</b>
<i>chocolate</i> <i>ice cream</i> (scoop)	<b>200</b>	<b>2</b>	<b>2</b>	<b>4</b>	<b>20</b>
<i>cola</i> (bottle)	<b>150</b>	<b>0</b>	<b>4</b>	<b>1</b>	<b>30</b>
<i>pineapple</i> <i>cheesecake</i> (piece)	<b>500</b>	<b>0</b>	<b>4</b>	<b>5</b>	<b>80</b>



# PROBLEM FORMULATION

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- ❑ Objective of the problem is to minimize the total costs of the diet
- ❑ Decision variables are defined for each day's purchases

$x_1 = \textit{number of brownies}$

$x_2 = \textit{number of chocolate ice cream scoops}$

$x_3 = \textit{number of bottles of soda}$

$x_4 = \textit{number of pineapple cheesecake pieces}$

# PROBLEM FORMULATION

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□ The problem statement is

$$\min \quad Z = 50 x_1 + 20 x_2 + 30 x_3 + 80 x_4$$

*s.t.*

$$400 x_1 + 200 x_2 + 150 x_3 + 500 x_4 \geq 500 \text{ cal}$$

$$3 x_1 + 2 x_2 \geq 6 \text{ oz}$$

$$2 x_1 + 2 x_2 + 4 x_3 + 4 x_4 \geq 10 \text{ oz}$$

$$2 x_1 + 4 x_2 + x_3 + 5 x_4 \geq 8 \text{ oz}$$

$$x_i \geq 0 \quad i = 1, 4$$

# EXAMPLE 3: DIET PROBLEM

□ The dual problem is

$$\max \quad W = 500 y_1 + 6 y_2 + 10 y_3 + 8 y_4$$

*s.t.*

$$400 y_1 + 3 y_2 + 2 y_3 + 2 y_4 \leq 50 \quad \text{brownie}$$

$$200 y_1 + 2 y_2 + 2 y_3 + 4 y_4 \leq 20 \quad \text{ice-cream}$$

$$150 y_1 + 4 y_3 + y_4 \leq 30 \quad \text{soda}$$

$$500 y_1 + 4 y_3 + 5 y_4 \leq 80 \quad \text{cheesecake}$$

$$y_1, y_2, y_3, y_4 \geq 0$$

# INTERPRETATION OF THE DUAL

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- We consider a salesperson of “nutrients” who is interested in assuming that each dieter meets daily requirements by purchasing calories, sugar, fat and chocolate as “goods”
- The decision is to determine the prices charged  
 $y_i$  = *price per unit of required nutrient to sell to dieters*
- Objective of the salesperson is to set the prices  $y_i$  so as to maximize revenues from selling to the dieter the daily ration of required nutrients

# INTERPRETATION OF DUAL

- ❑ Now, the dieter can purchase a brownie for 50 ¢ and have 400 *cal*, 3 *oz* of chocolate, 2 *oz* of sugar and 2 *oz* of fat
- ❑ The sales price  $y_i$  must be set sufficiently low to entice the buyer to get the required nutrients from the brownie:  
$$400y_1 + 3y_2 + 2y_3 + 2y_4 \leq 50 \longleftarrow \begin{array}{l} \text{brownie} \\ \text{constraint} \end{array}$$
- ❑ We derive similar constraints for the ice cream, the soda and the cheesecake

# DUAL PROBLEMS

$$\begin{array}{ll} \max & Z = \underline{c}^T \underline{x} \\ \text{s.t.} & \underline{A} \underline{x} \leq \underline{b} \\ & \underline{x} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \max \\ \text{s.t.} \end{array}} \right\} (P)$$
$$\begin{array}{ll} \min & W = \underline{b}^T \underline{y} \\ \text{s.t.} & \underline{A}^T \underline{y} \geq \underline{c} \\ & \underline{y} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} (D)$$

# WEAK DUALITY THEOREM

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□ For any  $\underline{x}$  feasible for  $(P)$  and any  $\underline{y}$  feasible for  $(D)$

$$\underline{c}^T \underline{x} \leq \underline{b}^T \underline{y} \quad \square$$

□ Proof:

$$\underline{A}^T \underline{y} \geq \underline{c} \Rightarrow \underline{c}^T \leq \underline{y}^T \underline{A} \Rightarrow \underline{c}^T \underline{x} \leq \underline{y}^T \underline{A} \underline{x}$$

$$\underline{c}^T \underline{x} \leq \underline{y}^T \underline{A} \underline{x} \leq \underline{y}^T \underline{b} = \underline{b}^T \underline{y} \quad \square$$

# COROLLARY 1 OF THE WEAK DUALITY THEOREM

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$$\underline{x} \text{ is feasible for } (P) \Rightarrow \underline{c}^T \underline{x} \leq \underline{y}^T \underline{b}$$

*for any feasible  $\underline{y}$  for  $(D)$*

$$\underline{c}^T \underline{x} \leq \underline{y}^{*T} \underline{b} = \min W$$

*for any feasible  $\underline{x}$  for  $(P)$ ,*

$$\underline{c}^T \underline{x} \leq \min W$$



# COROLLARY 2 OF THE WEAK DUALITY THEOREM

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$$\underline{y} \text{ is feasible for } (D) \Rightarrow \underline{c}^T \underline{x} \leq \underline{y}^T \underline{b}$$

*for every feasible  $\underline{x}$  for  $(P)$*

$$\max Z = \max \underline{c}^T \underline{x} = \underline{c}^T \underline{x}^* \leq \underline{y}^T \underline{b}$$

*for any feasible  $\underline{y}$  of  $(D)$ ,*

$$\underline{y}^T \underline{b} \geq \max Z$$

# COROLLARIES 3 AND 4 OF THE WEAK DUALITY THEOREM

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If  $(P)$  is feasible and  $\max Z$  is unbounded, i.e.,

$$Z \rightarrow +\infty ,$$

then,  $(D)$  has no feasible solution.

If  $(D)$  is feasible and  $\min Z$  is unbounded, i.e.,

$$Z \rightarrow -\infty ,$$

then,  $(P)$  is infeasible.

# DUALITY THEOREM APPLICATION

□ Consider the maximization problem

$$\begin{aligned} \max Z &= x_1 + 2x_2 + 3x_3 + 4x_4 = \underbrace{[1, 2, 3, 4]}_{\underline{c}^T} \underline{x} \\ \text{s.t.} \quad & \underbrace{\begin{bmatrix} 1 & 2 & 2 & 3 \\ 2 & 1 & 3 & 2 \end{bmatrix}}_{\underline{A}} \underline{x} \leq \underbrace{\begin{bmatrix} 20 \\ 20 \end{bmatrix}}_{\underline{b}} \\ & \underline{x} \geq \underline{0} \end{aligned} \quad \left. \vphantom{\begin{aligned} \max Z &= x_1 + 2x_2 + 3x_3 + 4x_4 = \underbrace{[1, 2, 3, 4]}_{\underline{c}^T} \underline{x} \\ \text{s.t.} \quad & \underbrace{\begin{bmatrix} 1 & 2 & 2 & 3 \\ 2 & 1 & 3 & 2 \end{bmatrix}}_{\underline{A}} \underline{x} \leq \underbrace{\begin{bmatrix} 20 \\ 20 \end{bmatrix}}_{\underline{b}} \\ & \underline{x} \geq \underline{0} \end{aligned}} \right\} (P)$$

# DUALITY THEOREM APPLICATION

□ The corresponding dual is given by

$$\begin{array}{ll} \min & W = \underline{b}^T \underline{y} \\ s.t. & \\ & \underline{A}^T \underline{y} \geq \underline{c} \\ & \underline{y} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \min \\ s.t. \\ \underline{A}^T \underline{y} \geq \underline{c} \\ \underline{y} \geq \underline{0} \end{array}} \right\} (D)$$

□ With the appropriate substitutions, we obtain

# DUALITY THEOREM APPLICATION

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$$\min \quad W = 20 y_1 + 20 y_2$$

*s.t.*

$$y_1 + 2 y_2 \geq 1$$

$$2 y_1 + y_2 \geq 2$$

$$2 y_1 + 3 y_2 \geq 3$$

$$3 y_1 + 2 y_2 \geq 4$$

$$y_1 \geq 0, \quad y_2 \geq 0$$

# GENERALIZED FORM OF THE DUAL

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□ Consider the primal decision

$$x_i = 1, \quad i = 1, 2, 3, 4 ;$$

decision is feasible for  $(P)$  with

$$Z = \underline{c}^T \underline{x} = 10$$

□ The dual decision

$$y_i = 1, \quad i = 1, 2$$

is feasible for  $(D)$  with

$$W = \underline{b}^T \underline{y} = 40$$

# DUALITY THEOREM APPLICATION

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□ Clearly,

$$Z(x_1, x_2, x_3, x_4) = 10 \leq 40 = W(y_1, y_2)$$

and so clearly, the feasible decision for  $(P)$  and  $(D)$

satisfy the *Weak Duality Theorem*

□ Moreover, we have

$$\text{corollary 1} \Rightarrow 10 \leq \min W = W(y_1^*, y_2^*)$$

$$\text{corollary 2} \Rightarrow \max Z = Z(x_1^*, x_2^*, x_3^*, x_4^*) \leq \underline{b}^T \underline{y} = 40$$

# COROLLARIES 5 AND 6

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**$(P)$  is feasible and  $(D)$  is infeasible, then,**

**$(P)$  is unbounded**



**$(D)$  is feasible and  $(P)$  is infeasible, then,**

**$(D)$  is unbounded**





# EXAMPLE

□ Consider the primal dual problems:

$$\left. \begin{array}{l} \max Z = x_1 + x_2 \\ s.t. \\ -x_1 + x_2 + x_3 \leq 2 \\ -2x_1 + x_2 - x_3 \leq 1 \\ x_1, x_2, x_3 \geq 0 \end{array} \right\} (P)$$
$$\left. \begin{array}{l} \min W = 2y_1 + y_2 \\ s.t. \\ -y_1 - 2y_2 \geq 1 \\ y_1 + y_2 \geq 1 \\ y_1 - y_2 \geq 0 \\ y_1, y_2 \geq 0 \end{array} \right\} (D)$$

□ Now

$\underline{x} = \underline{0}$  is feasible for  $(P)$

# EXAMPLE

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$\underline{x} = \underline{0}$  is feasible for  $(P)$

but

$$-y_1 - 2y_2 \geq 1$$

is impossible for  $(D)$  since it is inconsistent with

$$y_1, y_2 \geq 0$$

- Since  $(D)$  is infeasible, it follows from Corollary 5 that  $Z \rightarrow \infty$
- You are able to show this result by solving  $(P)$  using the simplex scheme

# OPTIMALITY CRITERION THEOREM

□ We consider the primal-dual problems  $(P)$  and  $(D)$  with

$$\left. \begin{array}{l} \underline{x}^0 \text{ is feasible for } (P) \\ \underline{y}^0 \text{ is feasible for } (D) \\ \underline{c}^T \underline{x}^0 = \underline{b}^T \underline{y}^0 \end{array} \right\} \Rightarrow \begin{array}{l} \underline{x}^0 \text{ is optimal for } (P) \\ \text{and} \\ \underline{y}^0 \text{ is optimal for } (D) \end{array}$$

□ We next provide the proof:

$$\begin{array}{l} \underline{x}^0 \text{ is feasible for } (P) \\ \underline{y}^0 \text{ is feasible for } (D) \end{array} \begin{array}{c} \text{Weak Duality} \\ \Rightarrow \\ \text{Theorem} \end{array} \underline{c}^T \underline{x}^0 \leq \underline{b}^T \underline{y}^0$$

# OPTIMALITY CRITERION THEOREM

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but we are given that

$$\underline{c}^T \underline{x}^0 = \underline{b}^T \underline{y}^0$$

and so it follows that  $\forall$  feasible  $\underline{x}$  with  $\underline{y}^0$  feasible

$$\underline{c}^T \underline{x} \leq \underline{b}^T \underline{y}^0 = \underline{c}^T \underline{x}^0$$

and so  $\underline{x}^0$  is optimal ;

similarly,  $\forall$  feasible  $\underline{y}$  with  $\underline{x}^0$  feasible

$$\underline{b}^T \underline{y} \geq \underline{c}^T \underline{x}^0 = \underline{b}^T \underline{y}^0$$

and so it follows that  $\underline{y}^0$  is optimal

# MAIN DUALITY THEOREM

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**$(P)$  is feasible and  $(D)$  is feasible; then,**

**$\exists \underline{x}^*$  *feasible for  $(P)$  which is optimal and***

**$\exists \underline{y}^*$  *feasible for  $(D)$  which is optimal such that***

$$\underline{c}^T \underline{x}^* = \underline{b}^T \underline{y}^*$$

# COMPLEMENTARY SLACKNESS CONDITIONS

□  $\underline{x}^*$  and  $\underline{y}^*$  are optimal for  $(P)$  and  $(D)$ ,

respectively, if and only if

$$0 = \left( \underline{y}^{*T} \underline{A} - \underline{c}^T \right) \underline{x}^* + \underline{y}^{*T} \left( \underline{b} - \underline{A} \underline{x}^* \right)$$

$$= \underline{y}^{*T} \underline{b} - \underline{c}^T \underline{x}^*$$

□ We prove this equivalence result by defining the slack variables  $\underline{u} \in \mathbb{R}^m$  and  $\underline{v} \in \mathbb{R}^n$  such that  $\underline{x}$  and  $\underline{y}$  are feasible; at the optimum,

$$\underline{A} \underline{x}^* + \underline{u}^* = \underline{b} \quad \underline{x}^*, \underline{u}^* \geq \underline{0}$$

$$\underline{A}^T \underline{y}^* - \underline{v}^* = \underline{c} \quad \underline{y}^*, \underline{v}^* \geq \underline{0}$$

# COMPLEMENTARY SLACKNESS CONDITIONS

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where the optimal values of the slack variables

$\underline{u}^*$  and  $\underline{v}^*$  depend on the optimal values

$\underline{x}^*$  and  $\underline{y}^*$

□ Now,

$$\underline{y}^{*T} \underline{A} \underline{x}^* + \underline{y}^{*T} \underline{u}^* = \underline{y}^{*T} \underline{b} = \underline{b}^T \underline{y}^*$$

$$\underbrace{\underline{x}^{*T} \underline{A}^T \underline{y}^*}_{\underline{y}^{*T} \underline{A} \underline{x}^*} - \underline{x}^{*T} \underline{v}^* = \underline{x}^{*T} \underline{c} = \underline{c}^T \underline{x}^*$$

$$\underline{y}^{*T} \underline{A} \underline{x}^*$$

# COMPLEMENTARY SLACKNESS CONDITIONS

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□ This implies that

$$\underline{y}^{*T} \underline{u}^* + \underline{v}^{*T} \underline{x}^* = \underline{b}^T \underline{y}^* - \underline{c}^T \underline{x}^*$$

□ We need to prove optimality which is true if and  
  
only if

$$\underline{y}^{*T} \underline{u}^* + \underline{v}^{*T} \underline{x}^* = 0$$

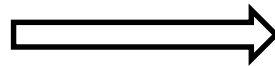


# COMPLEMENTARY SLACKNESS CONDITIONS

□ However,

$\underline{x}^*, \underline{y}^*$  are optimal

*Main*



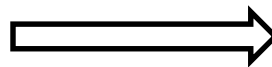
*Duality Theorem*

$$\underline{c}^T \underline{x}^* = \underline{b}^T \underline{y}^* \Rightarrow \underline{y}^{*T} \underline{u}^* + \underline{v}^{*T} \underline{x}^* = 0$$

□ Also,

$$\underline{y}^{*T} \underline{u}^* + \underline{v}^{*T} \underline{x}^* = 0 \Rightarrow \underline{b}^T \underline{y}^* = \underline{c}^T \underline{x}^*$$

*Optimality*



*Criterion Theorem*

$\underline{x}^*$  is optimal for (P) and  $\underline{y}^*$  is optimal for (D)

# COMPLEMENTARY SLACKNESS CONDITIONS

□ Note that

$$\underline{x}^*, \underline{y}^*, \underline{u}^*, \underline{v}^* > 0 \Rightarrow \text{component - wise each element} \geq 0$$

$$\underline{y}^{*T} \underline{u}^* + \underline{v}^* \underline{x}^* = 0 \Rightarrow y_i^* u_i^* = 0 \quad \forall i = 1, \dots, m$$

$$\text{and } v_j^* x_j^* = 0 \quad \forall j = 1, \dots, n$$

□ At the optimum,

$$y_i^* \left( b_i - \sum_{j=1}^n a_{ij} x_j^* \right) = 0 \quad i = 1, \dots, m$$

and

$$x_j^* \left( \sum_{i=1}^m a_{ji} y_i^* - c_j \right) = 0 \quad j = 1, \dots, n$$

# COMPLEMENTARY SLACKNESS CONDITIONS

□ Hence, for  $i = 1, 2, \dots, m$

$$y_i^* > 0 \Rightarrow b_i = \sum_{j=1}^n a_{ij} x_j^*$$

and

$$b_i - \sum_{j=1}^m a_{ij} x_j^* > 0 \Rightarrow y_i^* = 0$$

□ Similarly for  $j = 1, 2, \dots, n$

$$x_j^* > 0 \Rightarrow \sum_{i=1}^m a_{ji} y_i^* = c_j$$

and

$$\sum_{i=1}^m a_{ji} y_i^* - c_j > 0 \Rightarrow x_j^* = 0$$

# EXAMPLE

$$\max \quad Z = x_1 + 2x_2 + 3x_3 + 4x_4$$

*s.t.*

$$x_1 + 2x_2 + 2x_3 + 3x_4 \leq 20$$

$$2x_1 + x_2 + 3x_3 + 2x_4 \leq 20$$

$$x_i \geq 0 \quad i = 1, \dots, 4$$

(P)

# EXAMPLE

*min*

$$W = 20y_1 + 20y_2$$

*s.t.*

$$y_1 + 2y_2 \geq 1$$

$$2y_1 + y_2 \geq 2$$

$$2y_1 + 3y_2 \geq 3$$

$$3y_1 + 2y_2 \geq 4$$

$$y_1, y_2 \geq 0$$

(D)

# EXAMPLE

---

$\underline{x}^*, \underline{y}^*$  *optimal*  $\Rightarrow$

$$y_1^* \left( 20 - x_1^* - 2x_2^* - 2x_3^* - 3x_4^* \right) = 0$$

$$y_2^* \left( 20 - 2x_1^* - x_2^* - 3x_3^* - 2x_4^* \right) = 0$$

$\underline{y}^* = \begin{bmatrix} 1.2 \\ 0.2 \end{bmatrix}$  is given as an optimal solution with

$$\min W = 28$$

# EXAMPLE

$$x_1^* + 2x_2^* + 2x_3^* + 3x_4^* = 20$$

$$2x_1^* + x_2^* + 3x_3^* + 2x_4^* = 20$$

$$y_1^* + 2y_2^* = 1.2 + 0.4 > 1 \Rightarrow x_1^* = 0$$

$$2y_1^* + y_2^* = 2.4 + 0.2 > 2 \Rightarrow x_2^* = 0$$

$$2y_1^* + 3y_2^* = 2.4 + 0.6 = 3$$

$$3y_1^* + 2y_2^* = 3.6 + 0.4 = 4$$

so that

$$\left. \begin{aligned} 2x_3^* + 3x_4^* &= 20 \\ 3x_3^* + 2x_4^* &= 20 \end{aligned} \right\} \Rightarrow \begin{aligned} x_3^* &= 4 \\ x_4^* &= 4 \end{aligned}$$

# COMPLEMENTARY SLACKNESS CONDITION APPLICATIONS

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- ❑ Key uses of *c.s. conditions* are
  - finding optimal ( $P$ ) solution given optimal ( $D$ ) solution and vice versa
  - verification of optimality of solution (whether a feasible solution is optimal)
- ❑ We can start with a feasible solution and attempt to construct an optimal dual solution; if we succeed, then the feasible primal solution is *optimal*



# DUALITY

$$\begin{array}{ll} \max & Z = \underline{c}^T \underline{x} \\ & \text{s.t.} \\ & \underline{A} \underline{x} \leq \underline{b} \\ & \underline{x} \geq \underline{0} \\ \min & W = \underline{b}^T \underline{y} \\ & \text{s.t.} \\ & \underline{A}^T \underline{y} \geq \underline{c} \\ & \underline{y} \geq \underline{0} \end{array} \quad \left. \begin{array}{l} \text{ } \\ \text{ } \\ \text{ } \\ \text{ } \\ \text{ } \\ \text{ } \end{array} \right\} \begin{array}{l} (P) \\ (D) \end{array}$$

# DUALITY

□ Suppose the primal problem is minimization, then,

$$\min \quad Z = \underline{c}^T \underline{x} \quad (P)$$

s.t.

$$\underline{A} \underline{x} \geq \underline{b}$$

$$\underline{x} \geq \underline{0}$$

$$\max \quad W = \underline{b}^T \underline{y} \quad (D)$$

s.t.

$$\underline{A}^T \underline{y} \leq \underline{c}$$

$$\underline{y} \geq \underline{0}$$

# INTERPRETATION

- The economic interpretation is

$$Z^* = \max Z = \underline{c}^T \underline{x}^* = \underline{b}^T \underline{y}^* = W^* = \min W$$

$$\left. \begin{array}{l} b_i - \text{constrained resource quantities,} \\ y_i^* - \text{optimal dual variables} \end{array} \right\} i = 1, 2, \dots, m$$

- Suppose, we change

$$b_i \rightarrow b_i + \Delta b_i \Rightarrow \Delta Z = y_i^* \Delta b_i$$

- In words, the optimal dual variable for each primal constraint gives the net change in the optimal value of the objective function  $Z$  for a one unit change in the constraint on resources

# INTERPRETATION

---

- ❑ Economists refer to the dual variable as the *shadow price* on the constraint resource
- ❑ The *shadow price* determines the value/worth of having an additional quantity of a resource
- ❑ In the previous example, the optimal dual variables indicate that the worth of another unit of resource 1 is 1.2 while that of another unit of resource 2 is 0.2

# GENERALIZED FORM OF THE DUAL

□ We start out with

$$\begin{array}{ll} \max & \underline{Z} = \underline{c}^T \underline{x} \\ & \\ s.t. & \\ & \underline{A} \underline{x} = \underline{b} \\ & \underline{x} \geq \underline{0} \end{array} \quad \left. \vphantom{\begin{array}{l} \max \\ s.t. \end{array}} \right\} (P)$$

# GENERALIZED FORM OF THE DUAL

□ To find  $(D)$ , we first put  $(P)$  in symmetric form

$$\begin{array}{ll} \underline{y}_+ \leftrightarrow \underline{A} \underline{x} \leq \underline{b} \\ \underline{y}_- \leftrightarrow -\underline{A} \underline{x} \leq -\underline{b} \\ \underline{x} \geq \underline{0} \end{array} \quad \begin{bmatrix} \underline{A} \\ -\underline{A} \end{bmatrix} \underline{x} \leq \begin{bmatrix} \underline{b} \\ -\underline{b} \end{bmatrix} \quad \begin{array}{l} \textit{symmetric} \\ \textit{form} \end{array}$$

# GENERALIZED FORM OF THE DUAL

---

□ Let

$$\underline{y} = \underline{y}_+ - \underline{y}_-$$

□ We rewrite the problem as

$$\min W = \underline{b}^T \underline{y}$$

*s.t.*

$$\underline{A}^T \underline{y} \geq \underline{c}$$

*$\underline{y}$  is unsigned*

□ The *c.s.* conditions apply

$$\underline{x}^{*T} (\underline{A}^T \underline{y}^* - \underline{c}) = \underline{0}$$

# EXAMPLE 5: THE PRIMAL

$$\max Z = x_1 - x_2 + x_3 - x_4$$

*s.t.*

$$y_1 \Leftrightarrow x_1 + x_2 + x_3 + x_4 = 8$$

$$y_2 \Leftrightarrow x_1 \leq 8$$

$$y_3 \Leftrightarrow x_2 \leq 4$$

$$y_4 \Leftrightarrow -x_2 \leq 4 \quad (P)$$

$$y_5 \Leftrightarrow x_3 \leq 4$$

$$y_6 \Leftrightarrow -x_3 \leq 2$$

$$y_7 \Leftrightarrow x_4 \leq 10$$

$$x_1, x_4 \geq 0$$

$x_2, x_3$  *unsigned*



# EXAMPLE 5: THE DUAL

$$\min W = 8y_1 + 8y_2 + 4y_3 + 4y_4 + 4y_5 + 2y_6 + 10y_7\pi$$

*s.t.*

$$x_1 \leftrightarrow y_1 + y_2 \geq 1$$

$$x_2 \leftrightarrow y_1 + y_3 - y_4 = -1 \quad (D)$$

$$x_3 \leftrightarrow y_1 + y_5 - y_6 = 1$$

$$x_4 \leftrightarrow \phantom{y_1 + y_2 + y_3 - y_4 + y_5 - y_6} + y_7 \geq 1$$

$$y_2, \dots, y_7 \geq 0$$

*y<sub>1</sub> unsigned*

# EXAMPLE 5: *c.s. conditions*

---

□ We are given that

$$\underline{x}^* = \begin{bmatrix} 8 \\ -4 \\ 4 \\ 0 \end{bmatrix}$$

is optimal for  $(P)$

□ Then the *c.s. conditions* obtain

$$x_1^* (y_1^* + y_2^* - 1) = 0$$

## EXAMPLE 5: *c.s. conditions*

---

so that

$$x_1^* = 8 > 0 \Rightarrow y_1^* + y_2^* = 1$$

□ The other *c.s. conditions* obtain

$$y_i^* \left( \sum_{j=1}^4 a_{ij} x_j^* - b_i \right) = 0$$

□ Now,  $x_4^* = 0$  implies  $x_4^* - 10 < 0$  and so

$$y_7^* = 0$$

# EXAMPLE 5: *c.s. conditions*

---

□ Also,  $x_3^* = 4$  implies

$$y_6^* = 0$$

□ Similarly, the *c.s. conditions*

$$x_j^* \left( \sum_{i=1}^7 a_{ji} y_i^* - c_j \right) = 0$$

have implications on the  $y_i^*$  variable

# EXAMPLE 5: *c.s. conditions*

---

□ Since  $x_2^* = -4$ , then we have

$$y_3^* = 0$$

□ Now, with  $y_7^* = 0$  we have

$$y_1^* > -1$$

□ Since,  $W = \underline{b}^T \underline{y}$  we have

$$y_2^* = 1 - y_1^*$$

# EXAMPLE 5

---

□ Suppose

$$y_1^* = 1$$

and so,

$$y_2^* = 0$$

□ Furthermore,

$$y_1^* + y_3^* - y_4^* = 1 - y_4^* = -1$$

implies

$$y_4^* = 2$$

# EXAMPLE 5

---

□ Also

$$y_1^* + y_5^* - y_6^* = 1$$

implies

$$1 + y_5^* = 1$$

and so

$$y_5^* = 0$$

# EXAMPLE 5

---

□ Therefore

$$\begin{aligned} W(\underline{y}^*) &= (8)(1) + (8)(0) + (4)(0) + (4)(2) + \\ &\quad (4)(0) + (2)(0) + (10)(0) \\ &= 16 \end{aligned}$$

and so

$$W^* = 16 = Z^* \Leftrightarrow \text{optimality of } (P) \text{ and } (D)$$



# *PRIMAL – DUAL* TABLE

<i>primal</i> (maximize)	<i>dual</i> (minimize)
$\underline{A}$ ( coefficient matrix )	$\underline{A}^T$ ( transpose of the coefficient matrix )
$\underline{b}$ ( right-hand side vector )	$\underline{b}$ ( cost vector )
$\underline{c}$ ( price vector )	$\underline{c}$ ( right hand side vector )
$i^{th}$ constraint is = type	the dual variable $y_i$ is unrestricted in sign
$i^{th}$ constraint is $\leq$ type	the dual variable $y_i \geq 0$
$i^{th}$ constraint is $\geq$ type	the dual variable $y_i \leq 0$
$x_j$ is unrestricted	$j^{th}$ dual constraint is = type
$x_j \geq 0$	$j^{th}$ dual constraint is $\geq$ type
$x_j \leq 0$	$j^{th}$ dual constraint is $\leq$ type