## ECE 307- Techniques for Engineering Decisions

## Lecture 4. Duality Concepts in Linear Programming

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## DUALITY

$\square$ 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:


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## DUALITY DEFINITIONS

## $\square$ We first define the primal and dual problems

$$
\begin{align*}
& \text { max } \\
& Z=\underline{c}^{\boldsymbol{T}} \underline{x} \\
& \text { s.t. } \\
& \underline{\boldsymbol{A}} \underline{x} \leq \underline{\boldsymbol{b}} \\
& \underline{x} \geq \underline{0} \\
& W=\underline{b}^{T} \underline{y} \\
& \text { s.t. } \\
& \underline{\boldsymbol{A}}^{T} \underline{y} \geq \underline{\boldsymbol{c}}  \tag{D}\\
& \underline{y} \geq \underline{0}
\end{align*}
$$

## DUALITY DEFINITIONS

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

$$
\max Z=c_{1} x_{1}+c_{2} x_{2}+\ldots+c_{n} x_{n}
$$

s.t.

$$
\left.\begin{array}{c}
a_{11} x_{1}+a_{12} x_{2}+\ldots+a_{1 n} x_{n} \leq b_{1} \\
a_{21} x_{1}+a_{22} x_{2}+\ldots+a_{2 n} x_{n} \leq b_{2} \\
\vdots \\
a_{m 1} x_{1}+a_{m 2} x_{2}+\ldots+a_{m n} x_{n} \leq b_{m} \\
x_{1} \geq 0, \quad x_{2} \geq 0, \ldots, \quad x_{n} \geq 0
\end{array}\right\}(P)
$$

## DUALITY DEFINITIONS

$$
\min W=b_{1} y_{1}+b_{2} y_{2}+\ldots+b_{m} y_{m}
$$

sot.

$$
\left.\begin{array}{c}
a_{11} y_{1}+a_{21} y_{2}+\ldots+a_{m 1} y_{m} \geq c_{1} \\
a_{12} y_{1}+a_{22} y_{2}+\ldots+a_{m 2} y_{m} \geq c_{2} \\
\vdots \\
a_{1 n} y_{1}+a_{2 n} y_{2}+\ldots+a_{m n} y_{m} \geq c_{n} \\
y_{1} \geq 0, \quad y_{2} \geq 0, \ldots, \quad y_{m} \geq 0
\end{array}\right\}(D)
$$

## EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM



## EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM

$\square$ We are given that the supplies stored in warehouses
$W_{1}$ and $W_{2}$ satisfy

$$
\begin{aligned}
& \text { supply at } W_{1} \leq 300 \\
& \text { supply at } W_{2} \leq 600
\end{aligned}
$$

$\square$ We are also given the demands needed to be met at the retail stores $\boldsymbol{R}_{1}, \boldsymbol{R}_{2}$, and $\boldsymbol{R}_{3}$ :
demand at $R_{1} \geq 200$
demand at $R_{2} \geq 300$
demand at $R_{3} \geq 400$

## EXAMPLE 1: MANUFACTURER TRANSPORTATION PROBLEM

$\square$ The problem is to determine the least-cost shipping

## schedule

$\square$ We define the decision variable
$x_{i j}=$ quantity shipped from $W_{i}$ to $R_{j} i=1,2, j=1,2,3$
$\square$ The shipping costs may be viewed as
$c_{i j}=$ element $i, j$ of the transportation cost matrix

## FORMULATION STATEMENT

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

s.t.

$$
\begin{aligned}
& x_{11}+x_{12}+x_{13} \leq 300 \\
& x_{21}+x_{22}+x_{23} \leq 600 \\
& x_{11} \quad+x_{21} \quad \geq 200 \\
& \boldsymbol{x}_{12} \quad+\boldsymbol{x}_{22} \quad \geq \mathbf{3 0 0} \\
& \boldsymbol{x}_{13} \\
& +x_{23} \geq 400 \\
& x_{i j} \geq 0 \quad i=1,2, j=1,2,3
\end{aligned}
$$

## DUAL PROBLEM SETUP USING SYMMETRIC FORM

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

s.t.

$$
\begin{aligned}
& y_{1} \leftrightarrow-x_{11}-x_{12}-x_{13} \geq-300 \\
& y_{2} \leftrightarrow \\
& y_{3} \leftrightarrow \quad \boldsymbol{x}_{11} \\
& y_{4} \leftrightarrow \\
& x_{12} \\
& -x_{21}-x_{22}-x_{23} \geq-600 \\
& +x_{21} \\
& \geq \quad 200 \\
& y_{5} \leftrightarrow \\
& x_{13} \\
& +x_{23} \geq 400 \\
& x_{i j} \geq 0 \quad i=1,2 \quad j=1,2,3
\end{aligned}
$$

## DUAL PROBLEM SETUP

$\max W=-300 y_{1}-600 y_{2}+200 y_{3}+300 y_{4}+400 y_{5}$ sot.

$$
\begin{aligned}
& -y_{1} \quad+y_{3} \\
& -y_{1}+y_{4} \\
& \leq c_{11}=2 \\
& -y_{1} \\
& -y_{2}+y_{3} \quad \leq c_{21}=5 \\
& -y_{2}+y_{4} \\
& +y_{5} \leq c_{13}=3 \\
& \leq c_{12}=4 \\
& -y_{2}+y_{5} \leq c_{23}=4 \\
& y_{i} \geq 0 \quad i=1,2, \ldots, 5
\end{aligned}
$$

## 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 $\square$ 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

$$
\begin{aligned}
& -y_{1} \quad+y_{3} \\
& -y_{1} \\
& -y_{1} \\
& -y_{2}+y_{3} \\
& -y_{2} \quad+y_{4} \\
& -y_{2} \\
& +y_{4} \\
& \leq c_{11}=2 \\
& +y_{5} \leq c_{23}=4
\end{aligned}
$$

$\square$ 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

## $\square$ Resource requirements

| item | sales price (\$) |
| :---: | :---: |
| desks | 60 |
| tables | 30 |
| chairs | 20 |



## EXAMPLE 2: FURNITURE PRODUCTS

$\square$ The Dakota Furniture Company manufacturing:

| resource | desk | table | chair | available |
| :---: | :---: | :---: | :---: | :---: |
| lumber board (ft) | 8 | 6 | 1 | 48 |
| finishing (h) | 4 | 2 | 1.5 | 20 |
| carpentry $(h)$ | 2 | 1.5 | 0.5 | 8 |

$\square$ We assume that the demand for desks, tables and chairs is unlimited and the two required resources - lumber and labor - are already purchased

The decision problem is to maximize total revenues

## PRIMAL AND DUAL PROBLEM FORMULATION

$\square$ We define decision variables

$$
\begin{aligned}
& x_{1}=\text { number of desks produced } \\
& x_{2}=\text { number of tables produced } \\
& x_{3}=\text { number of chairs produced }
\end{aligned}
$$

$\square$ The Dakota problem is
$\max Z=60 x_{1}+30 x_{2}+20 x_{3}$
s.t.

\[

\]

## PRIMAL AND DUAL PROBLEM FORMULATION

## $\square$ The dual problem is

$$
\begin{aligned}
& \min W=48 y_{1}+20 y_{2}+8 y_{3} \\
& \text { s.t. } \\
& 8 y_{1}+4 y_{2}+2 y_{3} \geq 60 \\
& 6 y_{1}+2 y_{2}+1.5 y_{3} \text { desk } \\
& y_{1}+1.5 y_{2}+0.5 y_{3} \geq 20 \\
& \text { chable } \\
& y_{1}, y_{2}, y_{3} \geq 0
\end{aligned}
$$

## PRIMAL AND DUAL PROBLEM FORMULATION

$\max \quad Z=60 x_{1}+30 x_{2}+20 x_{3}$

$$
\begin{aligned}
y_{1} & \leftrightarrow 8 x_{1}+6 x_{2}+\quad x_{3} \leq 48 & \text { lumber } \\
y_{2} & \leftrightarrow 4 x_{1}+2 x_{2}+1.5 x_{3} \leq 20 & \text { finishing } \\
y_{3} & \leftrightarrow 2 x_{1}+1.5 x_{2}+0.5 x_{3} \leq 8 & \text { carpentry } \\
x_{1}, x_{2}, x_{3} & \geq 0 &
\end{aligned}
$$

$\max \quad W=48 y_{1}+20 y_{2}+8 y_{3}$

$$
\begin{aligned}
48 y_{1}+20 y_{2}+8 y_{3} & \geq 60 & \text { desk } \\
6 y_{1}+2 y_{2}+1.5 y_{3} & \geq 30 & \text { table } \\
y_{1}+1.5 y_{2}+0.5 y_{3} & \geq 20 & \text { chair } \\
y_{1}, y_{2}, y_{3} & \geq 0 &
\end{aligned}
$$

## INTERPRETATION OF THE DUAL PROBLEM

## $\square$ An entrepreneur wishes to purchase all of

 Dakota's resources$\square$ He needs, therefore, to determine the prices to pay for each unit of each resource

$$
\begin{aligned}
& y_{1}=\text { price paid per ft of lumber board } \\
& y_{2}=\text { price paid per } h \text { of finishing labor } \\
& y_{3}=\text { price paid per } h \text { of carpentry labor }
\end{aligned}
$$

$\square$ We solve the Dakota dual problem to determine $y_{1}, y_{2}$ and $y_{3}$

## INTERPRETATION OF THE DUAL PROBLEM

$\square$ To convince Dakota to sell the raw resources, the resource prices must be set sufficiently high
$\square$ For example, the entrepreneur must offer Dakota at least $\$ \mathbf{6 0}$ for a combination of resources that consists of 8 ft of lumber board, $4 \boldsymbol{h}$ of finishing and $2 h$ of carpentry, since Dakota could use this combination to sell a desk for $\$ \mathbf{6 0}$; this requires es the following dual constraint:

$$
8 y_{1}+4 y_{2}+2 y_{3} \geq 60
$$

# INTERPRETATION OF DUAL PROBLEM 

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

## EXAMPLE 3: DIET PROBLEM

## A new diet requires that all food eaten come from

one of the four "basic food groups":
O chocolate cake $\bigcirc$ soda

O ice cream
O cheesecake
$\square$ The four foods available for consumption are
O brownie
O cola

O chocolate ice cream $O$ pineapple cheesecake

## EXAMPLE 3: DIET PROBLEM

$\square$ Minimum requirements for each day are:

O 500 cal

O 6 oz chocolate

O $10 \boldsymbol{o z}$ sugar

O $8 \boldsymbol{o z}$ fat
$\square$ The objective is to minimize the diet costs

## EXAMPLE 3: DIET PROBLEM

| food | calories | $\begin{gathered} \text { chocolate } \\ (o z) \end{gathered}$ | sugar (oz) | fat (oz) | $\begin{gathered} c o s t s \\ (\text { cents) } \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| brownie | 400 | 3 | 2 | 2 | 50 |
| chocolate ice cream (scoop) | 200 | 2 | 2 | 4 | 20 |
| $\underset{\text { (bottle) }}{\substack{\text { cola } \\ \hline}}$ | 150 | 0 | 4 | 1 | 30 |
| pineapple cheesecake (piece) | 500 | 0 | 4 | 5 | 80 |

## PROBLEM FORMULATION

$\square$ Objective of the problem is to minimize the total costs of the diet

Decision variables are defined for each day's purchases

$$
\begin{aligned}
& x_{1}=\text { number of brownies } \\
& x_{2}=\text { number of chocolate ice cream scoops } \\
& x_{3}=\text { number of bottles of soda } \\
& x_{4}=\text { number of pineapple cheesecake pieces }
\end{aligned}
$$

## PROBLEM FORMULATION

## $\square$ The problem statement is

$\min Z=50 x_{1}+20 x_{2}+30 x_{3}+80 x_{4}$
s.t.
$400 x_{1}+200 x_{2}+150 x_{3}+500 x_{4} \geq 500$ cal

$$
\begin{array}{ll}
3 x_{1}+2 x_{2} & \\
2 x_{1}+2 x_{2}+4 x_{3}+4 x_{4} & \geq 10 \mathrm{oz} \\
2 x_{1}+4 x_{2}+x_{3}+5 x_{4} & \geq 8 \mathrm{oz}
\end{array}
$$

$$
x_{i} \geq 0 \quad i=1,4
$$

## EXAMPLE 3: DIET PROBLEM

## $\square$ The dual problem is

$\max W=500 y_{1}+6 y_{2}+10 y_{3}+8 y_{4}$
sot.
$400 y_{1}+3 y_{2}+2 y_{3}+2 y_{4} \leq 50 \quad$ brownie
$200 y_{1}+2 y_{2}+2 y_{3}+4 y_{4} \leq 20 \quad$ ice cream
$150 y_{1} \quad+4 y_{3}+y_{4} \leq 30 \quad$ soda
$500 y_{1} \quad+4 y_{3}+5 y_{4} \leq 80 \quad$ cheesecake

$$
y_{1}, y_{2}, y_{3}, y_{4} \geq 0
$$

## INTERPRETATION OF THE DUAL

$\square$ 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"
$\square$ The decision is to determine the prices charged $y_{i}=$ price per unit of required nutrient to sell to dieters
$\square$ 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 $50 \phi$ and have $400 \mathrm{cal}, 3 \mathrm{oz}$ of chocolate, 2 oz of sugar and 2 oz of fat
$\square$ The sales price $y_{i}$ must be set sufficiently low to entice the buyer to get the required nutrients from the brownie:

$$
400 y_{1}+3 y_{2}+2 y_{3}+2 y_{4} \leq 50 \longleftarrow \underset{\text { constraint }}{\text { brownie }}
$$

$\square$ We derive similar constraints for the ice cream, the soda and the cheesecake

## DUAL PROBLEMS

$\max \quad Z=\underline{\boldsymbol{c}}^{T} \underline{\boldsymbol{x}}$
sot.

$$
\begin{aligned}
\underline{A} \underline{x} & \leq \underline{b} \\
\underline{x} & \geq \underline{0}
\end{aligned}
$$

min

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

sot.

$$
\begin{aligned}
\underline{A}^{T} \underline{y} & \geq \underline{c} \\
\underline{y} & \geq \underline{0}
\end{aligned}
$$

## WEAK DUALITY THEOREM

$\square$ For any $\underline{x}$ feasible for $(P)$ and any $\underline{y}$ feasible for $(D)$, the following relation is satisfied

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

$\square$ Proof:

$$
\begin{gathered}
\underline{A}^{T} \underline{y} \geq \underline{c} \Rightarrow \underline{c}^{T} \leq \underline{y}^{T} \underline{A} \Rightarrow \underline{c}^{T} \underline{x} \leq \underline{y}^{T} \underline{A} \underline{x} \\
\underline{c}^{T} \underline{x} \leq \underline{y}^{T} \underline{A} \underline{x} \leq \underline{y}^{T} \underline{b}=\underline{b}^{T} \underline{y}
\end{gathered}
$$

## COROLLARY 1 OF THE WEAK DUALITY THEOREM

$\underline{x}$ 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

$\underline{y}$ 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 $(D)$,

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

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## COROLLARIES 3 AND 4 OF THE WEAK DUALITY THEOREM

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

## $\square$ Consider the maximization problem

$$
\max Z=x_{1}+2 x_{2}+3 x_{3}+4 x_{4}=\underbrace{[1,2,3,4]}_{\underline{c}^{T}} \underline{x}
$$

s.t.

$$
\underbrace{\left[\begin{array}{llll}
1 & 2 & 2 & 3 \\
2 & 1 & 3 & 2
\end{array}\right]}_{\underline{A}} \underline{\underline{x}} \leq \underbrace{\left[\begin{array}{c}
20 \\
20
\end{array}\right]}_{\underline{b}}
$$

$(P)$

## DUALITY THEOREM APPLICATION

The corresponding dual is given by

$$
\left.\begin{array}{ll}
\min & \\
\text { s.t. } & \\
&  \tag{D}\\
& \underline{A}^{T} \underline{y} \\
& \underline{y} \geq \underline{c} \\
& \underline{y} \geq \underline{0}
\end{array}\right\}
$$

$\square$ With the appropriate substitutions, we obtain

## DUALITY THEOREM APPLICATION

min

$$
W=20 y_{1}+20 y_{2}
$$

sot.

$$
\begin{aligned}
& y_{1}+2 y_{2} \geq 1 \\
& 2 y_{1}+y_{2} \geq 2 \\
& 2 y_{1}+3 y_{2} \geq 3 \\
& 3 y_{1}+2 y_{2} \geq 4 \\
& y_{1} \geq 0, y_{2} \geq 0
\end{aligned}
$$

## GENERALIZED FORM OF THE DUAL

$\square$ Consider the primal decision

$$
x_{i}=1, \quad i=1,2,3,4 ;
$$

decision is feasible for $(P)$ with

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

$\square$ 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

## $\square$ Clearly,

$$
Z\left(x_{1}, x_{2}, x_{3}, x_{4}\right)=10 \leq 40=W\left(y_{1}, y_{2}\right)
$$

and so clearly, the feasible decision for $(P)$ and (D)
satisfy the Weak Duality Theorem
$\square$ Moreover, we have
corollary $1 \Rightarrow 10 \leq \min W=W\left(y_{1}^{*}, y_{2}^{*}\right)$
corollary $2 \Rightarrow \max Z=Z\left(x_{1}^{*}, x_{2}^{*}, x_{3}^{*}, x_{4}^{*}\right) \leq \underline{b}^{T} \underline{y}=40$

## COROLLARIES 5 AND 6

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

## EXAMPLE

## $\square$ Consider the primal dual problems:

$\square$ Now

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

## EXAMPLE

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

but

$$
-y_{1}-2 y_{2} \geq 1
$$

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

$$
y_{1}, y_{2} \geq 0
$$

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

## OPTIMALITY CRITERION THEOREM

$\square$ We consider the primal-dual problems $(P)$ and (D) with
$\underline{x}^{0}$ is feasible for $(P) \quad \underline{x}^{0}$ is optimal for $(P)$ $\underline{y}^{0}$ is feasible for $(D)$
$\underline{c}^{T} \underline{x}^{0}=\underline{b}^{T} \underline{y}^{0}$ $\Rightarrow$ and
$\underline{y}^{0}$ is optimal for (D)
We next provide the proof:
$\underline{x}^{0}$ is feasible for (P) Weak Duality $\underline{y}^{0}$ is feasible for $(D) \underset{\text { the rem }}{\Rightarrow} \underline{c}^{T} \underline{x}^{0} \leq \underline{b}^{T} \underline{y}^{0}$

## OPTIMALITY CRITERION THEOREM

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

## $(P)$ 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

$\square \underline{x}^{*}$ and $\underline{y}$ * are optimal for $(P)$ and (D) respectively, if and only if

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

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

$$
\begin{array}{ll}
\underline{\boldsymbol{A}}^{*}+\underline{\boldsymbol{u}}^{*}=\underline{\boldsymbol{b}} \quad \underline{\boldsymbol{x}}^{*}, \underline{\boldsymbol{u}}^{*} \geq \underline{\boldsymbol{0}} \\
\boldsymbol{A}^{T} \underline{\boldsymbol{y}}^{*}-\underline{\boldsymbol{v}}^{*}=\underline{\boldsymbol{c}} \quad \underline{\boldsymbol{y}}^{*}, \underline{\boldsymbol{v}}^{*} \geq \mathbf{0}
\end{array}
$$

## COMPLEMENTARY SLACKNESS CONDITIONS

where the optimal values of the slack variables
$\underline{u}^{*}$ and $\underline{v}$ * depend on the optimal values
$\underline{x}^{*}$ and $\underline{y}^{*}$
$\square$ Now,

$$
\begin{aligned}
& \underline{\boldsymbol{y}}^{* T} \underline{\boldsymbol{A}}^{*}+\underline{y}^{* T} \underline{u}^{*}=\underline{y}^{* T} \underline{\boldsymbol{b}}=\underline{b}^{T} \underline{\boldsymbol{y}}^{*} \\
& \underbrace{\boldsymbol{x}^{* T}} \underline{\boldsymbol{A}}^{T} \underline{\underline{y}}^{*}-\underline{x}^{* T} \underline{\boldsymbol{v}}^{*}=\underline{x}^{* T} \underline{\boldsymbol{c}}=\underline{\boldsymbol{c}}^{T} \underline{x}^{*}
\end{aligned} \underline{\underline{x}}^{*} .
$$

## COMPLEMENTARY SLACKNESS CONDITIONS

$\square$ This implies that

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

$\square$ 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

## $\square$ However,

$$
\begin{gathered}
\underline{x}^{*}, \underline{y}^{*} \underset{\substack{\text { are optimal } \\
\underline{c}^{T} \\
\underline{x}^{*}=\\
\underline{b}^{T} \\
\underline{y}^{*} \\
\Rightarrow \underline{y}^{* T} \underline{u}^{*}+\underline{v}^{* T} \underline{x}^{*}=0}}{\text { Theorem }}=0
\end{gathered}
$$

$\square$ Also,

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

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

## COMPLEMENTARY SLACKNESS CONDITIONS

## - Note that

$$
\begin{gathered}
\underline{x}^{*}, \underline{y}^{*}, \underline{u}^{*}, \underline{v}^{*}>0 \Rightarrow \text { component }- \text { wise each element } \geq 0 \\
\underline{y}^{* T} \underline{u}^{*}+\underline{v}^{*} \underline{x}^{*}=0 \Rightarrow y_{i}^{*} u *=0 \forall i=1, \ldots, m \\
\text { and } v_{j}^{*} x_{j}^{*}=0 \forall j=1, \ldots, n
\end{gathered}
$$

$\square$ At the optimum,

$$
y_{i}^{*}\left(b_{i}-\sum_{j=1}^{n} a_{i j} x_{j}^{*}\right)=0 \quad i=1, \ldots, m
$$

and

$$
x_{j}^{*}\left(\sum_{i=1}^{m} a_{j i} y_{i}^{*}-c_{j}\right)=0 \quad j=1, \ldots, n
$$

## COMPLEMENTARY SLACKNESS CONDITIONS

$\square$ Hence, for $i=1,2, \ldots, m$

$$
y_{i}^{*}>0 \Rightarrow b_{i}=\sum_{j=1}^{n} a_{i j} x_{j}^{*}
$$

and

$$
b_{i}-\sum_{j=1}^{m} a_{i j} x_{i}^{*}>0 \Rightarrow y_{i}^{*}=0
$$

$\square$ Similarly for $j=1,2, \ldots, n$

$$
x_{i}^{*}>0 \Rightarrow \sum_{i=1}^{m} a_{j i} y_{i}^{*}=c_{j}
$$

and

$$
\sum_{i=1}^{m} a_{j i} y_{i}^{*}-c_{j}>0 \Rightarrow x_{j}^{*}=0
$$

## EXAMPLE

$\max \quad Z=x_{1}+2 x_{2}+3 x_{3}+4 x_{4}$
s.t.

$$
\left.\begin{array}{r}
x_{1}+2 x_{2}+2 x_{3}+3 x_{4} \leq 20 \\
2 x_{1}+x_{2}+3 x_{3}+2 x_{4} \leq 20 \\
x_{i} \geq 0 \quad(P) \\
i=1, \ldots, 4
\end{array}\right\}
$$

## EXAMPLE

$\min \quad W=20 y_{1}+20 y_{2}$
s.t.

$$
\left.\begin{array}{rl}
y_{1}+2 y_{2} & \geq 1 \\
2 y_{1}+y_{2} & \geq 2 \\
2 y_{1}+3 y_{2} & \geq 3 \\
3 y_{1}+2 y_{2} & \geq 4 \\
y_{1}, y_{2} & \geq 0
\end{array}\right\}(D)
$$

## EXAMPLE

$$
\begin{aligned}
& \underline{x}^{*}, \underline{y}^{*} \text { optimal } \Rightarrow \\
& y_{1}^{*}\left(20-x_{1}^{*}-2 x_{2}^{*}-2 x_{3}^{*}-3 x_{4}^{*}\right)=0 \\
& y_{2}^{*}\left(20-2 x_{1}^{*}-x_{2}^{*}-3 x_{3}^{*}-2 x_{4}^{*}\right)=0 \\
& \underline{y}^{*}=\left[\begin{array}{l}
1.2 \\
0.2
\end{array}\right] \text { is given as an optimal solution with }
\end{aligned}
$$

$$
\min W=28
$$

## EXAMPLE

$$
\begin{aligned}
& \boldsymbol{x}_{1}^{*}+2 \boldsymbol{x}_{2}^{*}+2 \boldsymbol{x}_{3}^{*}+\mathbf{3} \boldsymbol{x}_{4}^{*}=\mathbf{2 0} \\
& \mathbf{2} \boldsymbol{x}_{1}^{*}+\boldsymbol{x}_{2}^{*}+\mathbf{3} \boldsymbol{x}_{3}^{*}+\mathbf{2} \boldsymbol{x}_{4}^{*}=\mathbf{2 0}
\end{aligned}
$$

$y_{1}^{*}+2 y_{2}^{*}=1.2+0.4>1 \Rightarrow x_{1}^{*}=0$
$2 y_{1}^{*}+y_{2}^{*}=2.4+0.2>2 \Rightarrow x_{2}^{*}=0$
$2 y_{1}^{*}+3 y_{2}^{*}=2.4+0.6=3$
$3 y_{1}^{*}+2 y_{2}^{*}=3.6+0.4=4$ so that

$$
\left.\begin{array}{ll}
2 x_{3}^{*}+3 x_{4}^{*}=20 & \Rightarrow x_{3}^{*}=4 \\
3 x_{3}^{*}+2 x_{4}^{*}=20 & \Rightarrow x^{*}=4
\end{array}\right\}
$$

## COMPLEMENTARY SLACKNESS CONDITION APPLICATIONS

$\square$ Key uses of c.s. conditions are
O finding optimal ( $P$ ) solution given optimal (D) solution and vice versa

O verification of optimality of solution (whether a feasible solution is optimal)
$\square$ 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

[^0]
## DUALITY

## $\max$ <br> $$
z=\underline{\underline{c}}^{T} \underline{x}
$$

sot.

$$
\begin{aligned}
\underline{A} \underline{x} & \leq \underline{b} \\
\underline{x} & \geq \underline{0}
\end{aligned}
$$

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

$$
\begin{aligned}
\underline{A}^{T} \underline{y} & \geq \underline{c} \\
\underline{y} & \geq \underline{0}
\end{aligned}
$$

## DUALITY

## Suppose the primal problem is minimization, then,

$\min \quad Z=\underline{c}^{T} \underline{x}$
s.t.

$$
\begin{aligned}
\underline{\boldsymbol{A}} \underline{\boldsymbol{x}} & \geq \underline{\boldsymbol{b}} \\
\underline{\boldsymbol{x}} & \geq \underline{\boldsymbol{0}} \\
\boldsymbol{W} & =\underline{\boldsymbol{b}}^{T} \underline{\boldsymbol{y}}
\end{aligned}
$$

max
s.t.

$$
\begin{align*}
\underline{A}^{T} \underline{y} & \leq \underline{c}  \tag{D}\\
\underline{y} & \geq \underline{0}
\end{align*}
$$

## INTERPRETATION

$\square$ The economic interpretation is
$Z^{*}=\max Z=\underline{c}^{T} \underline{x}^{*}=\underline{b}^{T} \underline{y}^{*}=W^{*}=\min W$
$b_{i}-$ constrained resource quantities
$y_{i}^{*}-$ optimal dual variables
$\square$ Suppose, we change

$$
b_{i} \rightarrow b_{i}+\Delta b_{i} \Rightarrow \Delta Z=y_{i}^{*} \Delta b_{i}
$$

$\square$ 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

$\square$ 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
$\square$ 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



## GENERALIZED FORM OF THE DUAL

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

$$
\begin{aligned}
\underline{y}_{+} \leftrightarrow & \underline{A} \underline{x}
\end{aligned} \leq \underline{b}\left[\begin{array}{r}
\underline{A} \\
\underline{y}_{-} \leftrightarrow-\underline{A} \underline{x}
\end{array}\right]-\underline{b}\left[\underline{x} \leq\left[\begin{array}{r}
\underline{b} \\
-\underline{A}
\end{array}\right] \quad \underline{-\underline{b}}\right]
$$

## GENERALIZED FORM OF THE DUAL

- Let

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

- We rewrite the problem as

$$
\begin{aligned}
& \min W=\underline{b}^{T} \underline{y} \\
& \text { s.t. } \\
& \qquad \underline{A}^{T} \underline{y} \geq \underline{c} \\
& \underline{y} \text { is unsigned }
\end{aligned}
$$

$\square$ The c.s. conditions apply

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

## EXAMPLE 5: THE PRIMAL

s.t.

$$
\max Z=x_{1}-x_{2}+x_{3}-x_{4}
$$

$$
\begin{aligned}
& y_{1} \leftrightarrow x_{1}+x_{2}+x_{3}+x_{4}=8 \\
& y_{2} \leftrightarrow x_{1} \quad \leq 8 \\
& y_{3} \leftrightarrow \quad x_{2} \quad \leq 4 \\
& y_{4} \leftrightarrow \quad-x_{2} \quad \leq 4 \\
& y_{5} \leftrightarrow \\
& x_{3} \leq 4 \\
& -x_{3} \leq 2 \\
& x_{4} \leq 10 \\
& x_{1}, x_{4} \geq 0 \\
& x_{2}, x_{3} \text { unsigned } \\
& y_{7} \leftrightarrow
\end{aligned}
$$

## EXAMPLE 5: THE DUAL

$$
\min W=8 y_{1}+8 y_{2}+4 y_{3}+4 y_{4}+4 y_{5}+2 y_{6}+10 y_{7}
$$

sot.

$$
\begin{aligned}
& x_{1} \leftrightarrow y_{1}+y_{2} \\
& x_{2} \leftrightarrow y_{1} \quad+y_{3}-y_{4} \\
& x_{3} \leftrightarrow y_{1} \\
& x_{4} \leftrightarrow y_{1} \\
& +y_{7} \geq-1 \\
& y_{2}, \ldots \ldots ., y_{7} \geq 0 \\
& y_{1} \text { unsigned }
\end{aligned}
$$

## EXAMPLE 5: c.s. conditions

## We are given that

$$
\underline{x}^{*}=\left[\begin{array}{r}
8 \\
-4 \\
4 \\
0
\end{array}\right]
$$

is optimal for $(P)$
$\square$ Then the c.s. conditions obtain

$$
x_{1}^{*}\left(y_{1}^{*}+y_{2}^{*}-1\right)=0
$$

## EXAMPLE 5: c.s.conditions

so that

$$
x_{1}^{*}=8>0 \Rightarrow y_{1}^{*}+y_{2}^{*}=1
$$

and so $y_{2}^{*}=1-y_{1}^{*}$
$\square$ The other c.s. conditions require

$$
y_{i}^{*}\left(\sum_{j=1}^{4} a_{i j} x_{j}^{*}-b_{i}\right)=0
$$

$\square$ Now, $x_{4}^{*}=0$ implies $x_{4}^{*}-10<0$ and so $y_{7}^{*}=0$

## EXAMPLE 5: c.s.conditions

$\square$ Also, $x_{3}^{*}=4$ implies

$$
y_{6}^{*}=0
$$

$\square$ We similarly use the c.s. conditions

$$
x_{j}^{*}\left(\sum_{i=1}^{7} a_{j i} y_{i}^{*}-c_{j}\right)=0
$$

to provide implications on the $y_{i}^{*}$ variables

## EXAMPLE 5: c.s. conditions

$\square$ Since $x_{2}^{*}=-4$, then we have

$$
y_{3}^{*}=0
$$

$\square$ Now, with $y_{7}^{*}=0$ we have

$$
y_{1}^{*}>-1
$$

$\square$ Now, we have already shown that

$$
y_{2}^{*}=1-y_{1}^{*}
$$

## EXAMPLE 5

## $\square$ Suppose that

$$
y_{1}^{*}=1
$$

and so,

$$
y_{2}^{*}=0
$$

$\square$ Furthermore,

$$
y_{1}^{*}+y_{3}^{*}-y_{4}^{*}=1-y_{4}^{*}=-1
$$

implies that

$$
y_{4}^{*}=2
$$

## EXAMPLE 5

## $\square$ Also

$$
y_{1}^{*}+y_{5}^{*}-y_{6}^{*}=1
$$

## implies

$$
1+y_{5}^{*}=1
$$

and so

$$
y_{5}^{*}=0
$$

## EXAMPLE 5

Therefore, as $W=\underline{b}^{T} \underline{y}$

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

and so

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

## PRIMAL - DUAL TABLE

| primal (maximize) | dual (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 ) |
| $\boldsymbol{i}^{\text {th }}$ constraint is $=$ type | the dual variable $y_{i}$ is unrestricted in sign |
| $\boldsymbol{i}^{\text {th }}$ constraint is $\leq$ type | the dual variable $y_{i} \geq 0$ |
| $\boldsymbol{i}^{\text {th }}$ constraint is $\geq$ type | the dual variable $y_{i} \leq 0$ |
| $\boldsymbol{x}_{j}$ is unrestricted | $j^{\text {th }}$ dual constraint is $=$ type |
| $x_{j} \geq 0$ | $j^{\text {th }}$ dual constraint is $\geq$ type |
| $x_{j} \leq 0$ | $j^{\text {th }}$ dual constraint is $\leq$ type |


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