

# Plan

- Dual of a Primal
- Fundamental theorem of Duality
- The complementary slackness Theorem

## Dual of a Primal

- Given an LPP (P) (\*)  
Max  $\mathbf{c}^T \mathbf{x}$   
subject to  $\mathbf{A}_{m \times n} \mathbf{x} \leq \mathbf{b}_{m \times 1}$ ,  $\mathbf{x} \geq \mathbf{0}$ .
- $Fea(P) = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{A}\mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}$ .
- The **Dual** of this problem is given by (D) (\*\*)  
Min  $\mathbf{b}^T \mathbf{y}$   
subject to  $\mathbf{A}_{n \times m}^T \mathbf{y} \geq \mathbf{c}_{n \times 1}$ ,  $\mathbf{y} \geq \mathbf{0}$ .
- $Fea(D) = \{\mathbf{y} \in \mathbb{R}^m : \mathbf{A}^T \mathbf{y} \geq \mathbf{c}, \mathbf{y} \geq \mathbf{0}\}$ .
- The LPP(\*) is called a **Primal** problem.

- **Theorem 1:** If  $\mathbf{x} \in \text{Fea}(P)$  and  $\mathbf{y} \in \text{Fea}(D)$ , then  $\mathbf{c}^T \mathbf{x} \leq \mathbf{b}^T \mathbf{y}$ .
- **Corollary 1:** If both the Primal and the Dual have feasible solutions then both have optimal solutions.
- **Corollary 2:** Let  $\mathbf{x}_0 \in \text{Fea}(P)$  and  $\mathbf{y}_0 \in \text{Fea}(D)$ , be such that  $\mathbf{c}^T \mathbf{x}_0 = \mathbf{b}^T \mathbf{y}_0$ , then  $\mathbf{x}_0$  and  $\mathbf{y}_0$  are optimal for the Primal and the Dual, respectively.

- **Example 1:** Consider the problem (P)  
 Max  $5x_1 + 2x_2$  subject to  
 $3x_1 + 2x_2 \leq 6$   
 $x_1 + 2x_2 \leq 4$   
 $x_1 \geq 0, x_2 \geq 0$ .
- The dual of this problem is given by (D),  
 Min  $6y_1 + 4y_2$   
 subject to  
 $3y_1 + y_2 \geq 5$   
 $2y_1 + 2y_2 \geq 2$ ,  
 $y_1 \geq 0, y_2 \geq 0$ .
- $\mathbf{x} = [2, 0]^T$  is a feasible solution of (P) and  $\mathbf{y} = [\frac{5}{3}, 0]^T$  is a feasible solution of (D)  
 such that  $\mathbf{b}^T \mathbf{y} = \mathbf{c}^T \mathbf{x} = 10$ , where  $\mathbf{c} = [5, 2]^T$  and  $\mathbf{b} = [6, 4]^T$ .
- Hence  $\mathbf{x} = [2, 0]^T$  is optimal for the **Primal (P)** and  $\mathbf{y} = [1, 1]^T$  is optimal for the **Dual (D)**.

- **Theorem 2: ( Fundamental Theorem of Duality )**

If **both** the **Primal** (P) and the **Dual** (D) of the Primal (P) have **feasible solutions** then **both** have **optimal solutions** and the **optimal value** of (P) and (D) equal ( that is  $\text{Min } \mathbf{b}^T \mathbf{y} = \text{Max } \mathbf{c}^T \mathbf{x}$ ). If one of them (the Primal or the Dual) does not have a **feasible solution** then the other does not have an **optimal solution**.

- To complete the proof it is enough to show that
- when both (P) and (D) have **feasible solutions** then their **optimal values** are equal.
- if any one of (P) or (D) **does not** have a **feasible solution** then the other **does not** have an **optimal solution**.
- **Theorem 3:** The **Dual** of the **Dual** (D) (of the Primal (P)) is the **Primal** (P).

- **Example 2:** Consider (P)

$$\text{Min } -x + 2y$$

subject to

$$x + 2y \geq 1$$

$$-x + y \leq 1$$

$$x \geq 0, y \geq 0.$$

- (P) **does not** have an **optimal solution** although it has a **feasible solution**.

The dual (D) of (P) is given by

$$\text{Max } u - v$$

subject to

$$u + v \leq -1$$

$$2u - v \leq 2,$$

$$u \geq 0, v \geq 0.$$

- Clearly (D) **does not** have any **feasible solution**, since the first constraint cannot be satisfied by any non negative  $u$  and  $v$ .

- **Example 3:** The LPP (P)

$$\text{Max } -x + 2y$$

subject to

$$x + 2y \leq 1$$

$$-x + y \geq 1,$$

$$x, y \geq 0.$$

- (P) **does not** have a **feasible solution**, hence **does not** have an **optimal solution**.
- The dual of (P) is given by (D)

$$\text{Min } u - v$$

subject to

$$u + v \geq -1$$

$$2u - v \geq 2$$

$$u \geq 0, v \geq 0.$$

Check that (D) has a **feasible solution** but **no optimal solution**.

- **Definition:** A nonempty set  $T \subseteq \mathbb{R}^n$  is said to be a **cone** if  $\mathbf{x} \in T$  implies  $\lambda \mathbf{x} \in T$  for all  $\lambda \geq 0$ .
- So a cone always contains the **origin**.
- Also a **cone**  $T$  is said to be a **convex cone** if it is also a **convex subset** of  $\mathbb{R}^n$ .
- **Exercise:** If  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k$  are all vectors in  $\mathbb{R}^n$  then check that  

$$T = \{\alpha_1 \mathbf{a}_1 + \alpha_2 \mathbf{a}_2 + \dots + \alpha_k \mathbf{a}_k : \alpha_i \geq 0 \text{ for all } i = 1, \dots, k\}$$
is a **convex cone**.
- $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k$  are called **generators** of the **convex cone**  $T$ .



## Fundamental Theorem of Duality (revisited):

- **Theorem 4: (Farka's Lemma)** Exactly **one** of the following two systems has a solution.

$$A_{m \times n} \mathbf{x} = \mathbf{b}, \mathbf{x} \geq \mathbf{0} \quad (1)$$

$$\mathbf{y}^T A \geq \mathbf{0}_{1 \times n}, \mathbf{y}^T \mathbf{b} < 0 \quad (2)$$

- **Corollary 4:** Exactly **one** of the following two systems has a solution.

$$A_{m \times n} \mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0} \quad (1)$$

$$\mathbf{y}^T A \geq \mathbf{0}, \mathbf{y}^T \mathbf{b} < 0, \mathbf{y} \geq \mathbf{0} \quad (2)$$

- Since  $Fea(D) = \{\mathbf{y} \in \mathbb{R}^m : A^T \mathbf{y} \geq \mathbf{c}, \mathbf{y} \geq \mathbf{0}\}$ , the set of all directions of  $Fea(D)$  is given by
- $D_D = \{\mathbf{d} \in \mathbb{R}^m : A^T \mathbf{d} \geq \mathbf{0}, \mathbf{d} \geq \mathbf{0}\}$
- The **dual** (D) has an optimal solution if and only if either  $Fea(D)$  is **bounded** or if not then  $\mathbf{b}^T \mathbf{d} \geq 0$  for all  $\mathbf{d} \in D_D$ .
- From **Corollary 4** we get if the **Primal** does **not** have a **feasible solution** then the **dual** does not have an **optimal solution**.
- From the **fundamental theorem of duality** we get that if any one of **Primal** or **Dual** has an **optimal solution** then the other also has an **optimal solution**.

- To complete the proof of Fundamental Theorem of Duality we need to show that if  $Fea(P) \neq \phi$  and  $Fea(D) \neq \phi$ , then the following system (1) has a solution:

$$A_{m \times n} \mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0} \quad (1)$$

$$A^T \mathbf{y} \geq \mathbf{c}, \mathbf{y} \geq \mathbf{0}$$

$$\mathbf{c}^T \mathbf{x} = \mathbf{b}^T \mathbf{y}$$

- If  $\mathbf{x} \in Fea(P)$  and  $\mathbf{y} \in Fea(D)$  then  $\mathbf{c}^T \mathbf{x} \leq \mathbf{b}^T \mathbf{y}$ , hence system (1) has a solution  $\Leftrightarrow$  system (1)'' has a solution:

$$A_{m \times n} \mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0} \quad (1)''$$

$$A^T \mathbf{y} \geq \mathbf{c}, \mathbf{y} \geq \mathbf{0}$$

$$\mathbf{c}^T \mathbf{x} \geq \mathbf{b}^T \mathbf{y}$$

- System (1)'' can be written as:

$$\begin{bmatrix} A_{m \times n} & \mathbf{0}_{m \times m} \\ \mathbf{0}_{n \times n} & -A_{n \times m}^T \\ -\mathbf{c}_{1 \times n}^T & \mathbf{b}_{1 \times m}^T \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \leq \begin{bmatrix} \mathbf{b} \\ -\mathbf{c} \\ 0 \end{bmatrix}, \quad \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \geq \begin{bmatrix} \mathbf{0}_{n \times 1} \\ \mathbf{0}_{m \times 1} \end{bmatrix}$$

- By **corollary 4**, exactly one of the two systems, (1'') (given above) and (2'') (given below) has a solution.

$$\begin{bmatrix} \mathbf{z}_{1 \times m}^T & \mathbf{w}_{1 \times n}^T & a_{1 \times 1} \end{bmatrix} \begin{bmatrix} A_{m \times n} & \mathbf{0}_{m \times m} \\ \mathbf{0}_{n \times n} & -A_{n \times m}^T \\ -\mathbf{c}_{1 \times n}^T & \mathbf{b}_{1 \times m}^T \end{bmatrix} \geq \begin{bmatrix} \mathbf{0}_{1 \times n} & \mathbf{0}_{1 \times m} \end{bmatrix},$$

$$\begin{bmatrix} \mathbf{z}_{1 \times m}^T & \mathbf{w}_{1 \times n}^T & a_{1 \times 1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ -\mathbf{c} \\ 0 \end{bmatrix} < 0 \quad \begin{bmatrix} \mathbf{z}_{1 \times m}^T & \mathbf{w}_{1 \times n}^T & a_{1 \times 1} \end{bmatrix} \geq \mathbf{0}_1$$

- We show in the proof that given  $Fea(P) \neq \phi$  and  $Fea(D) \neq \phi$ , system (2)'' does not have a solution, hence system (1'') has solution.
- **Theorem 5 ( Complementary Slackness Theorem ):** Let  $\mathbf{x} \in Fea(P)$  and  $\mathbf{y} \in Fea(D)$ , then  $\mathbf{x}$  and  $\mathbf{y}$  are optimal for the primal and the dual respectively if and only if  
 $x_j = 0$  whenever  $(A^T \mathbf{y})_j > c_j, j = 1, 2, \dots, n$  (1)  
 and  
 $y_i = 0$  whenever  $(A\mathbf{x})_i < b_i, i = 1, 2, \dots, m.$  (1\*)

- **Example 1:** Consider the following primal problem (P):

$$\text{Max } 4x_1 + 4x_2 + 2x_3$$

subject to

$$2x_1 + 3x_2 + 4x_3 \leq 10 \quad (i)$$

$$2x_1 + x_2 + 3x_3 \leq 4 \quad (ii)$$

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

The dual (D) of the above problem is given by:

$$\text{Min } 10y_1 + 4y_2$$

subject to

$$2y_1 + 2y_2 \geq 4 \quad (i)'$$

$$3y_1 + y_2 \geq 4 \quad (ii)'$$

$$4y_1 + 3y_2 \geq 2 \quad (iii)'$$

$$y_1, y_2 \geq 0$$