

# 6.215/6.255J/15.093J/IDS.200J Optimization Methods

## Lecture 5: Duality Theory I

September 23, 2021

# Today's Lecture

## Outline

- Motivation of duality
- General form of the dual
- Weak and strong duality
- Relations between primal and dual
- Duality - economic interpretation
- Complementary slackness

# Motivating Duality

## Checking optimality

$$\min \quad x_1 + x_2 + \cdots + x_{300}$$

$$\text{s.t.} \quad x_1 + 2x_2 + \cdots + 300x_{300} \geq 1$$

$$300x_1 + 299x_2 + \cdots + x_{300} \geq 10$$

A friend claims to have found an optimal solution  $\mathbf{x}^*$  with an optimal value  $11/301$ . How to check this?

# Duality - motivation

An idea from Lagrange

Consider the LOP, called the **primal** with optimal solution  $\mathbf{x}^*$

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

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Relax the equality constraint

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

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

$$\Rightarrow g(\mathbf{p}) \leq \mathbf{c}^T \mathbf{x}^* + \mathbf{p}^T (\mathbf{b} - \mathbf{Ax}^*) = \mathbf{c}^T \mathbf{x}^*$$

# Duality - motivation

An idea from Lagrange

Get the tightest lower bound, i.e.,  $\max g(\mathbf{p})$  over all possible  $\mathbf{p}$ .

Let us rewrite  $g(\mathbf{p})$ :

$$\begin{aligned} g(\mathbf{p}) &= \min_{\mathbf{x} \geq 0} \left[ \mathbf{c}^T \mathbf{x} + \mathbf{p}^T (\mathbf{b} - \mathbf{A} \mathbf{x}) \right] \\ &= \mathbf{p}^T \mathbf{b} + \min_{\mathbf{x} \geq 0} (\mathbf{c}^T - \mathbf{p}^T \mathbf{A}) \mathbf{x} \end{aligned}$$

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Note that

$$\min_{\mathbf{x} \geq 0} (\mathbf{c}^T - \mathbf{p}^T \mathbf{A})\mathbf{x} = \begin{cases} 0 & \text{if } \mathbf{c}^T - \mathbf{p}^T \mathbf{A} \geq 0^T, \\ -\infty & \text{otherwise.} \end{cases}$$



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So we have

$$\text{dual : } \max_{\mathbf{p}} g(\mathbf{p}) \quad \Leftrightarrow \quad \max_{\substack{\mathbf{p} \\ \text{s.t. } \mathbf{p}^T \mathbf{A} \leq \mathbf{c}^T}} \mathbf{p}^T \mathbf{b}$$

# General form of the dual

## Primal

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} \geq b_i \quad i \in M_1 \\ & \mathbf{a}_i^T \mathbf{x} \leq b_i \quad i \in M_2 \\ & \mathbf{a}_i^T \mathbf{x} = b_i \quad i \in M_3 \\ & x_j \geq 0 \quad j \in N_1 \\ & x_j \leq 0 \quad j \in N_2 \\ & x_j \text{ free} \quad j \in N_3 \end{array}$$

## Dual

$$\begin{array}{ll} \max & \mathbf{p}^T \mathbf{b} \\ \text{s.t.} & p_i \geq 0 \quad i \in M_1 \\ & p_i \leq 0 \quad i \in M_2 \\ & p_i \text{ free} \quad i \in M_3 \\ & \mathbf{p}^T \mathbf{A}_j \leq c_j \quad j \in N_1 \\ & \mathbf{p}^T \mathbf{A}_j \geq c_j \quad j \in N_2 \\ & \mathbf{p}^T \mathbf{A}_j = c_j \quad j \in N_3 \end{array}$$

# General form of the dual

## Example

$$\begin{array}{ll}\min & x_1 + 2x_2 + 3x_3 \\ \text{s.t.} & -x_1 + 3x_2 = 5 \\ & 2x_1 - x_2 + 3x_3 \geq 6 \\ & x_3 \leq 4 \\ & x_1 \geq 0 \\ & x_2 \leq 0 \\ & x_3 \text{ free,}\end{array}$$

$$\begin{array}{ll}\max & 5p_1 + 6p_2 + 4p_3 \\ \text{s.t.} & p_1 \text{ free} \\ & p_2 \geq 0 \\ & p_3 \leq 0 \\ & -p_1 + 2p_2 \leq 1 \\ & 3p_1 - p_2 \geq 2 \\ & 3p_2 + p_3 = 3.\end{array}$$

# General form of the dual

## Summary

primal	min	max	dual
constraints	$\geq b_i$ $\leq b_i$ $= b_i$	$\geq 0$ $\leq 0$ free	variables
variables	$\geq 0$ $\leq 0$ free	$\leq c_j$ $\geq c_j$ $= c_j$	constraints

Note: The dual of the dual is the primal

# The dual of the dual is the primal

## Example

$$\begin{array}{llllll}
 \min & x_1 & + & 2x_2 & + & 3x_3 \\
 \text{s.t.} & -x_1 & + & 3x_2 & & = 5 \\
 & 2x_1 & - & x_2 & + & 3x_3 \geq 6 \\
 & & & & x_3 & \leq 4 \\
 & x_1 \geq 0 & & & & \\
 & x_2 \leq 0 & & & & \\
 & x_3 \text{ free} & & & & 
 \end{array}$$

$$\begin{array}{llllll}
 \max & 5p_1 & + & 6p_2 & + & 4p_3 \\
 \text{s.t.} & p_1 \text{ free} & & & & \\
 & p_2 \geq 0 & & & & \\
 & p_3 \leq 0 & & & & \\
 & -p_1 & + & 2p_2 & & \leq 1 \\
 & 3p_1 & - & p_2 & & \geq 2 \\
 & & & 3p_2 & + & p_3 = 3.
 \end{array}$$

$$\begin{array}{llllll}
 \min & -5x_1 & - & 6x_2 & - & 4x_3 \\
 \text{s.t.} & x_1 \text{ free} & & & & \\
 & x_2 \geq 0 & & & & \\
 & x_3 \leq 0 & & & & \\
 & x_1 & - & 2x_2 & & \geq -1 \\
 & -3x_1 & + & x_2 & & \leq -2 \\
 & & - & 3x_2 & - & x_3 = -3
 \end{array}$$

$$\begin{array}{llllll}
 \max & -p_1 & - & 2p_2 & - & 3p_3 \\
 \text{s.t.} & p_1 & - & 3p_2 & & = -5 \\
 & -2p_1 & + & x_2 & - & 3p_3 \leq -6 \\
 & & & & - & p_3 \geq -4 \\
 & p_1 \geq 0 & & & & \\
 & p_2 \leq 0 & & & & \\
 & p_3 \text{ free} & & & & 
 \end{array}$$

# General form of the dual

Compactly, in matrix notation, some pairs of primal-dual

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

$$\begin{array}{ll}\max & \mathbf{p}^T \mathbf{b} \\ \text{s.t.} & \mathbf{p}^T \mathbf{A} \leq \mathbf{c}^T\end{array}$$

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

$$\begin{array}{ll}\max & \mathbf{p}^T \mathbf{b} \\ \text{s.t.} & \mathbf{p}^T \mathbf{A} = \mathbf{c}^T \\ & \mathbf{p} \geq 0\end{array}$$

# Weak duality

## Theorem

*If  $\mathbf{x}$  is primal feasible and  $\mathbf{p}$  is dual feasible then  $\mathbf{p}^T \mathbf{b} \leq \mathbf{c}^T \mathbf{x}$*

Proof: If the primal is in standard form:  $\mathbf{p}^T \mathbf{b} = \mathbf{p}^T \mathbf{A} \mathbf{x} \leq \mathbf{c}^T \mathbf{x}$

## Corollary

*If  $\mathbf{x}$  is primal feasible,  $\mathbf{p}$  is dual feasible, and  $\mathbf{p}^T \mathbf{b} = \mathbf{c}^T \mathbf{x}$ , then  $\mathbf{x}$  is optimal in the primal and  $\mathbf{p}$  is optimal in the dual.*

# Strong duality

## Theorem

*If the primal has an optimal solution, then so does the dual, and the optimal costs are equal.*

Proof:

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

Apply (with any anticycling rule) simplex method; get optimal BFS  $\mathbf{x}$ , and corresponding final optimal basis  $\mathbf{B}$ , then:

$$\mathbf{c}^T - \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{A} \geq 0$$



# Strong duality

Define  $\mathbf{p}^T = \mathbf{c}_B^T \mathbf{B}^{-1} \Rightarrow \mathbf{p}^T \mathbf{A} \leq \mathbf{c}^T$

$\Rightarrow \mathbf{p}$  dual feasible for

$$\begin{array}{ll} \max & \mathbf{p}^T \mathbf{b} \\ \text{s.t.} & \mathbf{p}^T \mathbf{A} \leq \mathbf{c}^T \end{array}$$

Moreover,

$$\mathbf{p}^T \mathbf{b} = \mathbf{c}_B^T \mathbf{B}^{-1} \mathbf{b} = \mathbf{c}_B^T \mathbf{x}_B = \mathbf{c}^T \mathbf{x}$$

$\Rightarrow \mathbf{x}, \mathbf{p}$  are primal and dual optimal

# Relations between primal and dual

	<b>Finite opt.</b>	<b>Unbounded</b>	<b>Infeasible</b>
<b>Finite opt.</b>	*		
<b>Unbounded</b>			*
<b>Infeasible</b>		*	*

# Duality

## Economic interpretation

- Let  $\mathbf{x}$  optimal nondegenerate solution:  $\mathbf{B}^{-1}\mathbf{b} > 0$
- Suppose  $\mathbf{b}$  changes to  $\mathbf{b} + \epsilon$  for some small  $\epsilon$
- How is the optimal cost affected?

# Duality

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  - feasibility unaffected
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## Economic interpretation

- Let  $\mathbf{x}$  optimal nondegenerate solution:  $\mathbf{B}^{-1}\mathbf{b} > 0$
- Suppose  $\mathbf{b}$  changes to  $\mathbf{b} + \epsilon$  for some small  $\epsilon$
- How is the optimal cost affected?
- For small  $\epsilon$ :
  - feasibility unaffected
  - optimality conditions unaffected
- New cost  $\mathbf{c}_B^T \mathbf{B}^{-1}(\mathbf{b} + \epsilon) = \mathbf{p}^T(\mathbf{b} + \epsilon)$
- If resource  $i$  changes by  $\epsilon_i$ , cost changes by  $p_i \epsilon_i$ : “marginal cost”

# Complementary slackness

**Primal**

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} \geq b_i \quad i \in M_1 \\ & \mathbf{a}_i^T \mathbf{x} \leq b_i \quad i \in M_2 \\ & \mathbf{a}_i^T \mathbf{x} = b_i \quad i \in M_3 \\ & x_j \geq 0 \quad j \in N_1 \\ & x_j \leq 0 \quad j \in N_2 \\ & x_j \text{ free} \quad j \in N_3 \end{array}$$

**Dual**

$$\begin{array}{ll} \max & \mathbf{p}^T \mathbf{b} \\ \text{s.t.} & p_i \geq 0 \quad i \in M_1 \\ & p_i \leq 0 \quad i \in M_2 \\ & p_i \text{ free} \quad i \in M_3 \\ & \mathbf{p}^T \mathbf{A}_j \leq c_j \quad j \in N_1 \\ & \mathbf{p}^T \mathbf{A}_j \geq c_j \quad j \in N_2 \\ & \mathbf{p}^T \mathbf{A}_j = c_j \quad j \in N_3 \end{array}$$

## Theorem

Let  $\mathbf{x}$  primal feasible and  $\mathbf{p}$  dual feasible. Then  $\mathbf{x}, \mathbf{p}$  optimal if and only if

$$p_i(\mathbf{a}_i^T \mathbf{x} - b_i) = 0, \quad \forall i$$

$$(c_j - \mathbf{p}^T \mathbf{A}_j)x_j = 0, \quad \forall j$$

# Complementary slackness

Proof:

- If  $\mathbf{x}$  primal feasible and  $\mathbf{p}$  dual feasible, we have  $u_i = p_i(\mathbf{a}_i^T \mathbf{x} - b_i) \geq 0$  and  $v_j = (c_j - \mathbf{p}^T \mathbf{A}_j)x_j \geq 0$  for all  $i$  and  $j$ .
- Also  $\mathbf{c}^T \mathbf{x} - \mathbf{p}^T \mathbf{b} = \sum_i u_i + \sum_j v_j$ .
- By the strong duality theorem, if  $\mathbf{x}$  and  $\mathbf{p}$  are optimal, then  $\mathbf{c}^T \mathbf{x} = \mathbf{p}^T \mathbf{b} \Rightarrow u_i = v_j = 0$  for all  $i, j$ .
- Conversely, if  $u_i = v_j = 0$  for all  $i, j$ , then  $\mathbf{c}^T \mathbf{x} = \mathbf{p}^T \mathbf{b}$ ,  $\Rightarrow \mathbf{x}$  and  $\mathbf{p}$  are optimal.

# Complementary slackness

## Example

$$\begin{array}{ll}\min & 13x_1 + 10x_2 + 6x_3 \\ \text{s.t.} & 5x_1 + x_2 + 3x_3 = 8 \\ & 3x_1 + x_2 = 3 \\ & x_1, x_2, x_3 \geq 0\end{array}$$

$$\begin{array}{ll}\max & 8p_1 + 3p_2 \\ \text{s.t.} & 5p_1 + 3p_2 \leq 13 \\ & p_1 + p_2 \leq 10 \\ & 3p_1 \leq 6\end{array}$$

Is  $\mathbf{x}^* = (1, 0, 1)^T$  optimal?



# Complementary slackness

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$$\begin{array}{ll}\min & 13x_1 + 10x_2 + 6x_3 \\ \text{s.t.} & 5x_1 + x_2 + 3x_3 = 8 \\ & 3x_1 + x_2 = 3 \\ & x_1, x_2, x_3 \geq 0\end{array}$$

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Is  $\mathbf{x}^* = (1, 0, 1)^T$  optimal?

$$5p_1 + 3p_2 = 13, \quad 3p_1 = 6$$

$$\Rightarrow p_1 = 2, \quad p_2 = 1$$

It satisfies  $p_1 + p_2 \leq 10$ , so dual feasible.

Objective =  $2 \cdot 8 + 3 \cdot 1 = 19 = 13 + 6 = 19$ , so yes  $\mathbf{x}^*$  optimal.