

# MAT 3007 Tutorial 6: Duality Theory

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June 24, 2025

# Duality Theory

Primal	Dual
minimize $\mathbf{c}^T \mathbf{x}$	maximize $\mathbf{b}^T \mathbf{y}$
subject to $\mathbf{a}_i^T \mathbf{x} \geq b_i, \quad i \in M_1,$	subject to $y_i \geq 0, \quad i \in M_1$
$\mathbf{a}_i^T \mathbf{x} \leq b_i, \quad i \in M_2,$	$y_i \leq 0, \quad i \in M_2$
$\mathbf{a}_i^T \mathbf{x} = b_i, \quad i \in M_3,$	$y_i \text{ free}, \quad i \in M_3$
$x_j \geq 0, \quad j \in N_1,$	$A_j^T \mathbf{y} \leq c_j, \quad j \in N_1$
$x_j \leq 0, \quad j \in N_2,$	$A_j^T \mathbf{y} \geq c_j, \quad j \in N_2$
$x_j \text{ free}, \quad j \in N_3,$	$A_j^T \mathbf{y} = c_j, \quad j \in N_3$

Primal	minimize	maximize	Dual
Constraints	$\geq b_i$	$\geq 0$	Variables
	$\leq b_i$	$\leq 0$	
	$= b_i$	free	
Variables	$\geq 0$	$\leq c_j$	Constraints
	$\leq 0$	$\geq c_j$	
	free	$= c_j$	

# Duality Theory

## Weak Duality Theorem:

If  $x$  is feasible to the primal minimization and  $y$  is feasible to the dual maximization, then  $b^T y \leq c^T x$ .

## Strong Duality Theorem:

If a linear program has a finite optimal solution, so does its dual, and the optimal values of the primal and dual are equal.

## Optimal Conditions:

- (1)  $x$  is primal feasible,
- (2)  $y$  is dual feasible,
- (3)  $b^T y = c^T x$ .

## Exercise 1: Solving primal LP by the dual

Use the **dual problem** to completely solve the linear optimization problem:

$$\begin{aligned}
 \min \quad & 4x_1 + x_2 + x_3 \\
 \text{s.t.} \quad & 2x_1 + x_2 + 2x_3 = 4 \\
 & 3x_1 + 3x_2 + x_3 = 3 \\
 & x \geq 0
 \end{aligned}$$

Primal	minimize	maximize	Dual
Constraints	$\geq b_i$	$\geq 0$	Variables
	$\leq b_i$	$\leq 0$	
	$= b_i$	free	
Variables	$\geq 0$	$\leq c_j$	Constraints
	$\leq 0$	$\geq c_j$	
	free	$= c_j$	

## Solution to Exercise 1

**Make sure  $\mathbf{c}$  is nonnegative** (We are happily already in this situation),  
The corresponding dual problem is:

$$\begin{aligned} \max_{\mathbf{y}} \quad & 4y_1 + 3y_2 \\ \text{s.t.} \quad & \begin{bmatrix} 2 & 3 \\ 1 & 3 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \leq \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix} \end{aligned}$$

Transform to standard LP:

$$\begin{aligned} \min_{y^+, y^-, s} \quad & -4y_1^+ - 3y_2^+ + 4y_1^- + 3y_2^- \\ \text{s.t.} \quad & \begin{bmatrix} 2 & 3 \\ 1 & 3 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} y_1^+ - y_1^- \\ y_2^+ - y_2^- \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix} \\ & \mathbf{y}^+, \mathbf{y}^-, \mathbf{s} \geq 0 \end{aligned}$$

## Solution to Exercise 1

Construct the initial tableau for the standard dual problem with basis  $B=\{5,6,7\}$  ,  $c_B=\{0,0,0\}$ ,  $x=(0,0,0,0,1,1,1)$  and  $A_B$  is identity matrix.

Table: Iteration 1

B	-4	-3	4	3	0	0	0	0
5	2	3	-2	-3	1	0	0	4
6	1	3	-1	-3	0	1	0	1
7	2	1	-2	-1	0	0	1	1

Actually this is nicely a canonical form, since we add slack variables for initialization.

For convenience, we use  $x_i$  to represent index  $i$  in the following iteration.

## Solution to Exercise 1

We bring  $x_1$  into the basis and have  $x_7$  exit the basis, the new tableau is:

Table: Iteration 2

B	0	-1	0	1	0	0	2	2
5	0	2	0	-2	1	0	-1	3
6	0	5/2	0	$-\frac{5}{2}$	0	1	$-\frac{1}{2}$	$\frac{1}{2}$
1	1	$\frac{1}{2}$	-1	$-\frac{1}{2}$	0	0	$\frac{1}{2}$	$\frac{1}{2}$

We bring  $x_2$  into the basis and have  $x_6$  exit the basis, the new tableau is:

Table: Iteration 3

B	0	0	0	0	0	$\frac{2}{5}$	$\frac{9}{5}$	$\frac{11}{5}$
5	0	0	0	0	1	$-\frac{4}{5}$	$-\frac{3}{5}$	$\frac{13}{5}$
2	0	1	0	1	0	$\frac{2}{5}$	$-\frac{1}{5}$	$\frac{1}{5}$
1	1	0	-1	0	0	$-\frac{1}{5}$	$\frac{3}{5}$	$\frac{2}{5}$

## Solution to Exercise 1

Now we reach the optimality. Let's find what we want.

Optimal Value: By tableau, optimal value for **the standard min problem** is  $-\frac{11}{5}$ , then optimal value of **the original dual problem** is  $\frac{11}{5}$ . By strong duality of LP, optimal value of **the primal problem** is  $\frac{11}{5}$ .

i.e. **The number in the top right corner of dual tableau is just the primal optimal value.**

Optimal Solution: Since we use tableau in canonical form at first. **(if the primal is in standard form and  $c \geq 0$ , we will always do that!)** The **primal** optimal solution is the final reduced costs corresponding to the original identity matrix part in tableau, i.e..

$$x^* = \left(0, \frac{2}{5}, \frac{9}{5}\right)$$

P.S.: The precondition ' $c \geq 0$ ' is necessary for such nice property of Optimal solution.



## Exercise 2: Deriving the dual form

Consider the problem

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

1. Derive the dual problem of
  - the original problem.
  - the original problem with slack variable  $s$
  - the original problem with  $x$  replaced by  $x^+$  and  $x^-$ .
2. Show that these three dual problems are equal.

## Solution to Exercise 2

$$\begin{aligned}
 (P1) \quad & \min && c^T x \\
 & \text{s.t.} && Ax \geq b
 \end{aligned}$$

$$\begin{aligned}
 (D1) \quad & \max && b^T y \\
 & \text{s.t.} && A^T y = c \\
 & && y \geq 0
 \end{aligned}$$

$$\begin{aligned}
 (P2) \quad & \min && c^T x + 0^T s \\
 & \text{s.t.} && Ax - s = b \\
 & && s \geq 0
 \end{aligned}$$

$$\begin{aligned}
 (D2) \quad & \max && b^T y \\
 & \text{s.t.} && A^T y = c \\
 & && -I^T y \leq 0
 \end{aligned}$$

$$\begin{aligned}
 (P3) \quad & \min && c^T x^+ - c^T x^- \\
 & \text{s.t.} && Ax^+ - Ax^- \geq b \\
 & && x^+, x^- \geq 0
 \end{aligned}$$

$$\begin{aligned}
 (D3) \quad & \max && b^T y \\
 & \text{s.t.} && A^T y \leq c \\
 & \text{s.t.} && -A^T y \leq -c \\
 & && y \geq 0
 \end{aligned}$$

## Exercise 3 - Strong Duality and Optimality Conditions

Let  $A$  be a **symmetric** square matrix. Consider the linear programming problem:

$$\begin{aligned} \min \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & A\mathbf{x} \geq \mathbf{c} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned} \tag{1}$$

Prove that if  $\mathbf{x}^*$  satisfies  $A\mathbf{x}^* = \mathbf{c}$  and  $\mathbf{x}^* \geq \mathbf{0}$ , then  $\mathbf{x}^*$  is an optimal solution.

## Solution to Exercise 3

Primal:

$$\begin{aligned}
 \min \quad & \mathbf{c}^T \mathbf{x} \\
 \text{s.t.} \quad & A\mathbf{x} \geq \mathbf{c} \\
 & \mathbf{x} \geq \mathbf{0}
 \end{aligned} \tag{2}$$

Dual:

$$\begin{aligned}
 \max \quad & \mathbf{c}^T \mathbf{y} \\
 \text{s.t.} \quad & A^T \mathbf{y} \leq \mathbf{c} \\
 & \mathbf{y} \geq \mathbf{0}
 \end{aligned} \tag{3}$$

For  $\mathbf{x}^*$  s.t.  $A\mathbf{x}^* = \mathbf{c}$  and  $\mathbf{x}^* \geq \mathbf{0}$ ,  $\mathbf{x}^*$  is both primal and dual feasible, and both have the same objective value.

Therefore,  $\mathbf{x}^*$  is optimal solution to both.

Q.E.D.

## Simple Review for Mid-term Exam

### What we have learned

- Modeling
- Linear Programming (LP)
  - Definition of Standard form, Basic solution, Feasible solution, Basic feasible solution
  - Techniques for transformation to a standard LP
  - Solve LP graphically
  - Solve LP algorithmically: **Simplex method**
  - Duality Theorem and its application