

Duality Theorems

Theorem 1: The dual problem of the dual is the primal

Proof:

D

$$\begin{array}{ll}\text{Min } & \mathbf{y}\mathbf{b} \\ \text{S.t. } & \mathbf{y}\mathbf{A} \geq \mathbf{c} \\ & \mathbf{y} \geq 0\end{array}$$

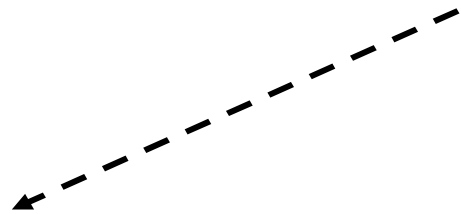
$$\begin{array}{ll}\Rightarrow & \text{Max } -\mathbf{y}\mathbf{b} \\ \text{S.t. } & -\mathbf{y}\mathbf{A} \leq -\mathbf{c} \\ & \mathbf{y} \geq 0\end{array}$$

D'

$$\begin{array}{ll}\text{Min } & -\mathbf{c}\mathbf{x} \\ \text{S.t. } & -\mathbf{A}\mathbf{x} \geq -\mathbf{b} \\ & \mathbf{x} \geq 0\end{array}$$

$$\begin{array}{ll}\Rightarrow & \text{Max } \mathbf{c}\mathbf{x} \\ \text{S.t. } & \mathbf{A}\mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \geq 0\end{array}$$

P



Primal [P]

$$\text{Max } Z = \mathbf{cx}$$

$$\text{S.t. } \mathbf{Ax} \leq \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0}$$

Dual variable

\mathbf{y}

Dual [D]

$$\text{Min } W = \mathbf{yb}$$

$$\text{S.t. } \mathbf{yA} \geq \mathbf{c}$$

$$\mathbf{y} \geq \mathbf{0}$$

Theorem 2: Weak Duality Theorem (WDT)

If \mathbf{x} is a f.s. to [P] and \mathbf{y} is a f.s. to [D], then $\mathbf{cx} \leq \mathbf{yb}$.

(i.e. maximum profit is constrained by the value of resources)

Proof: \mathbf{x} is feasible to [P] $\Rightarrow \mathbf{Ax} \leq \mathbf{b}$ (1)

\mathbf{y} is feasible to [D] $\Rightarrow \mathbf{yA} \geq \mathbf{c}$ (2)

Pre-multiply (1) by \mathbf{y} and post multiply (2) by \mathbf{x} , then

$$\mathbf{cx} \leq \mathbf{yAx} \leq \mathbf{yb} \quad \Rightarrow \quad \mathbf{cx} \leq \mathbf{yb}$$

Observations from WDT ($\mathbf{cx} \leq \mathbf{yb}$)

Note: Primal refers to the maximization problem and dual refers to the corresponding minimization problem

- 1) The Objective function value of the primal for its any feasible solution is a lower bound to the minimum value of the dual problem
- 2) Similarly, the Objective function value of the dual for its any feasible solution is an upper bound to the maximum value of the primal problem
- 3) If the primal is unbounded (i.e., $\text{Max } Z \rightarrow +\infty$), then the dual problem is infeasible.
- 4) If the dual is unbounded (i.e., $\text{Min } W \rightarrow -\infty$), then the primal is infeasible.

Theorem 3: Strong Duality Theorem (SDT)

If \mathbf{x}^* and \mathbf{y}^* are feasible to [P] and [D] respectively and $\mathbf{c}\mathbf{x}^* = \mathbf{y}^*\mathbf{b}$, then \mathbf{x}^* and \mathbf{y}^* are optimal to [P] and [D], respectively.

Proof: From WDT, $\mathbf{c}\mathbf{x} \leq \mathbf{y}\mathbf{b}$ for any feasible \mathbf{x} and \mathbf{y} .

Then $\mathbf{c}\mathbf{x} \leq \mathbf{y}^*\mathbf{b}$ for any feasible \mathbf{x} and \mathbf{y}^* .

But $\mathbf{c}\mathbf{x}^* = \mathbf{y}^*\mathbf{b}$ (Given)

Then $\mathbf{c}\mathbf{x} \leq \mathbf{c}\mathbf{x}^*$ for any feasible \mathbf{x}

$\Rightarrow \mathbf{x}^*$ is optimal to [P] {maximization problem}

Similarly, \mathbf{y}^* is optimal to [D]

Simplex Tableau for Primal at any iteration

BV	\mathbf{x}^T	\mathbf{x}_s^T	RHS
\mathbf{x}_B	$\mathbf{B}^{-1}\mathbf{A}$	\mathbf{B}^{-1}	$\mathbf{B}^{-1}\mathbf{b}$
Z	$\mathbf{c}_B\mathbf{B}^{-1}\mathbf{A} - \mathbf{c}$	$\mathbf{c}_B\mathbf{B}^{-1}$	$\mathbf{c}_B\mathbf{B}^{-1}\mathbf{b}$

Proposition: If \mathbf{B} is the optimal primal basis, then optimal solution of the dual problem is $\mathbf{y}^* = \mathbf{c}_B\mathbf{B}^{-1}$

Proof:

From SDT, $\mathbf{c}\mathbf{x}^* = \mathbf{y}^*\mathbf{b}$

$$\Rightarrow \mathbf{c}_B\mathbf{x}_B = \mathbf{y}^*\mathbf{b}$$

$$\Rightarrow \mathbf{c}_B\mathbf{B}^{-1}\mathbf{b} = \mathbf{y}^*\mathbf{b}$$

$$\Rightarrow \mathbf{y}^* = \mathbf{c}_B\mathbf{B}^{-1}$$

Economic interpretation of Duality (dual variable)

If \mathbf{B} is the optimal primal basis

At optimality, $\mathbf{X}_B = \mathbf{B}^{-1}\mathbf{b}$ and So $Z = \mathbf{C}_B \mathbf{B}^{-1}\mathbf{b} = \mathbf{C}_B \mathbf{B}^{-1}$

$$\begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \vdots \\ \mathbf{b}_i \\ \vdots \\ \mathbf{b}_m \end{bmatrix}$$

Suppose optimal basis \mathbf{B} does not change by changing the i th resource from b_i to $b_i + \Delta b_i$.

$$\mathbf{b}' = \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \vdots \\ \mathbf{b}_i + \Delta \mathbf{b}_i \\ \vdots \\ \mathbf{b}_m \end{bmatrix}$$

$$\text{So } Z' = \mathbf{C}_B \mathbf{B}^{-1} \mathbf{b}'$$

The Change in objective function,

$$\Delta Z = \mathbf{C}_B \mathbf{B}^{-1} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ \Delta \mathbf{b}_i \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} y_1^* & y_2^* & \cdots & y_i^* & \cdots & y_m^* \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ \Delta \mathbf{b}_i \\ \vdots \\ 0 \end{bmatrix}_{m \times 1} = y_i^* \Delta \mathbf{b}_i$$

$\Rightarrow \Delta Z / \Delta b_i = y_i^* = \text{Shadow Price or Dual Price}$

Complementary Slackness Theorem (CST)

- Suppose

Primal [P] Max $Z = \mathbf{c}\mathbf{x}$ s.t. $\mathbf{Ax} \leq \mathbf{b}$, $\mathbf{x} \geq \mathbf{0}$

Dual [D] Min $\mathbf{W} = \mathbf{y}\mathbf{b}$ s.t. $\mathbf{yA} \geq \mathbf{c}$, $\mathbf{y} \geq \mathbf{0}$

- If \mathbf{x} and \mathbf{y} are feasible solution to P and D, respectively. Then \mathbf{x} and \mathbf{y} are optimal to their respective problems iff

$$(\mathbf{yA} - \mathbf{c})\mathbf{x} + \mathbf{y}(\mathbf{b} - \mathbf{Ax}) = 0$$

Complementary Slackness Theorem (CST)

Proof: $\mathbf{Ax} + \mathbf{u} = \mathbf{b}$ (i), $\mathbf{yA} - \mathbf{v} = \mathbf{c}$ (ii)

Where $\mathbf{u}, \mathbf{v} \geq 0$

$$\mathbf{u} = \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \vdots \\ \mathbf{u}_m \end{bmatrix} \quad \mathbf{v} = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \cdots \quad \mathbf{v}_n]$$

Pre-multiply (i) by \mathbf{y} and post-multiply (ii) by \mathbf{x} , we get

$$\mathbf{yAx} + \mathbf{yu} = \mathbf{yb} \quad (\text{iii}) \quad \text{and} \quad \mathbf{yAx} - \mathbf{vx} = \mathbf{cx} \quad (\text{iv})$$

Now, (iii) – (iv),

$$\mathbf{yu} + \mathbf{vx} = \mathbf{yb} - \mathbf{cx} \Rightarrow \mathbf{y(b-Ax)} + (\mathbf{yA-c})\mathbf{x} = \mathbf{yb} - \mathbf{cx}$$

and from the given condition $\mathbf{y(b-Ax)} + (\mathbf{yA-c})\mathbf{x} = \mathbf{0}$

$\Rightarrow \mathbf{yb} - \mathbf{cx} = \mathbf{0} \Rightarrow \mathbf{yb} = \mathbf{cx}$, which implies optimality of \mathbf{x} and \mathbf{y} (from SDT)

The proof is complete when the other side of statement is also proved, i.e. if \mathbf{x} and \mathbf{y} are optimal ($\Rightarrow \mathbf{yb} = \mathbf{cx}$) it must be $\mathbf{yu} + \mathbf{vx} = 0$, which is straightforward.

Implications of CST

(i) Complementary slackness conditions

Since $\mathbf{x}, \mathbf{u}, \mathbf{y}, \mathbf{v} \geq 0$, if the sum of nonnegative terms equals zero, then each term is zero.

$$\begin{aligned} \mathbf{y}\mathbf{u} = 0 & \Rightarrow \mathbf{y}(\mathbf{b} - \mathbf{A}\mathbf{x}) = 0 \\ \text{and} \quad \mathbf{v}\mathbf{x} = 0 & \Rightarrow (\mathbf{y}\mathbf{A} - \mathbf{c})\mathbf{x} = 0 \end{aligned}$$

(ii) Using duality interpretations and the condition

If $y_i > 0 \Rightarrow u_i = 0$

Thus, if another party is willing to +ve prices, then it must be the case that the resource must have been utilized fully. (Valuable resource)

If $u_i > 0 \Rightarrow y_i = 0$ (resource not used fully)

(iii) CST conditions can be used to find an optimal primal solution from an optimal dual solution, and vice-versa

Example

[P]

$\text{Min } W = 2y_1 + 3y_2 + 5y_3 + 2y_4 + 3y_5$	Dual variable	Surplus variable
$\text{Subject to } y_1 + y_2 + 2y_3 + y_4 + 3y_5 \geq 4$	x_1	v_1
$2y_1 - 2y_2 + 3y_3 + y_4 + y_5 \geq 3$	x_2	v_2
$\text{and } y_1, y_2, y_3, y_4, y_5 \geq 0$		

[D]

$\text{Max } Z = 4x_1 + 3x_2$	Slack variable	
$\text{Subject to } x_1 + 2x_2 \leq 2$	(1) u_1	Optimal Solution: $x_1 = 4/5, x_2 = 3/5 \text{ and } Z = 5$
$x_1 - 2x_2 \leq 3$	(2) u_2	
$2x_1 + 3x_2 \leq 5$	(3) u_3	
$x_1 + x_2 \leq 2$	(4) u_4	
$3x_1 + x_2 \leq 3$	(5) u_5	
$x_1, x_2 \geq 0$		

For which constraints, slack $> 0 \Rightarrow u_2, u_3, u_4 > 0$. From CST, primal variable y 's corresponding to these constraints $= 0 \Rightarrow y_2 = y_3 = y_4 = 0$

Also, from CST, since $x_1, x_2 > 0 \Rightarrow$ primal constraint should be satisfied as equality, i.e. $v_1 = v_2 = 0$

Now, [P] reduces to $y_1 + 3y_5 = 4$ and $2y_1 + y_5 = 3 \Rightarrow y_1 = 1, y_5 = 1$ and $W = 5$

Dual Simplex Method

- Primal Problem:

Maximize $Z = \mathbf{c}\mathbf{x}$

Subject to $\mathbf{A}\mathbf{x} \leq \mathbf{b}$

$\mathbf{x} \geq 0$

- Basis \mathbf{B} is primal feasible if and only if $\mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b} \geq 0$
- Suppose this basis \mathbf{B} is also feasible for the dual problem (called dual feasible), then

$$\mathbf{y} = \mathbf{c}_B \mathbf{B}^{-1} \geq 0$$

and $\mathbf{y}\mathbf{A} \geq \mathbf{c}$

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

$$\Rightarrow z_j - c_j \geq 0, \forall j$$

which are optimality conditions for the primal problem

- Thus, if a Basis \mathbf{B} is dual feasible, then it is also primal optimal
- Since, this is primal optimal, using SDT ($\mathbf{c}\mathbf{x} = \mathbf{y}\mathbf{b}$ as $\mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b}$ and $\mathbf{y} = \mathbf{c}_B \mathbf{B}^{-1}$) \Rightarrow it is also dual optimal
- Thus, the main crux in the LP is to find the basis \mathbf{B} which is both primal and dual feasible
 \rightarrow Optimality

- **(Primal) simplex method:** Moves from one primal feasible basis to another till it achieves dual feasibility, i.e.

$$z_j - c_j \geq 0 \quad \longrightarrow \quad \text{optimality}$$

- **Dual simplex method:** Starts with a dual feasible Basis and move towards primal feasibility
- Difference from primal simplex
 - Entering variable / leaving rule
 - Optimality

Algorithm: Dual Simplex

- 1) The problem should maximization type with \leq constraints
- 2) Updated objective function coefficients are $\geq 0 \Rightarrow$ Dual feasibility
- 3) At least one RHS is negative \Rightarrow Primal infeasibility
- 4) Leaving variable: The basic variable with most (-ve) RHS
- 5) Entering variable: Maximum ratio rule, that x_j for which the ratio

$$\left\{ \frac{z_j - c_j}{a_{jk}} : a_{jk} < 0 \right\} \text{ is maximum, where } x_k \text{ is leaving variable}$$

- 6) Perform elementary row operations as usual

Example: Dual simplex

- Min $Z = 3x_1 + 2x_2$
Subject to $3x_1 + x_2 \geq 3$
 $4x_1 + 3x_2 \geq 6$
 $x_1 + x_2 \leq 3$
 $x_1, x_2 \geq 0$

- Convert into **maximization** type with all the constraints \leq type

$$\text{Max } Z' = -Z = -3x_1 - 2x_2$$

Subject to

$$-3x_1 - x_2 \leq -3$$

$$-4x_1 - 3x_2 \leq -6$$

$$x_1 + x_2 \leq 3$$

$$x_1, x_2 \geq 0$$

$$\text{max } Z' + 3x_1 + 2x_2 = 0$$

- **Augmented form**

$$\text{Max } Z' = -Z = -3x_1 - 2x_2$$

Subject to

$$-3x_1 - x_2 + x_3 = -3$$

$$-4x_1 - 3x_2 + x_4 = -6$$

$$x_1 + x_2 + x_5 = 3$$

$$x_1, x_2, x_3, x_4, x_5 \geq 0$$

Solving using Dual simplex

	Basis	x₁	x₂	x₃	x₄	x₅	RHS
Iteration 0	x₃	-3	-1	1	0	0	-3
	x₄	-4	-3	0	1	0	-6
	x₅	1	1	0	0	1	3
	Z'	3	2	0	0	0	0
	Ratio	-3/4	-2/3	-	-	-	
Iteration 1	x₃	-5/3	0	1	-1/3	0	-1
	x₂	4/3	1	0	-1/3	0	2
	x₅	-1/3	0	0	1/3	1	1
	Z'	1/3	0	0	2/3	0	-4
	Ratio	-1/5	-	-	-2	--	
Iteration 2	x₁	1	0	-3/5	1/5	0	3/5
	x₂	0	1	4/5	-3/5	0	6/5
	x₅	0	0	-1/5	2/5	1	6/5
	Z'	0	0	1/5	3/5	0	-21/5

Dual feasible and optimal: $x_1 = 3/5, x_2 = 6/5, Z = -Z' = 21/5$

Graphical Interpretation

