

Linear Programming

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Lecture 5 Duality and Sentivity Analysis

Outline:

1. Dual linear program

2. Duality theory

3. Sensitivity analysis

4. Dual simplex method

● Economic interpretation of duality

Is there any special meaning of the dual variables?

What is a dual problem tring to do?

What's the role of the complementary slackness in decision making?

● Dual variables

Consider a nondegenerate linear program:

$$\begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \quad (\text{minimize total cost}) \\ \text{(P)} \quad \text{s.t.} & \mathbf{A} \mathbf{x} = \mathbf{b} \quad (\text{satisfy demands}) \\ & \mathbf{x} \geq \mathbf{0} \quad (\text{different services}) \end{array}$$

Assume that \mathbf{x}^* is a nondegenerate optimal bfs

$$\mathbf{x}^* = \begin{pmatrix} \mathbf{x}_B^* \\ \mathbf{0} \end{pmatrix} = \begin{pmatrix} \mathbf{B}^{-1} \mathbf{b} \\ \mathbf{0} \end{pmatrix}$$

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

Since $\mathbf{x}_B^* = \mathbf{B}^{-1} \mathbf{b} > \mathbf{0}$, we have $\mathbf{B}^{-1}(\mathbf{b} + \Delta \mathbf{b}) > \mathbf{0}$ when $\Delta \mathbf{b}$ is small enough!

$$\begin{array}{l} \text{Thus } \bar{\mathbf{x}}^* = \begin{pmatrix} \bar{\mathbf{x}}^* \\ \mathbf{0} \end{pmatrix} = \begin{pmatrix} \mathbf{B}^{-1}(\mathbf{b} + \Delta \mathbf{b}) \\ \mathbf{0} \end{pmatrix} \text{ is an optima bfs to} \\ \begin{array}{ll} \min & \mathbf{c}^T \mathbf{x} \\ \text{(\bar{P})} \quad \text{s.t.} & \mathbf{A} \mathbf{x} = \mathbf{b} + \Delta \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array} \end{array}$$

with $\bar{z}^* = \mathbf{c}_B^T \mathbf{B}^{-1}(\mathbf{b} + \Delta \mathbf{b})$ (Why? No change in r_q !)

● **Dual variable for shadow price**

Moreover, we have

$$\begin{aligned}\Delta z &= \bar{z}^* - z^* \\ &= c_B^T B^{-1}(b + \Delta b) - c_B^T B^{-1}b \\ &= c_B^T B^{-1} \Delta b \\ &= w^T b\end{aligned}$$

where

$w^T = c_B^T B^{-1}$ is the simplex multiplier for (P) at optimum!

Hence w_i is the “marginal price” of the i th demand.

Note: dual variable w_i indicates the minimum unit price that one has to charge for additional demand i . It is also called shadow price or equilibrium price.

● **Dual LP problem**

Consider the following production scenario:

n products to be produced: x_j = amount of product j , $j = 1, \dots, n$

m resources in hand: b_i = amount of resource i , $i = 1, \dots, m$

market selling price for each product is known: c_1, c_2, \dots, c_n

technology matrix is given by $[a_{ij}]$: each product j consumes a_{ij} units of resources i , $i = 1, \dots, m, j = 1, \dots, n$.

● **A manufacturer's view**

Maximize total sales $\sum_{j=1}^n c_j x_j$

Resource limitation

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n \leq b_i, \quad i = 1, \dots, m.$$

Production requirement $x_j \geq 0$, $j = 1, 2, \dots, n$.

$$\begin{aligned} \max \quad & c^T x \\ \text{(P)} \quad \text{s.t.} \quad & Ax \leq b \\ & x \geq 0 \end{aligned}$$

● **A wholesaler's view(supplier)**

m resources to purchase from a supplier:

$$w_i = \text{unit price to purchase resource } i, \quad i = 1, 2, \dots, m.$$

Free information market:

Supplier knows your selling price c_j for product x_j and he/she wants to get the most out of you, *i.e.*

$$a_{1j}w_1 + a_{2j}w_2 + \cdots + a_{mj}w_m \geq c_j \quad j = 1, \dots, n$$

Therefore, we have the dual

$$\begin{aligned} \min \quad & \mathbf{b}^T \mathbf{w} && \text{(minimize total spending)} \\ \text{(D)} \quad \text{s.t.} \quad & \mathbf{A}^T \mathbf{w} \geq \mathbf{c} && \text{(price accepted by the supplier)} \\ & \mathbf{w} \geq \mathbf{0} \end{aligned}$$

Summary:

$$\begin{array}{ll} \max & \mathbf{c}^T \mathbf{x} \\ \text{(P)} \quad \text{s.t.} & \mathbf{A} \mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array} \qquad \begin{array}{ll} \min & \mathbf{b}^T \mathbf{w} \\ \text{(D)} \quad \text{s.t.} & \mathbf{A}^T \mathbf{w} \geq \mathbf{c} \\ & \mathbf{w} \geq \mathbf{0} \end{array}$$

Q: how to transform (D) to (P) using the definition of dual LP?

A:

$$\begin{aligned} \min \quad & \mathbf{b}^T \mathbf{w} \\ \text{(D)} \quad \text{s.t.} \quad & \mathbf{A}^T \mathbf{w} \geq \mathbf{c} \\ & \mathbf{w} \geq \mathbf{0} \end{aligned} \quad \Longleftrightarrow \quad \begin{aligned} \min \quad & \mathbf{b}^T \mathbf{w} + \mathbf{0}^T \mathbf{r} \\ \text{(D)} \quad \text{s.t.} \quad & \mathbf{A}^T \mathbf{w} - \mathbf{I} \mathbf{r} \geq \mathbf{c} \\ & \mathbf{w}, \mathbf{r} \geq \mathbf{0} \end{aligned}$$

\Updownarrow

$$\begin{aligned} \min \quad & \mathbf{c}^T \mathbf{x} \\ \text{(P)} \quad \text{s.t.} \quad & \begin{bmatrix} \mathbf{A} \\ -\mathbf{I} \end{bmatrix} \mathbf{x} \leq \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix} \end{aligned} \quad \Leftarrow \quad \begin{aligned} \min \quad & \begin{bmatrix} \mathbf{b}^T & \mathbf{0}^T \end{bmatrix} \begin{bmatrix} \mathbf{w} \\ \mathbf{r} \end{bmatrix} \\ \text{(D)} \quad \text{s.t.} \quad & \begin{bmatrix} \mathbf{A}^T & -\mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{w} \\ \mathbf{r} \end{bmatrix} = \mathbf{c} \\ & \mathbf{w}, \mathbf{r} \geq \mathbf{0} \end{aligned}$$

\Updownarrow

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

● Complementary slackness condition

1. w_i^* is the maximum marginal price the manufacturer is willing to pay the supplier.
2. When resource i is not fully utilized, i.e., $a_{i1}x_1^* + a_{i2}x_2^* + \cdots + a_{in}x_n^* < b_i$, the complementary slackness condition implies $w_i^* = 0$. This means the manufacturer is not willing to pay a penny for buying any additional amount!

3. When the supplier asks too much, *i.e.*, $a_{1j}w_1 + a_{2j}w_2 + \dots + a_{mj}w_m > c_j$, then $x_j = 0$. This means the manufacturer is not going to produce any product j .

● **Additional explanation for complementary slackness**

Consider a relaxed knapsack problem (the decision variables not must be integers, thus it is a relaxed problem):

$$\begin{aligned} \max \quad & 3x_1 + 4x_2 + 9x_3 + 2x_4 + 5x_5 \\ \text{s.t.} \quad & 4x_1 + 7x_2 + 10x_3 + 3x_4 + 7x_5 \leq 20 \\ & x_j \geq 0, \quad j = 1, 2, 3, 4, 5. \end{aligned}$$

Its dual becomes

$$\begin{aligned} \min \quad & 20y \\ \text{s.t.} \quad & 4y \geq 3 \\ & 7y \geq 4 \\ & 10y \geq 9 \\ & 3y \geq 2 \\ & 7y \geq 5 \\ & y \geq 0 \end{aligned}$$

Obviously, $y^* = 0.9$.

Dual slack:

$$\begin{aligned} 4y^* - 3 = 0.6 & \rightarrow x_1^* = 0 \\ 7y^* - 4 = 2.3 & \rightarrow x_2^* = 0 \\ 10y^* - 9 = 0 & \rightarrow x_3^* = 2 \\ 3y^* - 2 = 0.7 & \rightarrow x_4^* = 0 \\ 7y^* - 5 = 1.3 & \rightarrow x_5^* = 0 \end{aligned}$$

Dual variable: $y^* = 0.9 > 0$

Primal slack: $20 - (4x_1^* + 7x_2^* + 10x_3^* + 3x_4^* + 7x_5^*) = 0$, $x_3^* = 2$, $z^* = 18$.