Duality

1 Basics

Consider the standard form of linear programming:

subject to
$$Ax = b$$
, (1b)

$$x \ge 0,$$
 (1c)

where \boldsymbol{e}^{\top} is an n dimensional row vector, $\boldsymbol{x}=(x_1,x_2,\ldots,x_n)^{\top}$ is an n dimensional column vector, \boldsymbol{A} is an $m\times n$ matrix, and \boldsymbol{b} is an n dimensional column vector. By using an m dimensional row vector $\boldsymbol{y}=(y_1,y_2,\ldots,y_m)^{\top}$, we define a linear programming problem:

$$\underset{\boldsymbol{y}}{\text{maximize}} \ \boldsymbol{y}^{\top} \boldsymbol{b} \tag{2a}$$

subject to
$$A^{\top} y \leq c$$
. (2b)

Due to the equality Ax = b, the dual variable y is free (no constraint).

If the primal problem (1) has inequalities as:

$$\underset{x}{\text{minimize}} c^{\top} x \tag{3a}$$

subject to
$$Ax \ge b$$
, (3b)

$$x > 0$$
, (3c)

we obtain the dual problem displayed below:

$$\begin{array}{c}
\text{maximize } \mathbf{y}^{\top} \mathbf{b} \\
\end{array} \tag{4a}$$

subject to
$$A^{\top} y \leq c$$
. (4b)

$$y \ge 0.$$
 (4c)

Due to the inequality $Ax \ge b$, the dual variable y is non-negative.

If the primal problem (1) has equalities and inequalities as:

$$\underset{\boldsymbol{x}_1, \boldsymbol{x}_2}{\text{minimize}} \ \boldsymbol{c}_1^{\top} \boldsymbol{x}_1 + \boldsymbol{c}_2^{\top} \boldsymbol{x}_2 \tag{5a}$$

subject to
$$A_{11}x_1 + A_{12}x_2 \ge b_1$$
, $(:y_1)$ (5b)

$$A_{21}x_1 + A_{22}x_2 = b_2, \ (: y_2)$$
 (5c)

$$x_1 \ge 0$$
, $(: \lambda_1)$, $x_2 \ge 0$, $(: \lambda_2)$, (5d)

we obtain the dual problem displayed below:

$$\underset{\boldsymbol{y}_1, \boldsymbol{y}_2}{\text{maximize}} \ \boldsymbol{y}_1^{\top} \boldsymbol{b}_1 + \boldsymbol{y}_2^{\top} \boldsymbol{b}_2 \tag{6a}$$

subject to
$$\boldsymbol{A}_{11}^{\top}\boldsymbol{y}_1 + \boldsymbol{A}_{21}^{\top}\boldsymbol{y}_2 \leq \boldsymbol{c}_1,$$
 (6b)

$$A_{12}^{\top} y_1 + A_{22}^{\top} y_2 \le c_2, \tag{6c}$$

$$y_1 \ge 0. \tag{6d}$$

Due to the inequality $A_{11}x_1 + A_{12}x_2 \ge b_1$, the dual variable y_1 is non-negative. Due to the equality $A_{21}x_1 + A_{22}x_2 = b_2$, the dual variable y_2 is free (no constraint).

The Lagrangian function of (5) is

$$L = c_1^{\mathsf{T}} x_1 + c_2^{\mathsf{T}} x_2 + y_1^{\mathsf{T}} (b_1 - A_{11} x_1 - A_{12} x_2) + y_2^{\mathsf{T}} (b_2 - A_{21} x_1 - A_{22} x_2) - \lambda_1 x_1 - \lambda_2 x_2$$
(7)

The KKT condition is:

$$c_1 - A_{11}^{\top} y_1 - A_{21}^{\top} y_2 - \lambda_1 = 0,$$
 (8a)

$$c_2 - A_{12}^{\mathsf{T}} y_1 - A_{22}^{\mathsf{T}} y_2 - \lambda_2 = 0,$$
 (8b)

$$0 \le A_{11}x_1 + A_{12}x_2 - b_1 \perp y_1 \ge 0, \tag{8c}$$

$$\mathbf{0} \le \boldsymbol{x}_1 \perp \boldsymbol{\lambda}_1 \ge 0, \tag{8d}$$

$$\mathbf{0} \le \mathbf{x}_2 \perp \mathbf{\lambda}_2 \ge 0, \tag{8e}$$

If both the primal and dual problems are feasible, and the KKT condition (8) is satisfied, the strong duality theorem shows:

$$\boldsymbol{c}_{1}^{\top}\boldsymbol{x}_{1} + \boldsymbol{c}_{2}^{\top}\boldsymbol{x}_{2} = \boldsymbol{y}_{1}^{\top}\boldsymbol{b}_{1} + \boldsymbol{y}_{2}^{\top}\boldsymbol{b}_{2} \tag{9}$$

The dual variables y_1 and y_2 of equation (6), namely, the shadow prices y_1 and y_2 are the Lagrange multipliers.

参考文献

[1] https://people.orie.cornell.edu/dpw/orie6300/Lectures/lec08.pdf