

# MATH 4211/6211 – Optimization

## Duality

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Every LP has a corresponding **dual** LP.

Consider the primal LP given by

$$\begin{array}{ll} \text{minimize} & \mathbf{c}^\top \mathbf{x} \\ \text{subject to} & \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array}$$

Then the dual LP is given by

$$\begin{array}{ll} \text{maximize} & \mathbf{b}^\top \boldsymbol{\lambda} \\ \text{subject to} & \mathbf{A}^\top \boldsymbol{\lambda} \leq \mathbf{c} \\ & \boldsymbol{\lambda} \geq \mathbf{0} \end{array}$$

where  $\boldsymbol{\lambda} \in \mathbb{R}^m$ .

Note the exchanges  $\mathbf{c} \leftrightarrow \mathbf{b}$ ,  $\mathbf{x} \leftrightarrow \boldsymbol{\lambda}$ ,  $\mathbf{A} \leftrightarrow \mathbf{A}^\top$ , and  $\geq \leftrightarrow \leq$ .

The dual of the dual LP is the primal LP: first we write the dual LP as

$$\begin{aligned} & \text{minimize} && (-\mathbf{b})^\top \boldsymbol{\lambda} \\ & \text{subject to} && -\mathbf{A}^\top \boldsymbol{\lambda} \geq -\mathbf{c} \\ & && \boldsymbol{\lambda} \geq \mathbf{0} \end{aligned}$$

Then the dual of the dual LP above is

$$\begin{aligned} & \text{maximize} && (-\mathbf{c})^\top \mathbf{x} \\ & \text{subject to} && -\mathbf{A}\mathbf{x} \leq -\mathbf{b} \\ & && \mathbf{x} \geq \mathbf{0} \end{aligned}$$

which is equivalent to the primal LP

$$\begin{aligned} & \text{minimize} && \mathbf{c}^\top \mathbf{x} \\ & \text{subject to} && \mathbf{A}\mathbf{x} \geq \mathbf{b} \\ & && \mathbf{x} \geq \mathbf{0} \end{aligned}$$

We can also consider the standard form of LP and its dual.

The standard form of LP is

$$\begin{aligned} & \text{minimize} && \mathbf{c}^\top \mathbf{x} \\ & \text{subject to} && \mathbf{Ax} = \mathbf{b} \\ & && \mathbf{x} \geq \mathbf{0} \end{aligned}$$

where the equality constraint  $\mathbf{Ax} = \mathbf{b}$  can be written as two inequality constraints:  $\mathbf{Ax} \leq \mathbf{b}$  (or equivalently  $-\mathbf{Ax} \geq -\mathbf{b}$ ) and  $\mathbf{Ax} \geq \mathbf{b}$ , i.e.,

$$\begin{bmatrix} \mathbf{A} \\ -\mathbf{A} \end{bmatrix} \mathbf{x} \geq \begin{bmatrix} \mathbf{b} \\ -\mathbf{b} \end{bmatrix}$$

Therefore the dual LP is

$$\begin{aligned} & \text{maximize} && [\mathbf{b}^\top, -\mathbf{b}^\top] \begin{bmatrix} \mathbf{u} \\ \mathbf{v} \end{bmatrix} \\ & \text{subject to} && [\mathbf{A}^\top, -\mathbf{A}^\top] \begin{bmatrix} \mathbf{u} \\ \mathbf{v} \end{bmatrix} \leq \mathbf{c} \\ & && \begin{bmatrix} \mathbf{u} \\ \mathbf{v} \end{bmatrix} \geq \mathbf{0} \end{aligned}$$

where  $\begin{bmatrix} \mathbf{u} \\ \mathbf{v} \end{bmatrix}$  is the variable.

Letting  $\boldsymbol{\lambda} = \mathbf{u} - \mathbf{v}$ , we obtain

$$\begin{aligned} & \text{maximize} && \mathbf{b}^\top \boldsymbol{\lambda} \\ & \text{subject to} && \mathbf{A}^\top \boldsymbol{\lambda} \leq \mathbf{c} \end{aligned}$$

Note that we do not have  $\boldsymbol{\lambda} \geq \mathbf{0}$  anymore.

## Symmetric form of duality

### Primal

$$\begin{aligned} &\text{minimize} && c^\top x \\ &\text{subject to} && Ax \geq b \\ &&& x \geq 0 \end{aligned}$$

### Dual

$$\begin{aligned} &\text{maximize} && b^\top \lambda \\ &\text{subject to} && A^\top \lambda \leq c \\ &&& \lambda \geq 0 \end{aligned}$$

## Asymmetric form of duality

### Primal

$$\begin{aligned} &\text{minimize} && c^\top x \\ &\text{subject to} && Ax = b \\ &&& x \geq 0 \end{aligned}$$

### Dual

$$\begin{aligned} &\text{maximize} && b^\top \lambda \\ &\text{subject to} && A^\top \lambda \leq c \end{aligned}$$

**Lemma (Weak Duality Lemma).** Suppose  $x$  and  $\lambda$  are feasible solutions to the primal and dual LP problems, respectively, then

$$c^\top x \geq b^\top \lambda$$

That is, the primal objective value  $\geq$  dual objective value.

**Proof.** We prove the asymmetric form only. Since  $x$  and  $\lambda$  are both feasible in their corresponding problems, we know  $Ax = b$  and  $x \geq 0$ , as well as  $A^\top \lambda \leq c$ . Hence

$$b^\top \lambda = (Ax)^\top \lambda = x^\top A^\top \lambda \leq x^\top c$$

**Theorem.** Suppose  $x_0$  and  $\lambda_0$  are feasible points to the primal and dual LP problems, respectively. If  $c^\top x_0 = b^\top \lambda_0$ , then  $x_0$  and  $\lambda_0$  are optimal solutions to their respective problems.

**Proof.** By the weak duality lemma above, we know for every primal feasible point  $x$  there is

$$c^\top x \geq b^\top \lambda_0 = c^\top x_0$$

Therefore  $x_0$  is optimal. Similarly we can show the optimality of  $\lambda_0$ .



**Theorem.** If the primal problem has an optimal solution, then so does the dual, and the optimal primal objective value is equal to the optimal dual objective value.

**Proof.** Consider the asymmetric form first. If the primal problem has an optimal solution, then by the fundamental theorem of LP, it has an optimal basic feasible solution. That is  $A = [B, D]$  (WLOG we assume the first  $m$  columns of  $A$  are basic columns) and  $x = [x_B; 0]$  is optimal.

As  $x$  is optimal, we set  $\lambda$  such that  $\lambda^\top = c_B^\top B^{-1}$ , and have

$$0 \leq r_D^\top = c_D^\top - (c_B^\top B^{-1})D = c_D^\top - \lambda^\top D$$

i.e.,  $\lambda^\top D \leq c_D^\top$ . So  $\lambda$  is feasible and optimal:

$$\begin{aligned}\lambda^\top A &= \lambda^\top [B, D] = c_B^\top B^{-1} [B, D] \leq [c_B^\top, c_D^\top] = c^\top \\ \lambda^\top b &= c_B^\top B^{-1} b = c_B^\top x_B = c^\top x.\end{aligned}$$

For symmetric form, convert it to the asymmetric form and apply the proof.

**Theorem (Complementary Slackness Condition).** The feasible solution  $x$  and  $\lambda$  are optimal to a dual pair of LPs iff  $(c - A^\top \lambda)^\top x = 0$  and  $\lambda^\top (Ax - b) = 0$ .

**Proof.** We consider the asymmetric form.

( $\Rightarrow$ ) Since  $x$  is feasible, there is  $Ax = b$  and hence  $\lambda^\top (Ax - b) = 0$ . In addition, since both  $x$  and  $\lambda$  are optimal, we have  $c^\top x = b^\top \lambda = (Ax)^\top \lambda$ , which implies the first equality.

( $\Leftarrow$ ) The two equalities imply  $c^\top x = \lambda^\top Ax = \lambda^\top b$ . Hence  $x$  and  $\lambda$  are optimal.