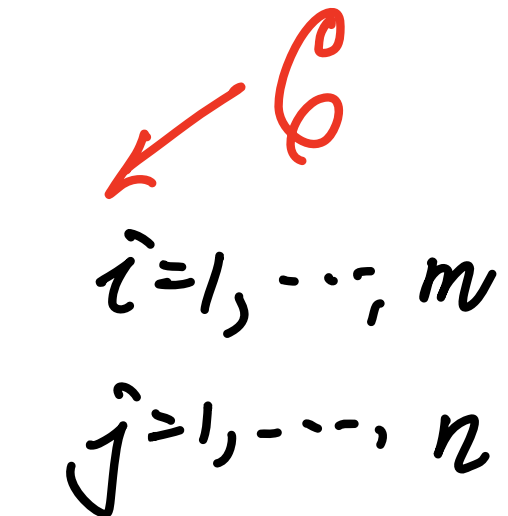



# Duality (General Form)

$$\begin{array}{ll} \text{maximize} & J = f_0(x) \\ \text{subject to} & f_i(x) \leq 0 \quad i=1, \dots, m \\ & g_j(x) = 0 \quad j=1, \dots, n \end{array}$$


$$\begin{array}{ll} \text{maximize} & J = f_0(x) \\ \text{subject to} & g_j(x) = 0 \quad j=1, \dots, n \end{array}$$


# Equality Constraints

$$\begin{aligned} & \text{maximize}_{x \in \mathbb{R}^d} \quad J = f_0(x) \\ & \text{subject to} \quad g_j(x) = 0 \quad j=1, \dots, n \end{aligned}$$

Lagrange Multiplier ( $\mu$ ) & Lagrangian ( $L$ )

$$\begin{aligned} & \text{maximize}_{x, \mu_1, \dots, \mu_n} \quad L(x, \mu) = f_0(x) - \sum_{j=1}^n \mu_j g_j(x) \end{aligned}$$

no constraints

# Equality Constraints

Compare  $\max f_0$  with constraints  
&  $\max L$  with no constraints

Let  $\tilde{x} \in \mathcal{C}$ , i.e.  $g_j(\tilde{x}) = 0$

$$\begin{aligned} \underline{f_0(\tilde{x})} &= f_0(\tilde{x}) - \sum_j \mu_j \underline{g_j(\tilde{x})} = 0 \\ &\leq \max_x \left[ f_0(x) - \sum_j \mu_j g_j(x) \right] \\ &= \max_x L(x, \mu) := \underline{\xi(\mu)} \end{aligned}$$

# Equality Constraints

Compare  $\max f_0$  with constraints  
&  $\max L$  with no constraints

Let  $\tilde{x} \in C$ , i.e.  $g_j(\tilde{x}) = 0$

$$J(\tilde{x}) = f_0(\tilde{x}) \leq \zeta(\mu)$$

any  $\tilde{x}$  satisfies constraint

no constraint for  $\mu$

$$J^* = \max_{\tilde{x} \in C} J(\tilde{x}) \leq \min_{\mu} \zeta(\mu) = \zeta^*$$

# Equality Constraints

Compare  $\max f_0$  with constraints  
&  $\max L$  with no constraints

Let  $\tilde{x} \in C$ , i.e.  $g_j(\tilde{x}) = 0$

$$J(\tilde{x}) = f_0(\tilde{x}) \leq J(\mu)$$

any  $\tilde{x}$  satisfies constraint

no constraint for  $\mu$

$$J^* \leq J^* \text{ (Weak Duality)}$$

# Equality Constraints

(P)  $\max_x \mathcal{J}(x) = f_0(x)$  6  
s.t.  $g_j(x) = 0 \quad j = 1, 2, \dots, n$

(D)  $\min_{\mu} \mathcal{Z}(\mu)$   $L(x, \mu)$

where  $\mathcal{Z}(\mu) = \max_x \left[ f_0(x) - \sum_j \mu_j g_j(x) \right]$   
no constraint.

Weak Duality  $\mathcal{J}^* = \max_{x \in \mathcal{C}} \mathcal{J}(x) \leq \min_{\mu} \mathcal{Z}(\mu) = \mathcal{Z}^*$

# Inequality Constraints

$$\begin{array}{ll} \underset{x \in \mathbb{R}^d}{\text{maximize}} & J = f_0(x) \\ \text{subject to} & f_i(x) \leq 0 \quad i=1, \dots, m \\ & g_j(x) = 0 \quad j=1, \dots, n \end{array}$$

$$L(x, \lambda, \mu) = f_0(x) - \sum_{i=1}^m \lambda_i f_i(x) - \sum_{j=1}^n \mu_j g_j(x)$$

$\lambda_i \geq 0$

Lagrangian fct

# Inequality Constraints

Let  $\tilde{x} \in \mathcal{C}$ , i.e. constraints are satisfied.  
 $\lambda \geq 0$

$$\underline{f_0(\tilde{x})} \leq f_0(x) - \sum_i \underbrace{\lambda_i}_{\geq 0} \underbrace{f_i(\tilde{x})}_{\leq 0} - \sum_j \underbrace{\mu_j}_{=0} \underbrace{g_j(\tilde{x})}_{=0}$$

$$= L(\tilde{x}, \lambda, \mu)$$

$$\leq \max_x L(x, \lambda, \mu) = \xi(\lambda, \mu)$$

(no constraint) ( $\lambda \geq 0$ )



# Inequality Constraints

Let  $\tilde{x} \in \mathcal{C}$ , i.e. constraints are satisfied.  
 $\lambda \geq 0$

$$\tilde{x} \in \mathcal{C} \quad f_0(\tilde{x}) \leq \xi(\lambda, \mu) \quad \lambda \geq 0$$

$$\max_{x \in \mathcal{C}} f_0(x) \leq \min_{\lambda \geq 0, \mu} \xi(\lambda, \mu)$$

$$J^* \leq \xi^*$$

Weak  
Duality

# Inequality Constraints

(P)

$$\max \quad \zeta(x) = f_0(x)$$

$$\text{s.t.} \quad f_i(x) \leq 0 \quad i=1, 2, \dots, m$$

$$g_j(x) = 0 \quad j=1, 2, \dots, n$$

$\mathcal{C}$

(D)

$$\min_{\lambda \geq 0, \mu} \quad \xi(\lambda, \mu)$$

constraint for  $\lambda$   $L(x, \lambda, \mu)$

$$\xi(\lambda, \mu) = \max_x \left[ f_0(x) - \sum_i \lambda_i f_i(x) - \sum_j \mu_j g_j(x) \right]$$

no constraint

Weak Duality:  $\zeta^* = \max_{x \in \mathcal{C}} \zeta(x) \leq \min_{\lambda \geq 0, \mu} \xi(\lambda, \mu) = \zeta^*$

# Duality for General LP

$$\max \zeta(x) = m^T X$$

$$\text{s.t.} \quad \left. \begin{array}{l} AX \leq \vec{b} \\ BX = \vec{c} \end{array} \right\} \delta$$

$$L(x, \lambda, \mu) = m^T X - \underbrace{\lambda^T}_{\lambda \geq 0, \mathcal{D}} (AX - b) - \mu^T (BX - c)$$

$$\zeta(\lambda, \mu) = \max_{X \text{ no constraint}} L(x, \lambda, \mu)$$

# Duality for General LP

$$\max J(x) = m^T x$$

$$\text{s.t.} \quad \left. \begin{array}{l} Ax \leq \vec{b} \\ Bx = \vec{c} \end{array} \right\} \mathcal{G}$$

$$L(x, \lambda, \mu) = m^T x - \underbrace{\lambda^T (Ax - b)}_{\lambda \geq 0, \mathcal{D}} - \mu^T (Bx - c)$$

$$x \in \mathcal{G} \quad \boxed{J(x) \leq J(\lambda, \mu)} \quad \lambda \geq 0$$

# Duality for General LP

$$\max J(x) = m^T x$$

$$\text{s.t.} \quad \left. \begin{array}{l} Ax \leq \vec{b} \\ Bx = \vec{c} \end{array} \right\} \mathcal{C}$$

$$L(x, \lambda, \mu) = m^T x - \underbrace{\lambda^T (Ax - b)}_{\lambda \geq 0, \mathcal{D}} - \mu^T (Bx - c)$$

$$x \in \mathcal{C} \quad \boxed{J(x) \leq J(\lambda, \mu)} \quad \lambda \geq 0$$

= ?

# Duality for General LP

$$L(x, \lambda, \mu) = m^T x - \lambda^T (Ax - b) - \mu^T (Bx - c)$$

$\lambda \geq 0, \mathcal{D}$

$$= (m^T - \lambda^T A - \mu^T B) x + \lambda^T b + \mu^T c$$

no constraint

If  $m^T - \lambda^T A - \mu^T B \neq 0$

$$\max_x L(x, \lambda, \mu) = +\infty$$

# Duality for General LP

$$L(x, \lambda, \mu) = m^T x - \lambda^T (Ax - b) - \mu^T (Bx - c)$$

$\lambda \geq 0, \mathcal{D}$

$$= (m^T - \lambda^T A - \mu^T B) x + \lambda^T b + \mu^T c$$

no constraint

If  $m^T - \lambda^T A - \mu^T B = 0$

$$\max_x L(x, \lambda, \mu) = \lambda^T b + \mu^T c$$

# Duality for General LP

$$L(x, \lambda, \mu) = m^T x - \lambda^T (Ax - b) - \mu^T (Bx - c)$$

$\lambda \geq 0, \mathcal{D}$

$$= (m^T - \lambda^T A - \mu^T B) x + \lambda^T b + \mu^T c$$

Hence

no constraint

$$\Sigma(\lambda, \mu) = \begin{cases} \lambda^T b + \mu^T c & \text{if } \lambda^T A + \mu^T B = m^T \\ +\infty & \text{if } \lambda^T A + \mu^T B \neq m^T \end{cases}$$



# Duality for General LP

$$L(x, \lambda, \mu) = m^T x - \lambda^T (Ax - b) - \mu^T (Bx - c)$$

$\lambda \geq 0, \mathcal{D}$

$$= (m^T - \lambda^T A - \mu^T B) x + \lambda^T b + \mu^T c$$

Hence

no constraint

$$\Sigma(\lambda, \mu) = \begin{cases} \lambda^T b + \mu^T c & \text{if } A^T \lambda + B^T \mu = m \\ +\infty & \text{if } A^T \lambda + B^T \mu \neq m \end{cases}$$

# Duality for General LP

$$\begin{aligned} \text{(P)} \quad & \max_X \quad J(X) = m^T X \\ & \text{s.t.} \quad AX \leq b \\ & \quad \quad BX = c \end{aligned}$$

$$\begin{aligned} \text{(D)} \quad & \min_{\lambda \geq 0, \mu} \quad \Sigma(\lambda, \mu) = b^T \lambda + c^T \mu \\ & \text{s.t.} \quad A^T \lambda + B^T \mu = m \end{aligned}$$

# Duality for "Traditional" LP

$$(P) \quad \max J(x) = c^T x$$

$$\text{s.t.} \quad Ax \leq b$$

$$x \geq 0 \iff -x \leq 0$$

$$L(x, y, \lambda) = c^T x - y^T (Ax - b) - \lambda^T (-x)$$

$$y, \lambda \geq 0$$

$$= (c^T - y^T A + \lambda^T) x + y^T b$$

# Duality for "Traditional" LP

$$L(x, y, \lambda) = c^T x - y^T (Ax - b) - \lambda^T (-x)$$

$y, \lambda \geq 0$

$$= (c^T - y^T A + \lambda^T) x + y^T b$$

$$\max_{x \geq 0} L(x, y, \lambda) = \begin{cases} y^T b & \text{if } c^T - y^T A + \lambda^T = 0 \\ +\infty & \text{if } c^T - y^T A + \lambda^T \neq 0 \end{cases}$$

$x \geq 0$   
no constraint

# Duality for "Traditional" LP

$$L(x, y, \lambda) = c^T x - y^T (Ax - b) - \lambda^T (-x)$$

$$y, \lambda \geq 0$$

$$= (c^T - y^T A + \lambda^T) x + y^T b$$

$$\max_{x, y, \lambda} L(x, y, \lambda) = \begin{cases} y^T b & \text{if } c - A^T y + \lambda = 0 \\ +\infty & \text{if } c - A^T y + \lambda \neq 0 \end{cases}$$

$x$   
no constraint

# Duality for "Traditional" LP

$$\begin{aligned} \text{(D)} \quad & \min_{\gamma} \quad \xi(\gamma) = b^T \gamma \\ & \text{s.t.} \quad c - A^T \gamma + \lambda = 0, \quad \gamma, \lambda \geq 0 \end{aligned}$$

$$\Leftrightarrow A^T \gamma = c + \lambda, \quad \gamma, \lambda \geq 0$$

$$\Leftrightarrow A^T \gamma \geq c \quad \gamma \geq 0$$

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