

Lecture 8

Primal

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax \leq b \\ & \downarrow \\ & m \times n \end{aligned}$$

$$x \in \mathbb{R}^n, \quad x^*, p^*$$

Dual

$$\begin{aligned} \max \quad & -b^T \lambda \\ \text{s.t.} \quad & A^T \lambda + c = 0 \\ & \downarrow \\ & n \times m \\ & \lambda \geq 0 \end{aligned}$$

$$\lambda \in \mathbb{R}^m, \quad \lambda^*, q^* \\ c^T = -\lambda^T A$$

Observations

of constraints in primal = # of variables in dual.

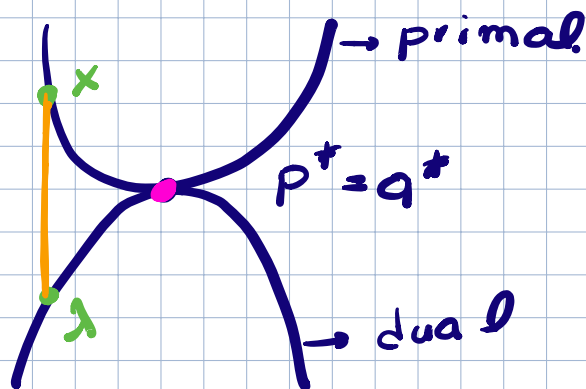
of constraints in dual = # of variables in the primal.

Weak duality

Strong duality

Weak duality says:

if x is feasible in primal
and λ is feasible in dual



$$\text{Duality gap: } c^T x + \lambda^T b \geq 0$$

$$c^T x \geq -\lambda^T b$$

Proof: We know that $c^T = -\lambda^T A$ since λ is feasible.

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

because x feasible $Ax \leq b \Rightarrow -Ax \geq -b$
and $\lambda \geq 0$

Weak duality implies

$$c^T x \geq p^* \geq q^* \geq -\lambda^T b$$

$$p^* \geq q^*$$

Strong duality: provided that either the primal or the dual are feasible, we have that $q^* = p^*$

- Both feasible, $p^* = q^*$ take a finite value.
- If primal is feasible, $p^* = -\infty$, dual infeasible.
- If dual is feasible & $q^* = +\infty$, primal infeasible.

(Proof → we need theorem of alternatives)

<p style="text-align: center; color: orange;">Primal</p> $\begin{aligned} \min \quad & c^T x \\ \text{st} \quad & Ax \leq b \\ & \downarrow \\ & m \times n \end{aligned}$ <p style="text-align: center;">$x \in \mathbb{R}^n, x^*, p^*$</p>	\longleftrightarrow	<p style="text-align: center; color: orange;">Dual</p> $\begin{aligned} \max \quad & -b^T \lambda \\ \text{st} \quad & A^T \lambda + c = 0 \\ & \downarrow \\ & n \times m \\ & \lambda \geq 0 \end{aligned}$ <p style="text-align: center;">$\lambda \in \mathbb{R}^m, \lambda^*, q^*$</p>
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- At the optimal x^* and λ^* , the following 3 conditions hold:

$\begin{aligned} \textcircled{1} \quad & x^* \text{ is feasible, } Ax^* \leq b \\ \textcircled{2} \quad & \lambda^* \text{ is feasible, } c + A^T \lambda^* = 0, \lambda^* \geq 0 \\ \textcircled{3} \quad & c^T x^* + b^T \lambda^* = 0 \end{aligned}$	}	certificate of optimality
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We can rewrite as follows:

$$\left. \begin{aligned} c^T x^* + b^T \lambda^* &= 0 \\ c^T &= -\lambda^{*T} A \end{aligned} \right\} \Rightarrow (-\lambda^{*T} A) x^* + \overbrace{b^T \lambda^*}^{\lambda^{*T} b} = 0$$

$$\Rightarrow \lambda^{*T} (A x^* - b) = 0$$

\downarrow
 $(\lambda_1^*, \lambda_2^*, \dots, \lambda_m^*)$

≥ 0

$$\begin{pmatrix} a_1^T x^* - b_1 \\ a_2^T x^* - b_2 \\ \vdots \\ a_m^T x^* - b_m \end{pmatrix} = 0$$

$$\Rightarrow \boxed{\lambda_i^* (a_i^T x^* - b_i) = 0} \text{ for all } i = 1, \dots, m$$

complementary slackness

$$\text{if } \lambda_i^* > 0 \Rightarrow a_i^T x^* - b_i = 0$$

$$\text{if } a_i^T x^* - b_i < 0 \Rightarrow \lambda_i^* = 0$$

Structure of the optimal solution.

x^* meets $J(x^*)$ constraints with equality: these exactly the indices where λ^* can take nonzero values

Second way to check optimality

Find x^* feasible and λ^* feasible such that $\lambda_i^* = 0$ when constraint i is not active for x^* .

Example.

$$\begin{aligned} \min_{x_1, x_2} & (-4 \ -5) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \\ & \begin{matrix} c_1 \\ c_2 \\ c_3 \\ c_4 \end{matrix} \begin{pmatrix} -1 & 0 \\ 2 & 1 \\ 0 & -1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \leq \begin{pmatrix} 0 \\ 3 \\ 0 \\ 3 \end{pmatrix} \end{aligned}$$

Show that $x_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is optimal.

x_0 is feasible ✓ it satisfies c_2 and c_4 with equality.
 \Rightarrow corresponding dual variables need to be of the form

$$\lambda = \begin{pmatrix} 0 \\ \lambda_2 \\ 0 \\ \lambda_4 \end{pmatrix}, \quad \lambda_2, \lambda_4 \geq 0$$

λ needs to be feasible $\Rightarrow A^T \lambda = -c$

$$\Rightarrow \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} \lambda_2 \\ \lambda_4 \end{pmatrix} = \begin{pmatrix} 4 \\ 5 \end{pmatrix} \Rightarrow \begin{pmatrix} \lambda_2 \\ \lambda_4 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

The vector $\lambda = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 2 \end{pmatrix}$ is feasible in dual
with correct sparseness

This proves the optimality of x_0 .