5.2 Simplex and Duality

The following linear programs form a primal dual pair:

$$z = \max\{c^T x \mid Ax = b, x \ge 0\}$$
$$w = \min\{b^T y \mid A^T y \ge c\}$$

This means for computing the dual of a standard form LP, we do not have non-negativity constraints for the dual variables.



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Proof of Optimality Criterion for Simplex

Suppose that we have a basic feasible solution with reduced cost

$$\tilde{c} = c^T - c_R^T A_R^{-1} A \le 0$$

This is equivalent to $A^T(A_B^{-1})^T c_B \ge c$

 $y^* = (A_B^{-1})^T c_B$ is solution to the dual $\min\{b^T y | A^T y \ge c\}$.

$$b^{T}y^{*} = (Ax^{*})^{T}y^{*} = (A_{B}x_{B}^{*})^{T}y^{*}$$
$$= (A_{B}x_{B}^{*})^{T}(A_{B}^{-1})^{T}c_{B} = (x_{B}^{*})^{T}A_{B}^{T}(A_{B}^{-1})^{T}c_{B}$$
$$= c^{T}x^{*}$$

Hence, the solution is optimal.

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Proof

Primal:

$$\max\{c^{T}x \mid Ax = b, x \ge 0\}$$

$$= \max\{c^{T}x \mid Ax \le b, -Ax \le -b, x \ge 0\}$$

$$= \max\{c^{T}x \mid \begin{bmatrix} A \\ -A \end{bmatrix} x \le \begin{bmatrix} b \\ -b \end{bmatrix}, x \ge 0\}$$

Dual:

$$\min\{ [b^{T} - b^{T}]y \mid [A^{T} - A^{T}]y \geq c, y \geq 0 \}$$

$$= \min \left\{ [b^{T} - b^{T}] \cdot \begin{bmatrix} y^{+} \\ y^{-} \end{bmatrix} \middle| [A^{T} - A^{T}] \cdot \begin{bmatrix} y^{+} \\ y^{-} \end{bmatrix} \geq c, y^{-} \geq 0, y^{+} \geq 0 \right\}$$

$$= \min \left\{ b^{T} \cdot (y^{+} - y^{-}) \middle| A^{T} \cdot (y^{+} - y^{-}) \geq c, y^{-} \geq 0, y^{+} \geq 0 \right\}$$

$$= \min \left\{ b^{T}y' \middle| A^{T}y' \geq c \right\}$$



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