## 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 $\tilde{c} = c^T - c_B^T A_B^{-1} A \le 0$ 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:

$$\min\{[b^{T} - b^{T}]y \mid [A^{T} - A^{T}]y \ge c, y \ge 0\}$$
  
= 
$$\min\{[b^{T} - b^{T}] \cdot \begin{bmatrix} y^{+} \\ y^{-} \end{bmatrix} \mid [A^{T} - A^{T}] \cdot \begin{bmatrix} y^{+} \\ y^{-} \end{bmatrix} \ge c, y^{-} \ge 0, y^{+} \ge 0\}$$
  
= 
$$\min\{b^{T} \cdot (y^{+} - y^{-}) \mid A^{T} \cdot (y^{+} - y^{-}) \ge c, y^{-} \ge 0, y^{+} \ge 0\}$$
  
= 
$$\min\{b^{T}y' \mid A^{T}y' \ge c\}$$

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