

## Discrete Optimization: Homework #7, Ex. #3

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We consider the following LP :  $\max\{c^T x : Ax \leq b, b \in \mathbb{R}^n\}$  and assume that it is feasible and bounded. Thus, we know that there exists an optimal solution, say  $x^*$ . We also know that the dual of our LP :  $\min\{b^T y : A^T y = c, y \geq 0\}$  is feasible and bounded by the Strong Duality theorem. That's why, we have that, for  $y^*$  an optimal solution of the dual LP :  $c^T x^* = b^T y^*$  i.e. the optimal values coincide.

But  $x^*$  is optimal for the primal LP and  $y^*$  for the dual LP. So for all  $(x^*, y^*)$  satisfying the previous equation and the respective constraints of primal and dual LP,  $(x^*, y^*)$  will be optimal. In other words, this means that all  $(x, y)$  in the following Polyhedron P are optimal.

$$c^T x = b^T y$$

$$Ax \leq b$$

$$A^T y = c$$

$$y \geq 0$$

$$\text{We have } P = \{\tilde{x} \in \mathbb{R}^n : \tilde{A}\tilde{x} \leq \tilde{b}\} \text{ with } \tilde{A} = \begin{pmatrix} c^T & -b^T \\ -c^T & b^T \\ A & 0 \\ 0 & A^T \\ 0 & -A^T \\ 0 & -I \end{pmatrix}, \tilde{x} = \begin{pmatrix} x \\ y \end{pmatrix} \text{ and}$$

$$\tilde{b} = \begin{pmatrix} 0 \\ 0 \\ b \\ c \\ -c \\ 0 \end{pmatrix}.$$

Since we know that there exists the optimal solution  $(x^*, y^*)$  of the LP, the vector  $w = \begin{pmatrix} x^* \\ y^* \end{pmatrix}$  is in P. So using the oracle algorithm on the Polyhedron P in a single call, we find an optimal solution of our LP.