

Lecture 21 Mar 29 2021

From last week: 1) $\max: c^T \vec{x}$

$$\text{s.t.: } A\vec{x} \leq b, \quad 0 \leq \vec{x}$$

$$2) \min b^T \vec{y}$$

$$\text{s.t.: } A^T \vec{y} \geq c, \quad 0 \leq \vec{y}$$

Weak Duality Thm //

$$\begin{aligned} c^T x &\leq y^T A x \leq b^T y \\ \text{"max"} &\leq \text{"min"} \end{aligned}$$

Strong Duality Thm //

If either 1) or 2) has finite optimal value, the
so does the other.

Their values coincide (equal).

Homework



Lemma 1/

Let $A \in \mathbb{R}^{m \times n}$. Only one has solution.

1) $\{ \vec{x} \mid A\vec{x} = \vec{b} \}$

2) $\{ \vec{\lambda} \mid A^T \vec{\lambda} = 0 \quad \vec{\lambda}^T \vec{b} < 0 \}$

Proof/

Suppose 1) has optimal value $|p^*| < \infty$, we would be done if we can show

3) $\rightarrow \begin{bmatrix} y^T b \geq p^* \\ A^T y \leq c \end{bmatrix}$ is feasible.

$\exists y$ s.t. $y^T b \leq p^*$, $A^T y \leq c$ by weak duality thm.
by proving 3), we can have $y^T b = p^*$

Rewrite 3) = $\begin{bmatrix} A^T \\ -b^T \end{bmatrix} y^T \leq \begin{bmatrix} c \\ -p^* \end{bmatrix}$

Prove by contradiction, assume 3) = \emptyset .

By lemma $\exists \lambda = \begin{bmatrix} \vec{\lambda} \\ \lambda \end{bmatrix} \geq 0$

$$\begin{aligned} \text{Then } \tilde{\lambda}^T A^T - \lambda_0 b^T &= 0 \\ \tilde{\lambda}^T c - \lambda_0 p^* &< 0 \end{aligned}$$

4)

show 4) DNE

$$\Rightarrow \begin{aligned} \tilde{\lambda}^T A^T &= \lambda_0 b^T \\ \tilde{\lambda}^T c &< \lambda_0 p^* \end{aligned}$$

Case 1: $\lambda_0 = 0$

$$\tilde{\lambda}^T A^T = 0 \Rightarrow A^T \tilde{\lambda} = 0$$

$$\tilde{\lambda}^T c < 0 \Rightarrow c^T \tilde{\lambda} < 0$$

Since p^* is a finite optimal, $\exists x^*$ that is a optimal value of 1).

Define $x = x^* + \tilde{\lambda}$, $x \geq 0$.

$$\begin{aligned} Ax &= Ax^* + A\tilde{\lambda} \\ &= Ax^* = b \end{aligned}$$

$$\begin{aligned} c^T x &= c^T (x^* + \tilde{\lambda}) \\ &= c^T x^* + c^T \tilde{\lambda} \\ &= p^* + c^T \tilde{\lambda} \\ &< 0 \end{aligned}$$

$< p^*$ which is a contradiction since

we assumed p^* is minimum optimal value.

case 2 $\lambda_0 > 0$

$$\text{let } x = \frac{\tilde{x}}{\lambda_0}$$

$$\text{Then } Ax = A\left(\frac{\tilde{x}}{\lambda_0}\right) = \frac{1}{\lambda_0} A\tilde{x}$$

$$= \frac{1}{\lambda_0} (\lambda_0 \cdot b^T)$$

$$= b^T$$

$$c^T x = c^T \left(\frac{\tilde{x}}{\lambda_0}\right)$$

$$< \frac{1}{\lambda_0} \lambda_0 p^*$$

$$= p^*$$

which is another contradiction.

Thus 4) is empty and 3) must be feasible.

