

CONVEX OPTIMISATION TUTORIAL 12

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(a)

Given $f(x) = \sup_{c \in \mathcal{C}} \bar{c}^T \bar{x}$. Clearly $\bar{c}^T \bar{x}$ is an affine function and it is convex. Operation \sup will preserve the convexity of the function. Thus f is convex.

(b)

Given primal (P),

$$\begin{aligned} \max \quad & \bar{c}^T \bar{x} \\ \text{subject to} \quad & \mathbf{F} \bar{c} \leq \bar{g} \end{aligned} \equiv \begin{aligned} \min \quad & -\bar{c}^T \bar{x} \\ \text{subject to} \quad & \mathbf{F} \bar{c} \leq \bar{g} \end{aligned} \equiv f(\bar{x})$$

Now we find the dual of this problem by taking infimum of Lagrange function over \bar{x} .

$$\begin{aligned} \mathcal{L}(\bar{x}, \bar{\lambda}) &= -\bar{c}^T \bar{x} + \bar{\lambda}^T (\mathbf{F} \bar{c} - \bar{g}) \\ g(\bar{\lambda}) &= \inf_{\bar{x}} \mathcal{L}(\bar{x}, \bar{\lambda}) \end{aligned}$$

Therefore the dual problem (D) is,

$$\begin{aligned} \min \quad & \bar{\lambda}^T \bar{g} \\ \text{subject to} \quad & \mathbf{F}^T \bar{\lambda} = \bar{c} \\ & \bar{\lambda} \geq 0 \end{aligned}$$

$f(\bar{x})$ is the optimal of (P) and dual gap for LPs is zero. Thus $f(\bar{x})$ is the optimal for (D) also.

(c)

We can replace the objective in the *Robust LP* with the dual problem.

$$\begin{aligned} \min_{\bar{x}} \min_{\bar{\lambda}} \quad & \bar{\lambda}^T \bar{g} \\ \text{subject to} \quad & \mathbf{A} \bar{x} \geq \bar{b} \\ & \mathbf{F}^T \bar{\lambda} = \bar{c} \\ & \bar{\lambda} \geq 0 \end{aligned}$$

Clearly both \bar{x} and $\bar{\lambda}$ thus the above problem can be written as,

$$\begin{aligned} \min_{\bar{x}, \bar{\lambda}} \quad & \bar{\lambda}^T \bar{g} \\ \text{subject to} \quad & \mathbf{A}\bar{x} \geq \bar{b} \\ & \mathbf{F}^T \bar{\lambda} = \bar{x} \\ & \lambda \geq 0 \end{aligned}$$

Initial problem was not convex because both \bar{x} and \bar{c} are variable and $\bar{c}^T \bar{x}$ is not convex. But once we transform the subproblem to dual we get a LP.