

## Problem 1

(a)

With the rank method we have:

$$J = \{1, 2, 3, 4\}$$

and corresponding  $A_J$  as:

$$A_J = \begin{bmatrix} -1 & -6 & 1 & 3 \\ -1 & -2 & 7 & 1 \\ 0 & 3 & -10 & -1 \\ -6 & -11 & -2 & 12 \end{bmatrix}$$

$A_J$  is full rank and therefore  $\tilde{x}$  is an extreme point of  $P$

(b)

To find the vector  $c$ , notice that  $\tilde{x}$  is the unique solution to linear equation  $A_J x = b$  since  $N(A_J) = \{0\}$ . Therefore we have  $\forall x \in P, A_J x \leq b$ . If we set  $c$  as any non-positive combination of the rows in  $A_J$  we would have:

$$c = -A_J^T * d$$

where  $d \in \mathbf{R}_+^4$ . One example is to set  $d = [1, 1, 1, 1]$ , and we would have  $c^T x \geq -13$ , and only when  $x = \tilde{x}$  we have  $c^T \tilde{x} = -13$ .

## Problem 2

By using the same rank method we have:

$$J = \{2, 3, 4\}$$

and corresponding  $A_J$  as:

$$A_J = \begin{bmatrix} -4 & -2 & -2 & -9 \\ -8 & -2 & 0 & -5 \\ 0 & -6 & -7 & -4 \end{bmatrix}$$

$R\left(\begin{bmatrix} A_J \\ C \end{bmatrix}\right) = 4$  and therefore  $\hat{x}$  is a extreme point of  $P$ .

### Problem 3

(a)

The equivalent could be written as:

$$\begin{aligned} & \text{minimize} && \mathbf{1}^T y \\ & \text{subject to} && -y_k \mathbf{1} \leq A_k^T x - b_k \leq y_k \mathbf{1}, \quad k = 1, \dots, m, \end{aligned}$$

with  $y \in \mathbf{R}^m$  as our auxiliary variable.

(b)

The Lagrangian could be formed as follow:

$$\begin{aligned} L(x, y, \lambda, \mu) &= \sum_{i=1}^m [\mathbf{1}^T y_k + \lambda_k^T (A_k^T x - b_k - y_k \mathbf{1}) + \mu_k^T (-y_k \mathbf{1} - A_k^T x + b_k)] \\ &= \sum_{i=1}^m \{(\mathbf{1} - \lambda_k - \mu_k)^T y_k + [A_k^T (\lambda_k - \mu_k)]^T x - (\lambda_k - \mu_k)^T (b_k)\} \end{aligned}$$

Therefore the Lagrangian dual function:

$$\begin{aligned} g(\lambda, \mu) &= \inf_{x, y} L(x, y, \lambda, \mu) \\ &= \inf_{x, y} \left\{ \sum_{i=1}^m [(\mathbf{1} - \lambda_k - \mu_k)^T y_k + [A_k^T (\lambda_k - \mu_k)]^T x - (\lambda_k - \mu_k)^T (b_k)] \right\} \\ &= \begin{cases} \sum_{i=1}^m b_k^T (\lambda_k - \mu_k) & \mathbf{1} - \lambda_k - \mu_k = 0 \text{ \& } A_k^T (\lambda_k - \mu_k) = 0 \\ -\infty & \text{otherwise.} \end{cases} \end{aligned}$$

And the dual problem and therefore be formed as:

$$\begin{aligned} & \text{maximize} && \sum_{i=1}^m b_k^T (\lambda_k - \mu_k) \\ & \text{subject to} && \mathbf{1}^T (\lambda_k + \mu_k) = 1 \\ & && \sum_{i=1}^m A_k^T (\lambda_k - \mu_k) = 0 \\ & && \lambda_k \geq 0, \quad \mu_k \geq 0, \quad k = 1, \dots, m. \end{aligned}$$

To have the simplified form we only have to set  $z_k = \lambda_k - \mu_k$ .

$$\begin{aligned} & \text{maximize} && \sum_{i=1}^m b_k^T z_k \\ & \text{subject to} && \sum_{i=1}^m A_k^T z_k = 0 \\ & && \|z_k\| \leq 1 \quad k = 1, \dots, m. \end{aligned}$$

(c)

For least square solution  $x_{ls}$ , it satisfies normal equation, i.e.,

$$A^T A x_{ls} = A^T b$$

or in our case:

$$\sum_{i=1}^m A_k^T (A_k x_{ls} - b_k) = 0$$

i.e.,

$$\sum_{i=1}^m A_k^T r_{ls} = 0$$

Therefore  $r_{ls}$  satisfy the first constraint of the dual problem. To find out a lower bound of the primal problem we need a feasible value of the dual problem and therefore we set:

$$z_k = -\frac{r_k}{\max_{k=1, \dots, m} \|r_k\|_1}$$

Obviously  $\|z_k\|_1 \leq 1$ , Therefore we have:

$$\sum_{i=1}^m b_k^T z_k \leq \sum_{i=1}^m (A_k x_{ls} - b_k) z_k = \frac{r_k^2}{\max_{k=1, \dots, m} \|r_k\|_1}$$

## Problem 4

(a)

The standard LP:

$$\begin{aligned} & \text{minimize} && y \\ & \text{subject to} && -y\mathbf{1} \leq x - x_0 \leq y\mathbf{1} \\ & && A^T x \leq b \end{aligned}$$

(b)

The Lagrangian could be formed as follow:

$$\begin{aligned} L(x, y, \lambda, \mu, \omega) &= y + \lambda^T(x - x_0 - y\mathbf{1}) + \mu^T(\lambda - \mu + A^T\omega) + \omega^T(A^T x - b) \\ &= (1 - \mathbf{1}^T \lambda - \mathbf{1}^T \mu)y + (\lambda - \mu + A^T\omega)^T x + (\mu - \lambda)^T x_0 - b^T \omega \end{aligned}$$

Therefore the Lagrangian dual function:

$$\begin{aligned} g(\lambda, \mu, \omega) &= \inf_{x, y} L(x, y, \lambda, \mu, \omega) \\ &= \inf_{x, y} \{[(1 - \mathbf{1}^T \lambda - \mathbf{1}^T \mu)y + (\lambda - \mu + A^T\omega)^T x + (\mu - \lambda)^T x_0 - b^T \omega]\} \\ &= \begin{cases} (\mu - \lambda)^T x_0 - b^T \omega & \mathbf{1} - \lambda_k - \mu_k = 0 \text{ \& } \lambda - \mu + A^T\omega = 0 \\ -\infty & \text{otherwise.} \end{cases} \end{aligned}$$

Then the dual problem could be formed as follow:

$$\begin{aligned} \text{maximize} \quad & (\mu - \lambda)^T x_0 - b^T \omega \\ \text{subject to} \quad & \mathbf{1}^T(\lambda_k + \mu_k) = 1 \\ & A^T\omega = \mu - \lambda \\ & \lambda \geq 0, \quad \mu \geq 0, \quad \omega \geq 0 \end{aligned}$$

Using the second constraint  $A^T\omega = \mu - \lambda$  and a simplified version could be formed as follow:

$$\begin{aligned} \text{maximize} \quad & (A^T\omega - b)^T x_0 \\ \text{subject to} \quad & \|A^T\omega\|_1 \leq 1 \\ & \omega \geq 0 \end{aligned}$$