

## 4.3

Let  $f_0(x) = (1/2)x^T P x + q^T x + r$ , which is quadratic and differentiable. The optimality condition for differentiable  $f_0$  is  $\nabla f_0(x^*)^T(y - x^*) \geq 0$  for all feasible  $y$ .

$\nabla f_0(x) = (1/2)(P + P^T)x + q = Px + q$  (because  $P$  is symmetric).

$$\begin{aligned} \nabla f_0(x^*)^T(y - x^*) &= (Px^* + q)^T(y - x^*) = \left( \begin{bmatrix} 13 & 12 & -2 \\ 12 & 17 & 6 \\ -2 & 6 & 12 \end{bmatrix} \begin{bmatrix} 1 \\ 1/2 \\ -1 \end{bmatrix} + \begin{bmatrix} -22 \\ -14.5 \\ 13 \end{bmatrix} \right)^T \left( \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} - \begin{bmatrix} 1 \\ 1/2 \\ -1 \end{bmatrix} \right) \\ &= -1(y_1 - 1) + 2(y_3 + 1) \geq -1(1 - 1) + 2(-1 + 1) = 0 \text{ (because } -1 \leq y_i \leq 1, i = 1, 2, 3). \end{aligned}$$

The optimality condition holds for all feasible  $y$  satisfying  $-1 \leq y_i \leq 1, i = 1, 2, 3$ .

Therefore,  $x^* = (1, 1/2, -1)$  is optimal.

## 4.11

a. Minimize  $\|Ax - b\|_\infty$

$\equiv$  Minimize  $\max |Ax - b|$  (i.e. minimize  $\max\{|a_1^T x - b_1|, \dots, |a_m^T x - b_m|\}$ )

$\equiv$  Minimize  $t$  subject to  $|Ax - b| \preceq t \cdot \mathbf{1}$  (i.e.  $|a_i^T x - b_i| \leq t, i = 1, \dots, m$ , so the max element of  $|Ax - b|$  is also upper-bounded by  $t$ )

$\equiv$  Minimize  $t$  subject to  $-t \cdot \mathbf{1} \preceq Ax - b \preceq t \cdot \mathbf{1}$

b. Minimize  $\|Ax - b\|_1$

$\equiv$  Minimize  $\mathbf{1}^T |Ax - b|$  (i.e. minimize  $\sum_{i=1}^m |a_i^T x - b_i|$ )

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $|Ax - b| \preceq t$  (i.e.  $|a_i^T x - b_i| \leq t_i, i = 1, \dots, m$ , so  $\sum_{i=1}^m |a_i^T x - b_i|$  is upper-bounded by  $\sum_{i=1}^m t_i = \mathbf{1}^T t$ )

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $-t \preceq Ax - b \preceq t$

c. Minimize  $\|Ax - b\|_1$  subject to  $\|x\|_\infty \leq 1$

$\equiv$  Minimize  $\mathbf{1}^T |Ax - b|$  subject to  $\max |x| \leq 1$  (i.e. minimize  $\sum_{i=1}^m |a_i^T x - b_i|$  subject to  $\max\{|x_1|, \dots, |x_m|\} \leq 1$ )

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $|Ax - b| \preceq t$  and  $|x| \preceq \mathbf{1}$  (i.e.  $|a_i^T x - b_i| \leq t_i, i = 1, \dots, m$ , so  $\sum_{i=1}^m |a_i^T x - b_i|$  is upper-bounded by  $\sum_{i=1}^m t_i = \mathbf{1}^T t$ , and  $|x_i| \leq 1, i = 1, \dots, m$ , so the max element of  $|x|$  is also upper-bounded by 1)

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $-t \preceq Ax - b \preceq t$  and  $-\mathbf{1} \preceq x \preceq \mathbf{1}$

d. Minimize  $\|x\|_1$  subject to  $\|Ax - b\|_\infty \leq 1$

$\equiv$  Minimize  $\mathbf{1}^T |x|$  subject to  $\max |Ax - b| \leq 1$  (i.e. minimize  $\sum_{i=1}^m |x_i|$  subject to  $\max\{|a_1^T x - b_1|, \dots, |a_m^T x - b_m|\} \leq 1$ )

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $|x| \preceq t$  and  $|Ax - b| \preceq 1$  (i.e.  $|x_i| \leq t_i, i = 1, \dots, m$ , so  $\sum_{i=1}^m |x_i|$  is upper-bounded by  $\sum_{i=1}^m t_i = \mathbf{1}^T t$ , and  $|a_i^T x - b_i| \leq 1, i = 1, \dots, m$ , so the max element of  $|Ax - b|$  is also upper-bounded by 1)

$\equiv$  Minimize  $\mathbf{1}^T t$  subject to  $-t \preceq x \preceq t$  and  $-1 \preceq Ax - b \preceq 1$

e. Minimize  $\|Ax - b\|_1 + \|x\|_\infty$

$\equiv$  Minimize  $\mathbf{1}^T |Ax - b| + \max |x|$  (i.e. minimize  $\sum_{i=1}^m |a_i^T x - b_i| + \max\{|x_1|, \dots, |x_m|\}$ )

$\equiv$  Minimize  $\mathbf{1}^T t + s$  subject to  $|Ax - b| \preceq t$  and  $|x| \preceq s \cdot \mathbf{1}$  (i.e.  $|a_i^T x - b_i| \leq t_i, i = 1, \dots, m$ , so  $\sum_{i=1}^m |a_i^T x - b_i|$  is upper-bounded by  $\sum_{i=1}^m t_i = \mathbf{1}^T t$ , and  $|x_i| \leq s, i = 1, \dots, m$ , so the max element of  $|x|$  is also upper-bounded by  $s$ )

$\equiv$  Minimize  $\mathbf{1}^T t + s$  subject to  $-t \preceq Ax - b \preceq t$  and  $-s \cdot \mathbf{1} \preceq x \preceq s \cdot \mathbf{1}$

## 4.23

Minimize  $\|Ax - b\|_4 = (\sum_{i=1}^m (a_i^T x - b_i)^4)^{1/4}$

$\equiv$  Minimize  $\sum_{i=1}^m (a_i^T x - b_i)^4$  (i.e. if we minimize  $\sum_{i=1}^m (a_i^T x - b_i)^4$ , we also minimize the original question because  $f(x) = x^{1/4}$  is strictly increasing)

$\equiv$  Minimize  $\sum_{i=1}^m t_i^2$  subject to  $(a_i^T x - b_i)^2 \leq t_i, i = 1, \dots, m$  (because if  $(a_i^T x - b_i)^2 \leq t_i, i = 1, \dots, m$ , then  $\sum_{i=1}^m (a_i^T x - b_i)^4$  is upper-bounded by  $\sum_{i=1}^m t_i^2$ )

$\equiv$  Minimize  $t^T t$  subject to  $a_i^T x - b_i = s_i$  and  $s_i^2 \leq t_i, i = 1, \dots, m$  ( $s$  is introduced such that the optimization problem is a QCQP, i.e. quadratic objective + quadratic inequality constraints + affine equality constraints)