

1.

$f(x) \geq f(\bar{x}) + \xi^T(x - \bar{x})$  for any subgradient  $\xi$ ,  
for any  $x \in \mathbb{R}^n$ ,  
Thus,  $\xi^T \bar{x} - f(\bar{x}) \geq \xi^T x - f(x) \forall x$ ,

so  $\xi^T \bar{x} - f(\bar{x}) = \sup_x \{ \xi^T x - f(x) \}$ ,

so  ~~$\xi^T \bar{x} - f(\bar{x}) = \sup_x \{ \xi^T x - f(x) \}$~~   $f^*(\xi) = \xi^T \bar{x} - f(\bar{x})$ ,

or  $f(\bar{x}) = \xi^T \bar{x} - f^*(\xi)$ .

Conversely, if  $f(\bar{x}) = x^{*\top} \bar{x} - f^*(x^*)$

then  $x^{*\top} \bar{x} - f(\bar{x}) = f^*(x^*)$

$\geq x^{*\top} x - f(x)$  for any  $x \in \mathbb{R}^n$

so  $f(x) \geq f(\bar{x}) + x^{*\top}(x - \bar{x})$ ,

so  $x^*$  is a subgradient of  $f$  at  $\bar{x}$ .

2.

Primal dual pair:

$$\begin{array}{ll} \min & -e^T d \\ \text{st.} & Ad = 0 \\ & d \geq 0 \end{array} \quad (P_0)$$

$$\begin{array}{ll} \max & 0 \\ \text{s.t.} & A^T \pi + s = -e \\ & s \geq 0 \end{array} \quad (D_0)$$

$K^P$  bounded  $\Rightarrow$  optimal value of  $(P_0)$  is zero  
( $d=0$  is feasible)

$\Rightarrow$  optimal value of  $(D_0)$  is zero

$\Rightarrow (D_0)$  is feasible,  
with soln  $\pi^0, s^0$ .

$\Rightarrow \pi^0 \neq 0$ , since else need  $s = -e$  and  $s \geq 0$ .

Let  $\bar{y}$  be feasible in  $(D)$ .

Then  $\bar{y} + \alpha \pi^0$  is feasible in  $(D)$  for any  $\alpha \geq 0$ .

So  $K^D$  is unbounded.

3.

$$H(\bar{x} + \alpha d) = A\bar{x} + \alpha Ad \leq b \quad \text{for any } \alpha \geq 0,$$

so  $\bar{x} + \alpha d$  is feasible for any  $\alpha \geq 0$ .

$$c^T(\bar{x} + \alpha d) + \frac{1}{2}(\bar{x} + \alpha d)^T Q(\bar{x} + \alpha d)$$

$$= c^T\bar{x} + \frac{1}{2}\bar{x}^T Q\bar{x} + \alpha(c^T d + \bar{x}^T Q d) + \frac{1}{2}\alpha^2 d^T Q d$$

$$\leq c^T\bar{x} + \frac{1}{2}\bar{x}^T Q\bar{x} + \alpha d^T(c + Q\bar{x})$$

$$\rightarrow -\infty \quad \text{as } \alpha \rightarrow \infty.$$

4.

$$\nabla f(x) = \begin{pmatrix} x_2 \\ x_1 \end{pmatrix}$$

$$\text{KKT points: } \begin{pmatrix} x_2 \\ x_1 \end{pmatrix} + u_1 \begin{pmatrix} -1 \\ 0 \end{pmatrix} + u_2 \begin{pmatrix} 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{matrix} \textcircled{1} \\ \textcircled{2} \end{matrix}$$

$$u_1(1+x_1) = 0, \textcircled{3} \quad u_2(1+x_2) = 0 \quad \textcircled{4}$$

$$-x_1 \leq 1, \textcircled{5} \quad -x_2 \leq 1 \quad \textcircled{6} \quad u_1 \geq 0, u_2 \geq 0 \quad \textcircled{7}, \textcircled{8}$$

Cases:

$$u_1 = 0 \xrightarrow{\textcircled{1}} x_2 = 0 \xrightarrow{\textcircled{4}} u_2 = 0 \xrightarrow{\textcircled{2}} x_1 = 0$$

Satisfies the conditions, so this is a KKT point,  $x = (0, 0)$ .

$$u_1 > 0 \xrightarrow{\textcircled{3}} x_1 = -1 \xrightarrow{\textcircled{2}} u_2 = -1 \text{ violates } \textcircled{8}.$$

So no KKT point with  $u_1 > 0$ .

Only KKT point is origin  $\boxed{\bar{x} = (0, 0)}$

QP unbounded: Let  $\bar{x} = (-1, 0)$ ,  $\bar{d} = (0, 1)$ .

$$\text{Now, } Q = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \text{ so } \bar{d}^T Q \bar{d} = 0.$$

$$\text{Also, } A\bar{x} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \leq \begin{pmatrix} -1 \\ -1 \end{pmatrix} \text{ and } A\bar{d} = \begin{pmatrix} 0 \\ -1 \end{pmatrix} \leq \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

$$\text{Finally, } (c + Q\bar{x})^T \bar{d} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}^T \begin{pmatrix} 0 \\ 1 \end{pmatrix} = -1 < 0.$$

So QP is unbounded.

5.

$$\text{KKT: } c + Qx + A^T y = 0 \quad (1)$$

$$0 \leq y \perp b - Ax \geq 0 \quad (2)$$

$$x^T (1) \Rightarrow c^T x + x^T Q x + x^T A^T y = 0$$

$$\stackrel{(2)}{\Rightarrow} c^T x + x^T Q x + b^T y = 0$$

$$\Rightarrow 2c^T x + x^T Q x = c^T x - b^T y$$

$$\Rightarrow c^T x + \frac{1}{2} x^T Q x = \frac{1}{2} (c^T x - b^T y).$$

So QP soln can be found by finding best KKT point (since constraints linear and feasible region bounded), which is

to solve

$$\begin{aligned} \min \quad & c^T x - b^T y \\ \text{s.t.} \quad & c + Qx + A^T y = 0 \\ & 0 \leq y \perp b - Ax \geq 0. \end{aligned}$$

6.

There are feasible points with value  $< 0$ , e.g.  $\bar{x} = (1, 1, 1)$ .

So any point with value 0 is not optimal. Need  $\bar{x}_i > 0 \forall i$ .

The constraints are linear, so CQ holds everywhere.

The feasible region is bounded, so the global minimum is a KKT point.

So look for all KKT points, and take the best.

KKT conditions:

$$-x_1 x_2 + u_0 - u_1 = 0 \quad (1)$$

$$-x_1 x_3 + 2u_0 - u_2 = 0 \quad (2)$$

$$-x_1 x_2 + 4u_0 - u_3 = 0 \quad (3)$$

$$u_0 (x_1 + 2x_2 + 4x_3 - 48) = 0 \quad (4) \quad u_0 \geq 0 \quad (5)$$

$$u_i x_i = 0, \quad i=1, 2, 3 \quad (6) \quad u_i \geq 0 \quad i=1, 2, 3 \quad (7)$$

$$x_1 + 2x_2 + 4x_3 - 48 \leq 0 \quad (8) \quad x_i \geq 0 \quad i=1, 2, 3 \quad (9)$$

Cases:

(i)  $u_i > 0$  for some  $i=1, 2, 3$ . Then  $x_i = 0$ , so value  $-x_1 x_2 x_3 = 0$ , so not optimal.

(ii)  $u_1 = u_2 = u_3 = 0$ .

(a)  $u_0 = 0 \Rightarrow x_2 x_3 = 0 \Rightarrow -x_1 x_2 x_3 = 0$ .

(b)  $u_0 > 0 \quad 2(1) - (2) \Rightarrow (x_1 - 2x_2) x_3 = 0 \quad (10)$

$4(1) - (3) \Rightarrow (x_1 - 4x_3) x_2 = 0 \quad (11)$

$\Rightarrow x_1 = 2x_2 \quad (12), \quad x_1 = 4x_3 \quad (13)$  for optimal  $x$ .

$\Rightarrow x_2 = 2x_3 \quad (14)$

$\Rightarrow 4x_3 + 4x_3 + 4x_3 = 48 \Rightarrow x_3 = 4$

$\Rightarrow \bar{x} = (16, 8, 4)$ , optimal since only KKT point  $> 0$ .

Optimal value:  
 $-16 \times 8 \times 4$   
 $= -2^9$   
 $= -512$ .