

1(a)

$$\min_x \frac{1}{2} x^T Q x + p^T x$$

$$\Rightarrow f(x) = \frac{1}{2} x^T Q x + p^T x$$

$$f'(x) = Qx + p$$

$$d_k = -f'(x) = -(Qx + p)$$

$$x_{k+1} = x_k + \alpha_k d_k$$

$$x_1 = x_0 + \alpha_0 d_0$$

Exact line search for α_k (α_0)

$$g(x) = \arg \min_{\alpha} f(x_1)$$

$$\frac{\partial f(x_1)}{\partial \alpha} = 0$$

$$f'(x_1)^T \frac{\partial x_1}{\partial \alpha} = 0$$

$$(Qx_1 + p)^T (d_0) = 0$$

$$[Q(x_0 + \alpha_0 d_0) + p]^T (d_0) = 0$$

$$[Qx_0 + p + Q\alpha_0 d_0]^T (d_0) = 0$$

$$[-d_0 + Q\alpha_0 d_0]^T (d_0) = 0$$

$$\alpha_0 d_0^T Q^T d_0 = d_0^T d_0$$

$$\alpha_0 = \frac{d_0^T d_0}{d_0^T Q d_0}$$

$$\text{i.e. } \alpha_k = \frac{d_k^T d_k}{d_k^T Q d_k}$$

This is exact line search α_k

$$x_{k+1} = x_k + \alpha_k d_k$$

Newton's method,

$$\nabla f(x) = Qx + P$$

$$\nabla^2 f(x) = Q$$

$$x_{k+1} = x_k - Q^{-1} (Qx + P)$$

~~$$\begin{pmatrix} 1000 & 1000 \\ 800 & 1 \end{pmatrix} \begin{pmatrix} 1000 \\ 800 \end{pmatrix}$$~~

2)

$$\min_x x^2 + 1$$

$$\text{sub to } (x-2)(x-4) \leq 0$$

$$L(x; \mu) = x^2 + 1 + \mu(x^2 - 6x + 8)$$

$$\nabla_x L(x; \mu) = 2x + (2x - 6)\mu = 0$$

$$2x + 2x\mu - 6\mu = 0$$

$$2x(1 + \mu) = 6\mu$$

$$x = \frac{3\mu}{1 + \mu}$$

If constraint is inactive, then $\mu = 0 \Rightarrow$
 which will lead to $x = 0$

$$\text{but } x^2 - 6x + 8 \leq 0$$

$$8 \leq 0 - \text{It is impossible.}$$

Thus, by complementary slackness,
 $\mu g(x) = 0$

$g_1(x)$ is active constraint, then.

$$x^2 - 6x + 8 = 0$$

$$\left(\frac{3M}{1+M}\right)^2 - 6\left(\frac{3M}{1+M}\right) + 8 = 0$$

$$\frac{9M^2}{(1+M)^2} - \frac{18M}{1+M} + 8 = 0$$

$$9M^2 - 18M(1+M) + 8(1+M)^2 = 0$$

$$9M^2 - 18M - 18M^2 + 8 + 16M + 8M^2 = 0$$

$$-M^2 - 2M + 8 = 0$$

$$M = -4, 2$$

According to KKT Condition,

x^* is a minima if

$$h_i(x^*) = 0$$

$$g_1(x^*) = 0 \quad \lambda g_1(x^*) = 0$$

$$\lambda \geq 0$$

Thus $M = -4$ is rejected, $M = 2$ is valid.

For $M = 2$, $g_1(x) = 0$.

$$x = \frac{3(2)}{1+2} = \underline{2}$$

$$\boxed{x^* = 2} \quad \text{at} \quad \boxed{M = 2}$$

(b)

$$g(x) = \min L(x, M)$$

$$= \min x^2 + 1 + M(x^2 - 6x + 8)$$

$$= \min (1+M)x^2 - 6xM + 1 + 8M$$

$$\nabla_x L(x, M) = 2(1+M)x - 6M = 0$$

$$x = \frac{3M}{1+M}$$

1. $(1+M)x = 3M$

$2(1+M)x = 6M$

$(1+M)x = 3M$

i.e

remin $(1+M)x^2 - 6xM + 1 + 8M$

i.e min $3xM - 6xM + 1 + 8M$

i.e min $-3xM + 1 + 8M$

i.e ~~max~~ $-3\left(\frac{3M}{1+M}\right)M + 1 + 8M$

$g(M) = \frac{-9M^2}{1+M} + 1 + 8M$

Dual $= g(M) = \max. \frac{9M+1-M^2}{(1+M)}$
 s.t ~~$(1+M)x = 3M$~~
 $M \geq 0$

primal problem $\Rightarrow \min_x x^2 + 1$
 s.t $(x-2)(x-4) \leq 0$
 $= \max - (x^2 + 1)$
 s.t $(x-2)(x-4) \leq 0$

Dual problem $\Rightarrow \max \frac{9M+1-M^2}{1+M}$
 s.t $(1+M)x \geq 3M$
 $= \min - \left(\frac{9M+1-M^2}{1+M} \right)$
 s.t $(1+M)x = 3M$

For $x=2, \mu=2$.

$$\begin{aligned} C^T x &\Rightarrow \text{primal solution} \\ &= -(x^2 + 1) \\ &= -5 \\ &= \underline{\underline{-5}} \end{aligned}$$

$$\begin{aligned} \text{Dual solution} &\Rightarrow b^T y \\ &= -\frac{(9\mu + 1 - \mu^2)}{1 + \mu} \\ &= -\left[\frac{9(2) + 1 - 4}{3} \right] \\ &= -5. \end{aligned}$$

$$\therefore C^T x \leq b^T y.$$

Thus this verify weak duality.

(c) $g(\mu) = \frac{9\mu + 1 - \mu^2}{1 + \mu}$

$$\nabla g(\mu) = \frac{(1+\mu)(9-2\mu) - (9\mu+1-\mu^2)}{(1+\mu)^2} = 0.$$

$$(1+\mu)(9-2\mu) = 9\mu+1-\mu^2$$

$$9 + 7\mu - 2\mu^2 = 9\mu + 1 - \mu^2$$

$$\mu^2 + 2\mu - 8 = 0$$

$$(\mu+4)(\mu-2) = 0$$

$$\mu = -4, \mu = 2$$

but since $\mu \geq 0$,

$$\boxed{\mu = 2}$$

$$\text{Dual value} = \frac{9\mu + 1 - \mu^2}{1 + \mu} = \underline{\underline{5}}.$$

Since $C^T x = b^T y \Rightarrow \text{Primal sol}^n = \text{Dual sol}^n$.

Strong Duality holds.

$g(u)$ is a maximization problem.

$$\nabla^2 g(u) = \frac{d}{du} \left[\frac{9-2u}{1+u} - \frac{(4u+1-u^2)}{(1+u)^2} \right]$$

$$= \frac{(1+u)(-2) - (9-2u)}{(1+u)^2} - \frac{(1+u)^2(9-2u) - (4u+1-u^2)2(1+u)}{(1+u)^4}$$

$$\nabla^2(g(u)) \Big|_{u=2} = \frac{3(-2) - (9-4)}{9} - \frac{3^2(3) - (15)(2)(3)}{3^4}$$

$$= \frac{-6-5}{9} - \frac{27-90}{3^4} = \frac{-11}{9} + \frac{7}{9} = -\frac{4}{9}$$

$$= -\frac{4}{9}$$

Since it is negative, $g(u)$ is concave function
 $g(u)$ is a concave maximization problem 2

Q.4.

$$(a) \quad \min (x_1 - 4)^2 + (x_2 - 4)^2$$

$$\text{Sub to } 5 - x_1 - x_2 \geq 0$$

$$\text{Here } f(x) = (x_1 - 4)^2 + (x_2 - 4)^2$$

$$\phi(x) = -\log(-5 + x_1 + x_2)$$

$$L(x) = \min_{x_1, x_2} (x_1 - 4)^2 + (x_2 - 4)^2 + \frac{1}{\epsilon_K} \log(-5 + x_1 + x_2)$$

$$\nabla L(x) = \begin{bmatrix} 2(x_1 - 4) - \frac{\epsilon_K}{x_1 + x_2 - 5} \\ 2(x_2 - 4) - \frac{\epsilon_K}{x_1 + x_2 - 5} \end{bmatrix}$$

$$\nabla^2 L(x) = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

$$A = 2 + \frac{\epsilon_K}{(x_1 + x_2 - 5)^2}$$

$$B = \frac{\epsilon_K}{(x_1 + x_2 - 5)^2} = C$$

$$D = 2 + \frac{\epsilon_K}{(x_1 + x_2 - 5)^2}$$

Penalty function:-

$$L(x) = \min_{x_1, x_2} (x_1 - 4)^2 + (x_2 - 4)^2 + CK (5 - x_1 - x_2)^2$$

$$\nabla L(x) = \begin{bmatrix} 2(x_1 - 4) + 2CK(5 - x_1 - x_2) \\ 2(x_2 - 4) - 2CK(5 - x_1 - x_2) \end{bmatrix}$$

$$\nabla^2 L(x) = \begin{bmatrix} 2 + 2CK & 2CK \\ 2CK & 2 + 2CK \end{bmatrix}$$

$$S(b) \quad \min_x \quad 5x_1^2 + 5x_2^2 + 5x_3^2 + 5x_4^2 - 4x_1x_3 - 4x_2x_4 + x_1 - x_2 + 2x_3 - 3x_4$$

$$\begin{bmatrix} x_1 & x_2 & x_3 & x_4 \end{bmatrix} \begin{bmatrix} 5 & 0 & -2 & 0 \\ 0 & 5 & 0 & -2 \\ -2 & 0 & 5 & 0 \\ 0 & -2 & 0 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

$$+ \begin{bmatrix} 1 & -1 & 2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

$$Q = \begin{bmatrix} 2.5 & 0 & -1 & 0 \\ 0 & 2.5 & 0 & -1 \\ -1 & 0 & 2.5 & 0 \\ 0 & -1 & 0 & 2.5 \end{bmatrix}$$

$$P^T = [1 -1 2 -3]$$

$$Q_2 = \begin{bmatrix} 10 & 0 & -4 & 0 \\ 0 & 10 & 0 & -4 \\ -4 & 0 & 10 & 0 \\ 0 & -4 & 0 & 10 \end{bmatrix}$$