

SOLUTIONS TO FINAL

Q1. (a)

Primal LP

$$\begin{aligned} &\text{minimize} && c^T x \\ &x \in \mathbb{R}^4 \\ &\text{subject to} \\ &Ax = b \\ &x \geq 0 \end{aligned}$$

where

$$A = \begin{bmatrix} 1 & 2 & 1 & 0 \\ 3 & 1 & -1 & 1 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad c = \begin{bmatrix} -1 \\ 3 \\ 1 \\ 1 \end{bmatrix}$$

Dual LP

$$\begin{aligned} &\text{maximize} && b^T \pi \\ &\pi \in \mathbb{R}^2, s \in \mathbb{R}^4 \\ &\text{subject to} \\ &A^T \pi + s = c \\ &s \geq 0 \end{aligned}$$

that is

$$\begin{aligned} &\text{maximize} && \pi_1 + 2\pi_2 \\ &\pi \in \mathbb{R}^2, s \in \mathbb{R}^4 \\ &\text{subject to} \end{aligned}$$

$$\begin{bmatrix} 1 & 3 \\ 2 & 1 \\ 1 & -1 \end{bmatrix} \pi + \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} = \begin{bmatrix} -1 \\ 3 \\ 1 \end{bmatrix}, \quad \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} \geq 0$$

①

(b)

$$\text{minimize } -\pi_1^+ + \pi_1^- = 2\pi_2^+ + 2\pi_2^-$$

$$\pi^+, \pi^- \in \mathbb{R}^2, s \in \mathbb{R}^4$$

subject to

$$\begin{bmatrix} 1 & 3 & -1 & -3 \\ 2 & 1 & -2 & -1 \\ 1 & -1 & -1 & 1 \\ 0 & 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} \pi^+ \\ \pi^- \end{bmatrix} + \begin{bmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{bmatrix} = \begin{bmatrix} -1 \\ 3 \\ 1 \\ 1 \end{bmatrix}$$

$$\pi^+, \pi^-, s \geq 0$$

Q2. (a)

$$\text{minimize } 4 \frac{y_1}{2} + 4 \frac{y_2 + y_1}{2} + \frac{3 + y_2}{2}$$

$$x_1, x_2, y_1, y_2 \in \mathbb{R}$$

subject to

$$(x_1 - 0)^2 + (y_1 - 0)^2 = 4$$

$$(x_2 - x_1)^2 + (y_2 - y_1)^2 = 4$$

$$\left(\frac{3}{2} - x_2\right)^2 + (3 - y_2)^2 = 1$$

$$y_2 - \frac{x_2^2}{6} - \frac{x_2}{4} + 1 \geq 0$$

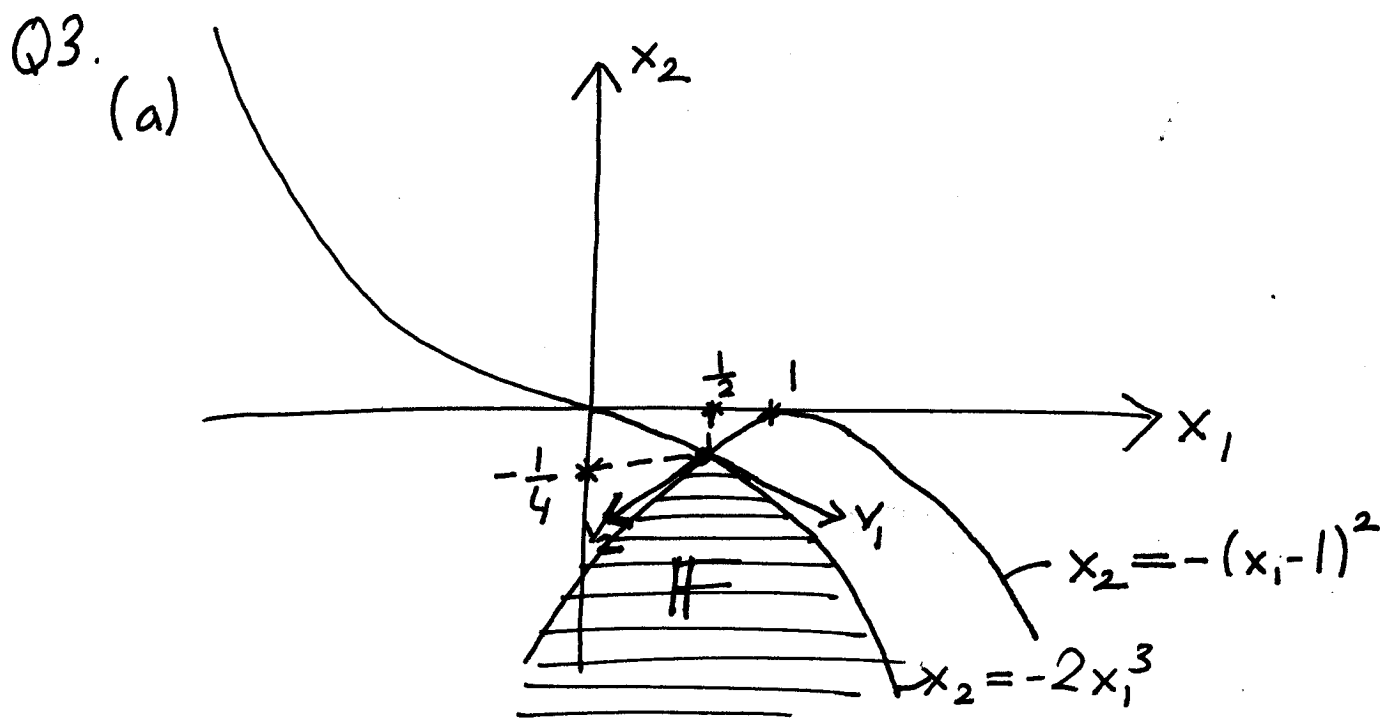
$$y_1 - \frac{x_1^2}{6} - \frac{x_1}{4} + 1 \geq 0$$

(b) Mixed Logarithmic Barrier-Penalty Function

$$M(x_1, x_2, y_1, y_2; \mu) = 4y_1 + \frac{5y_2}{2} + \frac{3}{2} + \frac{1}{2\mu} (x_1^2 + y_1^2 - 4)^2$$

(2)

$$\begin{aligned}
& + \frac{1}{2M} \left( (x_2 - x_1)^2 + (y_2 - y_1)^2 - 4 \right)^2 \\
& + \frac{1}{2M} \left( (3 - x_2)^2 + (3 - y_2)^2 - 1 \right)^2 \\
& - M \ln \left( y_2 - \frac{x_2^2}{6} - \frac{x_2}{4} + 1 \right) \\
& - M \ln \left( y_1 - \frac{x_1^2}{6} - \frac{x_1}{4} + 1 \right)
\end{aligned}$$



$$\begin{aligned}
-2x_1^3 &= -(x_1 - 1)^2 \iff 2x_1^3 - x_1^2 + 2x_1 - 1 = 0 \\
&\iff (2x_1 - 1)(x_1^2 + 1) = 0 \\
&\iff x_1 = \frac{1}{2}
\end{aligned}$$

Line tangent to  $x_2 = -2x_1^3$  at  $(\frac{1}{2}, -\frac{1}{4})$  has slope

$$\left. \frac{d(-2x_1^3)}{dx_1} \right|_{x_1 = \frac{1}{2}} = -\frac{3}{2}$$

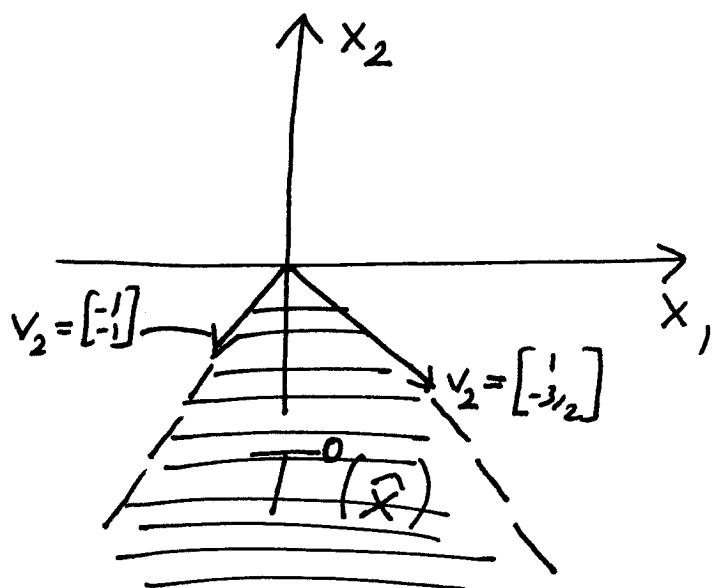
$$v_1 = \left(1, -\frac{3}{2}\right) \quad (\text{OR ANY POSITIVE MULTIPLE})$$

Line tangent to  $x_2 = -(x_1 - 1)^2$  at  $\left(\frac{1}{2}, -\frac{1}{4}\right)$  has slope

$$\left. \frac{d(-(x_1 - 1)^2)}{dx_1} \right|_{x_1 = 1/2} = 1$$

$$v_2 = (-1, -1) \quad (\text{OR ANY POSITIVE MULTIPLE})$$

Tangent cone ~~is the any~~ consists of all nonnegative linear combinations of  $v_1$  and  $v_2$ .



$$T^0(\bar{x})$$

$$=$$

$$\left\{ \alpha_1 \begin{bmatrix} -1 \\ -1 \end{bmatrix} + \alpha_2 \begin{bmatrix} 1 \\ -3/2 \end{bmatrix} : \alpha_1, \alpha_2 \geq 0 \right\}$$

$$(b) \quad J_a(\bar{x}) = \begin{bmatrix} \nabla c_1(\bar{x})^T \\ \nabla c_2(\bar{x})^T \end{bmatrix} = \begin{bmatrix} -3/2 & -1 \\ 1 & -1 \end{bmatrix}$$

Since  $\{\nabla c_1(\bar{x}), \nabla c_2(\bar{x})\}$  is linearly independent, LICQ holds. Consequently

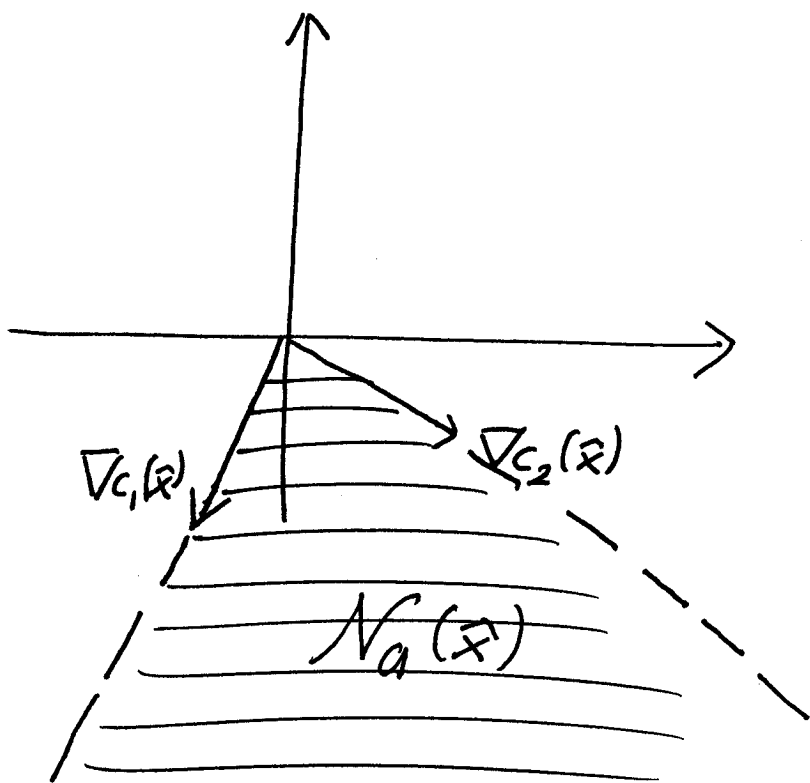
constraint qualification holds, that is

$$T^{\circ}(\bar{x}) = \{ p \in \mathbb{R}^2 : J_a(\bar{x}) p \geq 0 \}$$

(c)

$$\mathcal{N}_a(\bar{x}) = \{ \alpha_1 \nabla_{c_1}(\bar{x}) + \alpha_2 \nabla_{c_2}(\bar{x}) : \alpha_1, \alpha_2 \geq 0 \}$$

$$= \left\{ \alpha_1 \begin{bmatrix} -3/2 \\ -1 \end{bmatrix} + \alpha_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix} : \alpha_1, \alpha_2 \geq 0 \right\}$$



(d)  $\nabla_f(\bar{x}) = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

(1)  $\nabla_f(\bar{x}) = \lambda_1 \nabla_{c_1}(\bar{x}) + \lambda_2 \nabla_{c_2}(\bar{x})$

holds for  $\lambda_1 = 0$  and  $\lambda_2 = 1 \geq 0$   
(that is  $\nabla_f(\bar{x}) = J_a(\bar{x})^T \lambda$  for some  $\lambda \geq 0$ )

(2)  $c_1(\bar{x}) = c_2(\bar{x})$

Consequently first order necessary conditions hold.

Q4. (a)

Since constraints are linear, constraint qualification holds at all  $x$ .

If  $x_*$  is a local minimizer, then following KKT conditions holds for some  $\lambda \in \mathbb{R}^m$  and  $s \in \mathbb{R}^n$ .

KKT conditions

$$(1) Ax_* = b$$

$$(2) x_* \geq 0$$

$$(3) \underbrace{Hx_*}_{\nabla f(x_*)} = A^T \lambda + s$$

$$(4) s \geq 0$$

$$(5) s^T x_* = 0 \quad \boxed{\text{COMPLEMENTARITY}}$$

(b)

Suppose  $x$  is any feasible point and  $x_*$  is a point satisfying the KKT conditions. Then

$$\begin{aligned} f(x) &= \frac{1}{2} x^T H x \\ &= \frac{1}{2} (x_*^T H x_* + (x-x_*)^T H (x-x_*) \\ &\quad + 2(x-x_*)^T H x_*) \\ &\quad = \frac{1}{2} (x-x_*)^T H x_* + \frac{1}{2} x_*^T H (x-x_*) \\ &\quad \quad \quad \text{(since } H \text{ is symmetric)} \\ &= \frac{1}{2} x_*^T H x_* + \frac{1}{2} (x-x_*)^T H (x-x_*) \\ &\quad \quad \quad \geq 0 \quad \text{SINCE } H \succ 0 \\ &\quad + (x-x_*)^T (A^T \lambda + s) \\ &\quad \quad \quad \underbrace{(x^T A^T - x_*^T A^T)}_b \lambda + \underbrace{x^T s}_{\geq 0} - \underbrace{x_*^T s}_0 \\ &\quad \quad \quad \boxed{\text{SINCE } x \text{ IS FEASIBLE i.e. } Ax=b} \quad \boxed{\text{SINCE } x_* \text{ IS FEASIBLE}} \quad \boxed{\text{SINCE } x \text{ IS FEASIBLE, i.e. } x \geq 0} \quad \boxed{\text{COMPLEMENT}} \\ &\geq f(x_*) = \frac{1}{2} x_*^T H x_* \end{aligned}$$

Consequently  $x_*$  is a global minimizer.

Q5.

First consider the unperturbed problem.

$$f_*(0) = c^T x_*(0)$$

where  $x_*(0)$  satisfies

$$(1) \quad Ax_*(0) = b \quad \text{with } b \in \mathbb{R}^m$$

$$(2) \quad x_*(0) \geq 0$$

$$(3) \quad c = A^T \pi + s$$

$$(4) \quad s \geq 0$$

$$(5) \quad s^T x_*(0) = 0$$

for some  $\pi \in \mathbb{R}^m$  and  $s \in \mathbb{R}^n$ .

From (3)

$$\begin{aligned} f_*(0) &= (A^T \pi + s)^T x_*(0) \\ &= \pi^T A x_*(0) + \underbrace{s^T x_*(0)}_{0 \text{ (4)}} \\ &= \pi^T b \end{aligned}$$

Focus on the perturbed problem

$$\begin{aligned} f_*(\epsilon) &= c^T x_*(\epsilon) \\ &= (A^T \pi + s)^T x_*(\epsilon) \\ &= \pi^T A x_*(\epsilon) + s^T x_*(\epsilon) \end{aligned}$$



$$= \Pi^T (b + \epsilon e_m) + s^T x_*(\epsilon)$$

Notice that

$$(s^T x_*(\epsilon))_j = \begin{cases} 0 & s_j = 0 \\ & \text{(CONSTRAINT IS INACTIVE)} \\ s_j(x_*(\epsilon))_j & s_j \neq 0 \\ & \text{(CONSTRAINT IS ACTIVE)} \end{cases}$$

$$= \begin{cases} 0 & s_j = 0 \\ s_j(2\epsilon) & s_j \neq 0 \end{cases}$$

$\implies$

$$s^T x_*(\epsilon) = (2\epsilon) s^T e_n$$

Consequently

$$f_*(\epsilon) = \Pi^T (b + \epsilon e_m) + (2\epsilon) s^T e_n$$

and the derivative is given by

$$\begin{aligned} f_*'(0) &= \lim_{\epsilon \rightarrow 0} \frac{f_*(\epsilon) - f_*(0)}{\epsilon} = \lim_{\epsilon \rightarrow 0} \frac{(\epsilon)\Pi^T e_m + (2\epsilon)s^T e_n}{\epsilon} \\ &= \underline{\underline{\Pi^T e_m + 2s^T e_n}} \end{aligned}$$