

Past Exam Paper 1

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2. (a) (I) $g_1(x) = x_1^2 + x_2^2 - 4$ $g_2(x) = x_1 - 5x_2 + 5$

$\nabla^2 g_1(x) = 2I \succeq 0$ is convex

$\nabla^2 g_2(x) = 0$ is convex

Find $x_1 = 0, x_2 = 1$ as Slater point.

Slater Condition holds \Rightarrow MFCR holds at every feasible point

(II) The KKT is shown below

$$\begin{cases} \begin{bmatrix} 1 \\ 3x_2^2 \end{bmatrix} + \lambda_1 \begin{bmatrix} 2x_1 \\ 2x_2 \end{bmatrix} + \lambda_2 \begin{bmatrix} 1 \\ -5 \end{bmatrix} = 0 \quad (1) \\ \lambda_1 (x_1^2 + x_2^2 - 4) = 0 \quad (2) \\ \lambda_2 (x_1 - 5x_2 + 5) = 0 \quad (3) \\ \lambda_1 \geq 0, \lambda_2 \geq 0 \end{cases}$$

If $\lambda_1 = 0, \lambda_2 = 0$ From (1), the result is unfeasible

If $\lambda_1 = 0, \lambda_2 > 0$ From (2), $x_1 = 0, x_2 = 1$

From (1), $3 - 5\lambda_2 = 0 \Rightarrow \lambda_2 = \frac{3}{5}$

If $\lambda_1 > 0, \lambda_2 = 0$ From (1), $1 + 2\lambda_1 x_1 = 0 \Rightarrow x_1 = -\frac{1}{2\lambda_1}$

$3x_2^2 + 2\lambda_1 x_2 = 0, x_2 = \frac{-2\lambda_1 \pm \sqrt{4\lambda_1^2}}{6}$

$x_2 = 0$ OR $x_2 = -\frac{4\lambda_1}{6} = -\frac{2}{3}\lambda_1$

If $x_2 = 0$, from (2) $(-\frac{1}{2\lambda_1})^2 \cdot \lambda_1 - 4\lambda_1 = 0$

$\Rightarrow \frac{1}{4\lambda_1^2} \cdot \lambda_1 - 4\lambda_1 = 0, \frac{1}{4\lambda_1} - 4\lambda_1 = 0$

$\Rightarrow 1 - 16\lambda_1^2 = 0, \lambda_1 = \frac{1}{4}$

If $x_2 = -\frac{2}{3}\lambda_1$, from (2) $\lambda_1 (\frac{1}{4\lambda_1^2} + \frac{4}{9}\lambda_1^2 - 4) = 0$

$\Rightarrow \frac{1}{4\lambda_1^2} + \frac{4}{9}\lambda_1^2 - 4 = 0, \frac{1}{4} + \frac{4}{9}\lambda_1^4 - 4\lambda_1^2 = 0$

$\Rightarrow \frac{1}{4} + \frac{4}{9}t^2 - 4t = 0, t = \frac{4 \pm \sqrt{16 - \frac{1}{9}}}{2 \cdot \frac{4}{9}}$

$\Rightarrow \lambda_1 = \dots$

2. (b) $h(x) = \|Ax\|_2^2 - 1, g_i(x) = -x_i$

The MFCR def is:

$\mu \nabla h(x) + \sum -\lambda_i \nabla g_i(x) = 0, \lambda_i g_i(x) = 0, \text{ for } i \in I(x), \lambda_i \geq 0$

$\Leftrightarrow \mu (A^T A x) + [-\lambda_1, \dots, -\lambda_n]^T = 0, \lambda_i g_i(x) = 0, \lambda_i \geq 0$

$\Leftrightarrow \mu (A^T A x) - [\lambda_1, \dots, \lambda_n]^T = 0, \lambda_i g_i(x) = 0, \lambda_i \geq 0$

$$\begin{aligned} \Leftrightarrow \mathcal{U}(A^T A x) - [\lambda_1 \dots \lambda_n]^T &= 0, \lambda_i g_i(x) = 0, \lambda_i \geq 0 \\ \Leftrightarrow \mathcal{U}(A^T A x) x - [\lambda_1 \dots \lambda_n]^T x &= 0 \\ \Leftrightarrow \mathcal{U} = 0, \lambda_i = 0 &\Rightarrow \text{MFCQ holds} \end{aligned}$$

3. (a)

Let $h(y) = \ln(y^2 + 1)$

$$\begin{aligned} h''(y) &= \left(\frac{2y}{y^2 + 1} \right)' = \left(\frac{2(y^2 + 1) - 2y \cdot 2y}{y^4 + 2y^2 + 1} \right) = \frac{2y^2 - 4y^2 + 2}{y^4 + 2y^2 + 1} \\ &= \frac{-2y^2 + 2}{y^4 + 2y^2 + 1} \leq \frac{-2y^2 + 2}{2y^2 + 1} \leq m \end{aligned}$$

$$\|\nabla^2 f(x)\|_2 = \|A^T \nabla^2 h(Ax - b) A\|_2 \leq m \|A^T A\|_2 \leq L \dots$$

(b) Compare $\frac{2}{L}$, $L \dots$

4. t^2 is convex, $\sum x_i^2$ is convex

$-\ln(t-5)$ is convex. $x_2 + 2x_3$ is a linear map

g_μ is convex

$$\nabla g_\mu = \begin{bmatrix} 2x_1 & 2x_2 - \mu \frac{1}{x_2 + 2x_3 - 5} & 2x_3 - \mu \frac{2}{x_2 + 2x_3 - 5} & 2x_4 \end{bmatrix} = 0$$

$$x_1 = 0$$

$$x_2 = \frac{\mu}{2} \cdot \frac{1}{x_2 + 2x_3 - 5} \quad x_3 = \mu \frac{1}{x_2 + 2x_3 - 5} \quad x_4 = 0$$

$$2x_2 = x_3$$

$$x_2 = \frac{\mu}{2} \cdot \frac{1}{x_2 + 4x_2 - 5} \Rightarrow x_2 = \frac{\mu}{2} \cdot \frac{1}{5x_2 - 5}$$

$$5x_2^2 - 5x_2 = \frac{\mu}{2}, \text{ let } \mu \rightarrow 0 \quad x_2 = 1, \quad x_3 = 2$$

So the result is $(0 \ 1 \ 2 \ 0)$

5. (a)

Min $y_1 + y_2$

$$y_1 \geq (|x_1| + 1)^5 \quad y_2 \geq \sqrt{x_1^2 + 3x_2^2 + x_3^2 + 1}$$

$$y_3 \geq |x_1| \quad y_4 \geq (y_3 + 1)^5$$

$$y_1 (y_3 + 1) \geq (y_3 + 1)^6 \quad y_4 \geq (y_3 + 1)^2$$

$$y_1 (y_3 + 1) \geq y_4^3$$

$$y_1 y_4 (y_3 + 1) \geq y_4^4 \quad y_5 \geq y_4^2$$

$$u_1 y_{12} (y_2 + 1) \geq u_2^2$$

$$y_1 y_4 (y_3 + 1) \geq y_4^2 \quad y_5 \leq y_4$$

$$y_1 y_4 (y_3 + 1) \geq y_5^2$$

(b) $b = [5 \ 1]$

$$A_1 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \quad A_2 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Dual $\max b^T y$

$$\text{s.t.} \quad \begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} - y_1 \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} - y_2 \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \succeq 0$$

$U_p = V_d$.

$\text{tr}(X^*)$ satisfy the primal question, $X^* \succeq 0$

$$\text{tr}(CX) = \text{tr} \left(\begin{pmatrix} 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 4 & 0 & 0 & 0 \end{pmatrix} \right) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 4 \end{pmatrix} = 6$$

If it is the optimal sol. $b^T y = 6$

$$\begin{bmatrix} 5 \\ 1 \end{bmatrix}^T (y_1, y_2) = 6 \quad \Leftrightarrow y_1 + y_2 = 6$$

Dual s.t. $\begin{bmatrix} 2-y_1 & 1 & 0 & 0 \\ 1 & 2-y_1 & -\frac{y_2}{2} & 0 \\ 0 & -\frac{y_2}{2} & 2-y_1-y_2 & 0 \\ 0 & 0 & 0 & 1-y_1 \end{bmatrix} \succeq 0$. $\begin{matrix} 1-y_1 \geq 0 & y_1 \leq 1 \\ 2-y_1-y_2 \geq 0 & y_1+y_2 \leq 2 \end{matrix}$

$$5y_1 + y_2 = 4y_1 + (y_1 + y_2) \leq 4 + 2 \leq 6$$

Let $y_1=1, y_2=1$, it achieve the optimal value

Consider max's s.t

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 1 & -\frac{1}{2} & 0 \\ 0 & -\frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \left| \begin{matrix} 1 & 1 & 0 \\ 1 & 1 & -\frac{1}{2} \\ 0 & -\frac{1}{2} & 0 \end{matrix} \right| = 0 + 0 + 0 - \frac{1}{4} = -\frac{1}{4} < 0$$

It is not feasible

So x^* is not a optimal solution