

# TUTORIAL ∴

Today:

- ① discuss a non-linear problem and show how the c-g algorithm - works
- ② Demo's (2) (last)  
c-g (QM)  
} Rosen <sup>gen</sup> -brock (Non-linear problem)
- ③ problem set - 6 (QN-8)  
↪ QN-(Last QN)

(Q1.) Midsem-1 (QN-5) of Newton's method

$$\begin{aligned} f: \mathbb{R}^2 &\rightarrow \mathbb{R} \\ f(x) &= \min_{x \in \mathbb{R}^2} \{ x_1^4 + 2x_1^2 x_2^2 + x_2^4 \} \\ x^0 &= \begin{pmatrix} 2 \\ 1 \end{pmatrix} \\ \nabla f(x) &= \begin{Bmatrix} 4x_1^3 + 4x_1 x_2^2 \\ 4x_1^2 x_2 + 4x_2^3 \end{Bmatrix} \\ \nabla f(x^0) &= \begin{Bmatrix} 40 \\ 20 \end{Bmatrix} \\ d^{(0)} &= -\nabla f(x^0) = \begin{pmatrix} -40 \\ -20 \end{pmatrix} \\ \alpha_0 &= \arg \min_{\alpha \in \mathbb{R}^+} f(x^0 + \alpha d^0) \end{aligned}$$
$$\begin{aligned} x_{k+1} &= x_k + \alpha_k d_k \\ x_1 &= \boxed{x_0} + \boxed{\alpha_0} d_0 \end{aligned}$$

$$\tilde{z} = \min_{\alpha \in \mathbb{R}^+} f\left(\begin{pmatrix} 2 \\ 1 \end{pmatrix} + \alpha \begin{pmatrix} -40 \\ -20 \end{pmatrix}\right)$$

$$g(\alpha) = \min_{\alpha \in \mathbb{R}^+} (\alpha - 40\alpha)^2 + 2(\alpha - 40\alpha)^2(1 - 20\alpha)^2 + (1 - 20\alpha)^4$$

first-order-necess

$$\frac{dg(\alpha)}{d\alpha} = 0$$

$$25(1 - 20\alpha) = 0$$

$$\boxed{\alpha^* = \frac{1}{20}}$$

$$x^{(1)} = \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \frac{1}{20} \begin{pmatrix} -40 \\ -20 \end{pmatrix}$$

$$= \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \begin{pmatrix} -2 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

C.G. - got to minima in exactly one-step.

$$x^0 = \begin{pmatrix} a \\ b \end{pmatrix}$$

$$a, b \in \mathbb{R}$$

$$\boxed{\alpha = \frac{-1}{4(b^2 - a^2)}}$$

can check it?  
out

Non-linear - CG method.

Q 1

(Levenberg)

$$x^0 \rightarrow x^1 = -\nabla f(x^0)$$

$$k \geq 0, x_{k+1} = x_k + \alpha_k d_k$$

$$\alpha_k = \arg \min_{\alpha \in \mathbb{R}^+} f(x_k + \alpha d_k)$$

$$d_k = -\nabla f(x_k)$$

$$d_{k+1} = -\nabla f(x_{k+1}) + \beta_{k+1} d_k$$

$$\beta_{k+1} = \frac{\nabla f(x_{k+1})^T \nabla f(x_{k+1})}{\nabla f(x_k)^T \nabla f(x_k)}$$

$$\{f: \mathbb{R}^n \rightarrow \mathbb{R}\}$$

$$x_{k+1} = x_k + \alpha_k \{d_k\} = \{-\nabla f(x_k)\}$$

$$\alpha_k = \arg \min_{\alpha \in \mathbb{R}^+} f(x_k + \alpha d_k)$$

$$f(x) = \left\{ \frac{1}{2} x^T A x - b^T x + c \right\} \quad A \in \text{symmetric}$$

$$\nabla f = \{Ax - b\} \quad d_k = -\nabla f = b - Ax$$

$$\alpha_k = \arg \min_{\alpha \in \mathbb{R}^+} f(x_k + \alpha d_k)$$

$$= g(\alpha) = \frac{1}{2} (x_k + \alpha d_k)^T A (x_k + \alpha d_k) - b^T (x_k + \alpha d_k) + c$$

$$\left\{ \frac{dg(\alpha)}{d\alpha} = 0 \right\}$$

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$\{f(x) = \ln(x_1)\}$$

$\alpha \left\{ \sin(\pi/2) \dots \right\}$   


---

 show - Demos c-a works (QP)  
 (N-L?)

(Q2) Last on problem-set-6  
 { Find the Dual }

$H \Rightarrow \{ \text{invertible} \}$

$$\min_x \quad \frac{1}{2} x^T H x + c^T x$$

s.t.  $Ax \geq b$

$$\mathcal{L}(x, \mu) = \frac{1}{2} x^T H x + c^T x + \mu^T (b - Ax)$$

$$\mathcal{D}(\mu) = \min_x \quad \frac{1}{2} x^T H x + c^T x + \mu^T (b - Ax)$$

$$Hx + c - A^T \mu = 0$$

$$x = H^{-1} (A^T \mu - c)$$

Dual problem

$$\begin{aligned} \max_{\mu \geq 0} \quad & \frac{1}{2} (\mu^T A - c^T) H^{-1} H^{-1} (A^T \mu - c) \\ & + c^T H^{-1} A^T \mu - c^T H^{-1} c \\ & + \mu^T b - \mu^T A H^{-1} (A^T \mu) \\ & + \mu^T A H^{-1} c \end{aligned}$$

$$\max_{\mu \geq 0} \quad \frac{1}{2} \mu^T (A H^{-1} A^T) \mu + \mu^T (b + A H^{-1} c) - \frac{1}{2} c^T H^{-1} c$$

$$\max_{\mu \geq 0} -\frac{1}{2} \mu^T (A \mu^T A^T) \mu + \mu^T (b + A \mu^T c)$$

Q8.) problem-sheet-6:-

$$\text{solve } \min_{x \in \mathbb{R}^2} \{ 14x_1^2 - x_1 + 6x_2 - x_2^2 + 7 \}$$

$$\text{s.t. } \left. \begin{aligned} x_1 + x_2 &\leq 2 \\ x_1 + 2x_2 &\leq 3 \end{aligned} \right\}$$

primal problem (P)

$$\mathcal{L}(x, \mu_1, \mu_2) = 14x_1^2 - x_1 + 6x_2 - x_2^2 + 7 + \mu_1(x_1 + x_2 - 2) + \mu_2(x_1 + 2x_2 - 3), \mu_1 \geq 0, \mu_2 \geq 0$$

{ strong - Duality concept in QP }

primal problem  
obje

Dual problem  
objective.

{ P.S.D }

→ strong - Duality  
hold }

$$f(x) = 14x_1^2 - x_1 + 6x_2 - x_2^2 + 7$$

$$= \frac{1}{2} \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{bmatrix} 28 & 0 \\ 0 & -2 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} -1 & 6 \end{pmatrix}^T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + 7$$

$$= \frac{1}{2} x^T A x + C^T x + b$$

$\left\{ \begin{array}{l} A \text{ is neither PSD} \\ \text{nor N.S.D} \end{array} \right\}$   
 $\left\{ f \text{ is not convex nor concave} \right\}$

$\nabla_x \mathcal{L}(x, \mu) = 0$  (first-order condition)  
 (Lagrange-Duality theorem)

$$\begin{cases} 2x_1 - 1 + \mu_1 + \mu_2 \\ -2x_2 + 6 + \mu_1 + 2\mu_2 \end{cases} = 0$$

$$\begin{cases} x_1 = \frac{1 - \mu_1 - \mu_2}{2} \\ x_2 = \frac{6 + \mu_1 + 2\mu_2}{2} \end{cases}$$

complementary-slackness condition

$$\mu_1 (g_1(x_1(\mu), x_2(\mu))) = 0$$

$$\begin{aligned} \mu_1 \left\{ \frac{1 - \mu_1 - \mu_2}{2} + \frac{6 + \mu_1 + 2\mu_2}{2} - 2 \right\} &= 0 \\ \mu_2 \left\{ \frac{1 - \mu_1 - \mu_2}{2} + 6 + \mu_1 + 2\mu_2 - 3 \right\} &= 0 \end{aligned}$$

$\mu_1, \mu_2$   
(0,0)

( $\mu_1 = 0$ )

( $\mu_2 = 0$ )

$$\left\{ \begin{array}{l} \frac{1 - \mu_2 + 6 + 2\mu_2 - 3}{28} = 0 \\ \mu_2 = \end{array} \right\}$$

$$\left[ \begin{array}{l} \frac{1 - \mu_1 - \mu_2 + 6 + \mu_1 + 2\mu_2 - 2}{28} = 0 \\ \frac{1 - \mu_1 - \mu_2 + 6 + \mu_1 + 2\mu_2 - 3}{28} = 0 \end{array} \right]$$

$$AX = b \quad \{(\mu_1, \mu_2)\}$$

$$(M_1, M_2 = ?)$$

①  $\mu_1 = 0, \mu_2 = 0$  (Lecture - 16 page no - 5)  
(not satisfy constraints)

②  $\mu_1 = \frac{-1429}{14}, \mu_2 = \frac{673}{14}$   
(violated  $\mu_1 \geq 0$ )

③  $\mu_1 = 0, \mu_2 = -1.545$  (violated  $\mu_2 \geq 0$ )

④  $\mu_2 = 0, \mu_1 = -2.2307$  (violated  $\mu_1 \geq 0$ )

(no candidate for minima)

sufficiency condition for global.

$(\mu_1, \mu_2) \rightarrow f(\mu_1^*, \mu_2^*)$

$(\mu_1, \mu_2)$   
 $(x_1, x_2)$

$$\left\{ \begin{array}{l} \min_{x, \mu_1, \mu_2} \mathcal{L}(x, \mu_1^*, \mu_2^*) \\ \text{sufficient} \end{array} \right\}$$

$\mu_1^* = 0, \mu_2^* = 0$

$$\mathcal{L}(x^*) \left\{ \begin{array}{l} g_1(x^*) \leq 0 \\ g_2(x^*) \leq 0 \end{array} \right\} \quad (x^*, \mu^*) = \text{ideal pair}$$

primal - objective - fn value

Dual problem

$$D(\lambda) = \min_{x \in \mathbb{R}^2} 14x_1^2 - x_1 + 6x_2 - x_2^2 + 7 + \mu_1(x_1 + x_2 - 2) + \mu_2(x_1 + 2x_2 - 3)$$

$$x_1 = \frac{1 - \mu_1 - \mu_2}{28}$$

$$\mu_2 = \frac{\mu_1 + 2\mu_2 + 6}{2}$$

Dual problem

$$\max \mu_1, \mu_2 \geq 0$$

$$\left\{ \begin{array}{l} 14 \left( \frac{1 - \mu_1 - \mu_2}{28} \right)^2 - \left( \frac{1 - \mu_1 - \mu_2}{28} \right) + 6 \left( \frac{\mu_1 + 2\mu_2 + 6}{2} \right) - \left( \frac{\mu_1 + 2\mu_2 + 6}{2} \right)^2 + 7 + \mu_1 \left( \frac{1 - \mu_1 - \mu_2}{28} - 2 \right) + \mu_2 \left( \frac{\mu_1 + 2\mu_2 + 6}{2} - 3 \right) \end{array} \right\}$$



$$+ u_2 \left( \frac{1 - u_1 - u_2}{28} + u_1 f_2 u_2 \right)$$

first -

second.

Dual-problem. objective-function.

value

# Tutorial-10

Saturday, 19 March 2022 11:02 AM

①  $f(x) = -\ln(x) + x^2$ . Use Conjugate gradient method to find the optimal sol. let  $x^0 = 1$

$$\nabla f(x) = -\frac{1}{x} + 2x = 0 \Rightarrow -1 + 2x^2 = 0$$

$$\Rightarrow 2x^2 = 1$$

$$\Rightarrow x = \frac{1}{\sqrt{2}} \text{ (optimal)}$$

$$x^{k+1} = x^k + \alpha_k d^k$$

$$d^0 = -\nabla f(x^0)$$

$$d^0 = -1$$

$$\nabla f(x^0) = -1 + 2 = 1$$

$$x^1 = 1 + \alpha_0 (-1) = 1 - \alpha_0 \text{ optimline search}$$

$$f(x^1) = g_{x^1}(\alpha_0) = -\ln(1 - \alpha_0) + (1 - \alpha_0)^2$$

$$g'_{x^1}(\alpha_0) = \frac{1}{1 - \alpha_0} - 2(1 - \alpha_0) = 0$$

$$-1 + 2(1 - \alpha_0)^2 = 0$$

$$\Rightarrow 1 - \alpha_0 = \frac{1}{\sqrt{2}}$$

$$\alpha_0 = 1 - \frac{1}{\sqrt{2}}$$

$$x^1 = 1 - 1 + \frac{1}{\sqrt{2}} = \frac{1}{\sqrt{2}}$$

⑨ a Dual of the LP prob

$$\min_{x \in \mathbb{R}^n} C^T x$$

$$\begin{cases} Ax = b \\ A_1 x \leq b_1 \end{cases}$$

$$\begin{cases} \min C^T x \\ \text{s.t. } Ax \leq b \\ x \geq 0 \end{cases}$$

$$\begin{cases} \min C^T x \\ \text{s.t. } Ax = b \\ x \geq 0 \end{cases}$$

$$\lambda \in \mathbb{R}^m; \mu \in \mathbb{R}^m$$

$$c \in \mathbb{R}^n; \quad A \text{ and } A_1 \in \mathbb{R}^{m \times n}; \quad b \text{ and } b_1 \in \mathbb{R}^m.$$

$$L(x, \lambda, y) = c^T x + \lambda^T (Ax - b) + y^T (A_1 x - b_1)$$

$$= (c^T + \lambda^T A + y^T A_1) x - \lambda^T b - y^T b_1$$

Dual fun:-

$$D(\lambda, y) = \min_{x \in \mathbb{R}^n} L(x, \lambda, y)$$

$$\nabla_x L(x, \lambda, y) = 0 \\ = c^T + \lambda^T A + y^T A_1 = 0$$

$$= \begin{cases} -\lambda^T b - y^T b_1 \\ -\infty \end{cases} \quad \begin{cases} c^T + \lambda^T A + y^T A_1 = 0 \\ \text{otherwise} \end{cases}$$

Dual

$$\max_{\lambda, y} D(\lambda, y) \quad y \geq 0$$

$$= \max \begin{cases} -\lambda^T b - y^T b_1 \\ y \geq 0 \\ c^T + A^T \lambda + A_1^T y = 0 \end{cases} \quad \text{dual problem}$$

(b)  $x \in \mathbb{R}^4$

$$\begin{aligned} \min \quad & 18x_1 + 12x_2 + 2x_3 + 6x_4 \\ \text{subto} \quad & 3x_1 + x_2 - 2x_3 + x_4 = 2 \\ & x_1 + 3x_2 - x_4 = 2 \\ & x_1 \geq 0; x_2 \geq 0; x_3 \geq 0; x_4 \geq 0 \end{aligned}$$

Dual

$$\min \quad c^T x$$

$$\begin{aligned} Ax &= b \\ A_1 x &\leq b_1 \end{aligned}$$

$$C = \begin{bmatrix} 18 \\ 12 \\ 2 \\ 6 \end{bmatrix}$$

$$A = \begin{bmatrix} 3 & 1 & -2 & 1 \\ 1 & 3 & 0 & -1 \end{bmatrix}$$

$$b = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

2x4  
here ...

$L^*$

$$b = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$\begin{matrix} \leq & \wedge & \geq \\ \# & & \# \\ \text{eq} & & \text{vars} \end{matrix}$

$$\lambda = \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}$$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$$A_1 = \begin{bmatrix} -1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}_{4 \times 4}$$

$$\max -\lambda^T b - y^T b$$

$$y \geq 0$$

$$c + A^T \lambda + A_1^T y = 0$$

$$\rightarrow \max -[\lambda_1 \ \lambda_2] \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$$y_1 \geq 0; y_2 \geq 0; y_3 \geq 0; y_4 \geq 0 \quad \text{--- (1)}$$

$$\begin{bmatrix} 18 \\ 12 \\ 2 \\ 6 \end{bmatrix} + \begin{bmatrix} 3 & 1 \\ 1 & 3 \\ -2 & 0 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} - \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = 0 \quad \text{--- (2)}$$

$$\Rightarrow \max -2\lambda_1 - 2\lambda_2$$

$$\left. \begin{aligned} 18 + 3\lambda_1 + \lambda_2 - y_1 &= 0 \\ 12 + \lambda_1 + 3\lambda_2 - y_2 &= 0 \\ 2 - 2\lambda_1 - y_3 &= 0 \\ 6 + \lambda_1 - \lambda_2 - y_4 &= 0 \end{aligned} \right\} \begin{aligned} y_1 &\geq 0; y_2 \geq 0; y_3 \geq 0; \\ &y_4 \geq 0 \end{aligned}$$

$$\Rightarrow \max -2\lambda_1 - 2\lambda_2$$

Sub

$$\left. \begin{aligned} 18 + 3\lambda_1 + \lambda_2 &\geq 0 \\ 12 + \lambda_1 + 3\lambda_2 &\geq 0 \\ 2 - 2\lambda_1 &\geq 0 \\ 6 + \lambda_1 - \lambda_2 &\geq 0 \end{aligned} \right\}$$

Dual

KKT - conditions :-

$$\min_x \quad 18x_1 + 12x_2 + 2x_3 + 6x_4$$

$$3x_1 + x_2 - 2x_3 + x_4 = 2$$

$$x_1 + 3x_2 - x_4 = 2$$

$$x_1 \geq 0; x_2 \geq 0, x_3 \geq 0; x_4 \geq 0$$

active,  $x_1, x_2, x_3, x_4, (x_1, x_2), (x_1, x_3)$

$$\begin{aligned} \mathcal{L}(x, \lambda, \mu) = & 18x_1 + 12x_2 + 2x_3 + 6x_4 + \lambda_1(3x_1 + x_2 - 2x_3 + x_4 - 2) \\ & + \lambda_2(x_1 + 3x_2 - x_4 - 2) - \sum_{i=1}^4 \mu_i x_i \end{aligned}$$

$$\nabla_x \mathcal{L}(x, \lambda, \mu) = \begin{bmatrix} 18 + 3\lambda_1 + \lambda_2 - \mu_1 \\ 12 + \lambda_1 + 3\lambda_2 - \mu_2 \\ 2 - 2\lambda_1 - \mu_3 \\ 6 + \lambda_1 - \lambda_2 - \mu_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

complementary slack

$$\mu_i x_i = 0 \quad \forall i = 1, 2, 3, 4$$



$$\mu_1 = 0 = \mu_2 = \mu_3 = \mu_4$$

- ① feasibility conditions
- ②  $\nabla_x \mathcal{L}(x, \lambda, \mu) = 0; \mu \geq 0 \rightarrow ( \quad )$
- ③ complementary slackness  $\Rightarrow$  case