

MAE-598

Design Optimization

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Homework - 2

①

given,

$$f(x_1, x_2) = 2x_1^2 - 4x_1x_2 + 1.5x_2^2 + x_2$$

now,

$$\text{grad } f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 4x_1 - 4x_2 \\ -4x_1 + 3x_2 + 1 \end{bmatrix}$$

$$\text{now, } f=0 \Rightarrow \begin{bmatrix} 4x_1 - 4x_2 \\ -4x_1 + 3x_2 + 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{aligned} 4x_1 - 4x_2 &= 0 \\ -4x_1 + 3x_2 + 1 &= 0 \end{aligned}$$

$$x_1 = x_2 \Rightarrow \frac{4x_2}{4} \Rightarrow x_1 = x_2$$

so,

$$-4x_1 + 3x_2 + 1 = 0$$

$$-4x_1 + 3x_1 = -1$$

$$x_1(-1) = -1$$

$$\Rightarrow \boxed{x_1 = 1} \text{ so, } \boxed{x_2 = 1}$$

now,

$$H = \begin{bmatrix} 4 & -4 \\ -4 & 3 \end{bmatrix} \text{ now, } |H| = \begin{vmatrix} 4 & -4 \\ -4 & 3 \end{vmatrix} = 4(3) - (-4)(-4) = 12 - 16 = -4 < 0$$

Indefinite eigen value so, saddle point exist.

$$\text{at } f(\vec{x}_1, \vec{x}_2) = f(1, 1) = 2(1)^2 - 4(1)(1) + 1.5(1)^2 + 1 = 0.5$$

The Taylor expansion,

$$f(x) = f_0 + g^T(x - x_0) + \frac{1}{2}(x - x_0)^T H (x - x_0)$$

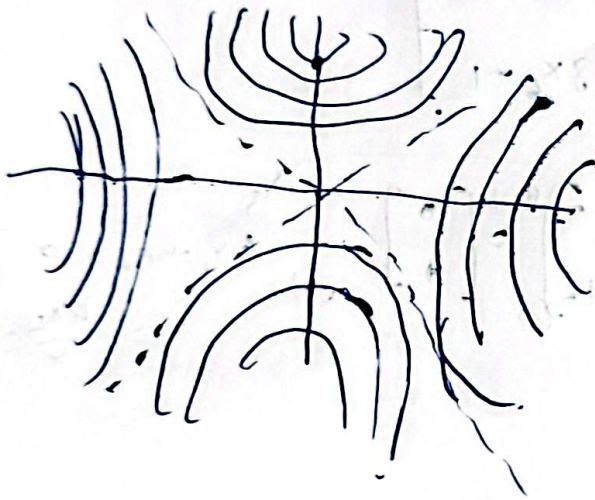
$$= 0.5 + [0 \ 0] (x - x_0) + \frac{1}{2}(x - x_0)^T \begin{bmatrix} 4 & -4 \\ -4 & 3 \end{bmatrix} (x - x_0)$$

$$= 0.5 + (0)(x_1 - 1) + (0)(x_2 - 1) + \frac{1}{2}(x_1 - 1)^2(4) + (x_1 - 1)(x_2 - 1)(-4) + \frac{1}{2}(x_2 - 1)^2(3)$$

$$\Rightarrow f(x) - 0.5 = 2(x_1^2 - 1)(x_2^2 + 1) - 4(x_1 x_2 - x_2 - x_1 + 1) + \frac{3}{2}(x_2^2 - 2x_2 + 1) \leq 0$$

$$0 \geq 2x_1^2 + \frac{3}{2}x_2^2 + x_2 - 4x_1 x_2 + 3$$

se ită zău,



It ca de boli yley

② (a) given

$$x_1 + 2x_2 + 3x_3 = 1$$

$$(-1, 0, 1)^T$$

$$\min_{\vec{x}} d^2 = (x_1 + 1)^2 + x_2^2 + (x_3 - 1)^2$$

$$\text{s.t. } x_1 + 2x_2 + 3x_3 = 1$$

$$f_1 = x_1^2 + 2x_1 + x_2^2 + x_3^2 - 2x_3 + 2$$

$$\nabla f_1 = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \frac{\partial f}{\partial x_3} \end{bmatrix} = \begin{bmatrix} 2x_1 + 2 \\ 2x_2 \\ 2x_3 - 2 \end{bmatrix}$$

$$H = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \quad \lambda = 2, 2, 2 \Rightarrow \text{P.D.}$$

↓
objective is convex

constraint as $x_1 + 2x_2 + 3x_3 = 1 \Rightarrow$ Hyper Plane which is convex

So, the problem is convex.

to convert to unconstrained prob.

$$x_1 = 1 - 2x_2 - 3x_3$$

$$\begin{aligned} f_2 &= (1 - 2x_2 - 3x_3)^2 + 2(1 - 2x_2 - 3x_3) + x_2^2 + x_3^2 - 2x_3 + 2 \\ &= 5x_2^2 + 12x_2x_3 - 8x_2 + 10x_3^2 - 14x_3 + 5 \end{aligned}$$

min f_2
w.r.t x_2, x_3

$$\nabla f_1 = \begin{bmatrix} 10x_2 + 12x_3 - 2 \\ 12x_2 + 20x_3 - 14 \end{bmatrix}$$

$$H = \begin{bmatrix} 10 & 12 \\ 12 & 20 \end{bmatrix} \Rightarrow \lambda = 2, 12 \Rightarrow \text{p.d.}$$

convex prob

So, sol. will be m/n.

$\nabla f_1 = 0$ when

$$10x_2 + 12x_3 - 2 = 0 \Rightarrow x_2 = -1/7$$

$$12x_2 + 20x_3 - 14 = 0 \Rightarrow x_3 = 11/14$$

$$x_1 = 1 - 2(-1/7) - 3(11/14)$$

$$= \frac{-15}{14}$$

$$\bar{x} = \left(\frac{-15}{14}, -1/7, 11/14 \right)$$

(3) Hyperplane of a convex set -

$$a^T x = c \quad \{ a^T y = c$$

no.

consider line segment join x & y by t

$$f(t) = tx + (1-t)y$$

no.

$$a^T f(t) = a^T (tx + (1-t)y)$$

$$= t(a^T x) + (1-t)(a^T y)$$

no.

$$a^T f(t) = t(c) + (1-t)(c)$$

$$[\because a^T x = c, a^T y = c]$$

$$a^T f(t) = tx + (1-t)c$$

$$a^T f(t) = c \rightarrow \text{Hyperplane}$$

$a^T x = c$ is a convex set because it contains all the line segment joining any 2 points within it.

(4) given,

$$\min_p \max_{I_L} \{ h(a^T p, I_L) \}$$

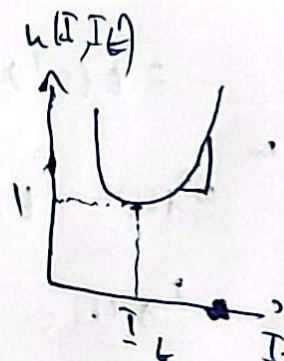
$$\text{St. } 0 \leq p_i \leq p_{\max}$$

$$h(I_L) = \begin{cases} I_L / I & I \leq \bar{I}_L \\ E(I_L) & I \geq \bar{I}_L \end{cases}$$

(a) no,

$$\frac{\partial h}{\partial p} = \frac{dh}{dI} = \frac{\partial(a^T p)}{\partial I} = h' a$$

$$\frac{\partial^2 h}{\partial p^2} = \frac{\partial^2 h}{\partial I^2} = \frac{\partial(a^T p)}{\partial I} a^T = h'' a a^T$$



$$h'' \geq 0 \rightarrow h \text{ is a convex set w.r.t } I$$

$$\begin{cases} I \leq \bar{I}_L \rightarrow h'' > 0 \\ I > \bar{I}_L \rightarrow h'' = 0 \end{cases} \quad \left. \begin{array}{l} \\ \end{array} \right\} h'' \text{ is P.S.D}$$

since $h(I, I_L)$ is a convex set $h(a^T p, I_L)$ is also a

convex set

(b) " $C_n^{16} \left\{ \begin{array}{l} p_1 + p_2 + \dots + p_n \leq p^* \\ p_1 + p_2 + \dots + p_n \leq p^* \end{array} \right\}$

$$[1, \dots, 1, 0, \dots, 0] \begin{bmatrix} p_1 \\ p_2 \\ \vdots \\ p_n \end{bmatrix}$$

linear constraints so, strictly convex so, unique sol.

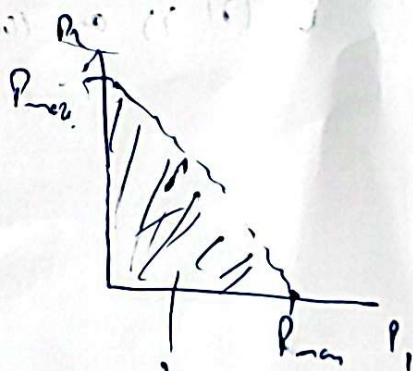
(c) no more than 10 lamps to be switched on.

2 lamps p_1, p_2 but no more than 1 can be on.

The region is not a convex set

→ if expanded to 10 lamps still it will not be a convex set

so may (or) may not have unique sol. not a good region.



(5) Given,

$$C^*(y) = \max_x \{xy - (1+x)\}$$

$$\frac{dC^*(y)}{dy} = x$$

$$\frac{d^2C^*(y)}{dy^2} = 0$$

So,

$C^*(y)$ is a linear function wrt to y .

So, $C^*(y)$ is a convex set wrt to y .

