

Q) Determine whether the following function is Convex:-

$$f(u) = \underbrace{(x_1 - 4x_2)^4}_{\text{convex}} + \underbrace{e^{(9x_1^2 + 6x_1x_2 + 4x_2^2)}}_{\text{convex}}$$

$e^u \rightarrow \text{convex}$

let  $f(u) = f_1(u) + f_2(u)$

$$f_1(u) = (x_1 - 4x_2)^4$$

$$g_1(u) = u^4 \quad (\mathbb{R} \rightarrow \mathbb{R})$$

$$h_1(u) = \underline{x_1 - 4x_2} \quad (\mathbb{R}^2 \rightarrow \mathbb{R})$$

affine (Both convex)  
convex

$$g_1'(u) = 4u^3$$

$$g_1''(u) = 12u^2 \geq 0 \quad (\text{convex})$$

$$\nabla h_1(u) = \begin{bmatrix} 1 \\ -4 \end{bmatrix}$$

$$\nabla^2 h_1(u) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \text{convex}$$

$$f_1(u) = g_1 \circ h_1(u) \quad \text{convex}$$

$(9x_1^2 + 6x_1x_2 + 4x_2^2)$

$$f_2(u) = e$$

$$g_2(u) = e^u \quad (\mathbb{R} \rightarrow \mathbb{R})$$

$$g_2'(u) = e^u; \quad g_2''(u) = e^u > 0$$

convex

$$h_2(u) = \underline{9x_1^2 + 6x_1x_2 + 4x_2^2} \quad (\mathbb{R}^2 \rightarrow \mathbb{R})$$

$$\nabla^2 h = \underline{X^T A X} \quad ; \quad (A)$$

$$= \begin{bmatrix} 9x_1 + 3x_2 & 3x_1 + 4x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$h_2(u) \cong \underline{X^T A X} = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 2.9 & 2.3 \\ 2.3 & 2.4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$h_2(x) \cong (x^T A x) = [x_1, x_2] \begin{bmatrix} 2.9 & 2.3 \\ 2.3 & 2.4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\nabla^T h_2(x) = 2.4$$

$$A' = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\begin{bmatrix} 1.8 & 6 \\ 6 & 8 \end{bmatrix} \text{ convex}$$

$$\text{tra}(A) = a + d < 0$$

$$|A'| > 0 \rightarrow \begin{matrix} +ve \\ -ve \end{matrix}$$

$$f(x) = \underbrace{-(x_1 - 4x_2)^4}_{\text{convex } f_1} + \underbrace{e^{(9x_1^2 + 6x_1x_2 + 4x_2^2)}}_{\text{convex } f_2}$$

(cannot say if it is convex or concave need to check)

$$\nabla^T f(x) = \nabla^T f_1(x) + \nabla^T f_2(x)$$

$$\nabla^T f(x) = (x_1 - 4x_2)^3 \begin{bmatrix} -12 & 48 \\ 48 & -192 \end{bmatrix} + e^{9x_1^2 + 6x_1x_2 + 4x_2^2} \begin{bmatrix} 18x_1 + 6x_2 + 18 & 6x_1 + 8x_2 + 6 \\ 6x_1 + 6x_2 + 6 & 6x_1 + 8x_2 + 6 \end{bmatrix}$$

$$\text{at } (x_1, x_2) = (-1, 1) \quad x_1 = -1; x_2 = 1$$

$$\nabla^T f(x) \Big|_{(-1,1)} < 0 \quad \therefore \text{not convex}$$

Q) find local minima, for the following functions

$$f(x) = \underbrace{3x^4 + 5x^3 - 4x^2 + 2}_{\text{convex}} \quad \text{in the interval } [0.3, 1] \quad \text{Global mini??}$$

$$f'(x) = 12x^3 + 15x^2 - 8x = 0$$

$$x = 0 \text{ (stationary)}$$

$$f''(x) = 36x^2 + 30x - 8$$

$$-30 \pm \sqrt{900 + 4(36)(8)}$$

$$12x^2 + 15x - 8 = 0$$

$$x_1 = \dots$$

$$x_2 = \dots$$

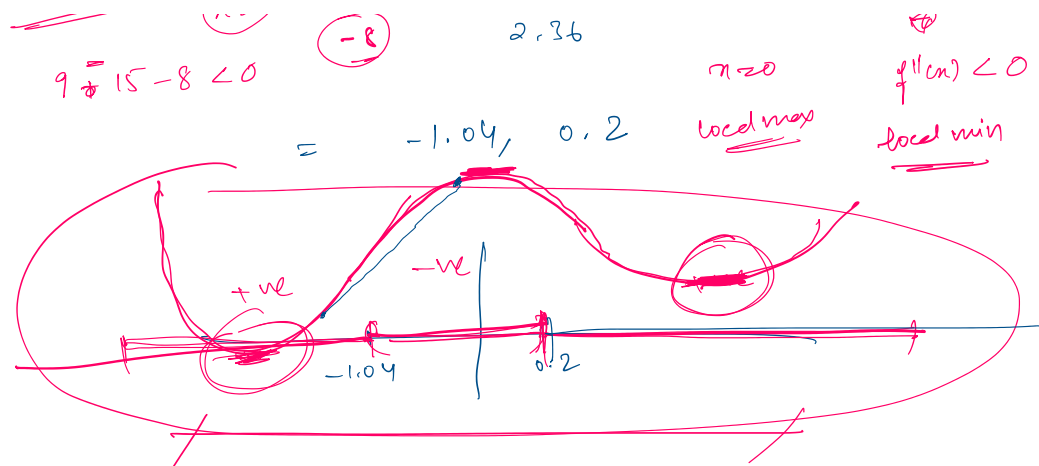
$$x = -0.5$$

$$x = 0 = -8$$

$$9 + 15 - 8 < 0$$

$$x = 0$$

$$f''(x) < 0$$



# TUTORIAL 5

Q1. Tutorial 4:

$$f: (-\infty, \infty) \rightarrow (-\pi/2, \pi)$$

$$f(x) = \tan^{-1}(x)$$

Lipschitz gradient exists

$$f: S \rightarrow \mathbb{R} \quad C^1$$

Defn

$$\| \nabla f(y) - \nabla f(x) \|_2 \leq L \| y - x \|_2$$

for some constant  $L$

$$f(x) = \tan^{-1}(x)$$

$$\frac{d}{dx} f(x) = \frac{1}{1+x^2}$$

$$\| \frac{d}{dx} f(y) - \frac{d}{dx} f(x) \| \leq L \|y - x\|$$

M.V.T  $(y, x) \quad \forall (x, y) \in \mathbb{R}$

$$f'(z) = \frac{f(y) - f(x)}{y - x}$$

$$\left[ \frac{1}{1+y^2} - \frac{1}{1+x^2} \right] = (y-x) f'(z)$$

L.H.S  $\leq L |y - x|$

$$\left| \frac{1}{1+y^2} - \frac{1}{1+x^2} \right| = |f''(z)| |y-x| \rightarrow (1)$$

$z \in \mathbb{R}$

$$\max_{z \in \mathbb{R}} f''(z)$$

$$\max_{z \in \mathbb{R}} \left( \frac{-1(2x)}{(1+x^2)^2} \right) = g(x)$$

first order - necessary - condition,  $g'(x) = 0$

$$g(x) = -2 \left[ \frac{(1+x^2)^2(1) - x(2)(1+x^2)(2x)}{(1+x^2)^4} \right] x$$

$$g'(x) = 0 \Rightarrow \frac{-2}{(1+x^2)^4} [1+x^2] [-4x^2 + x^2 + 1]$$

$$\boxed{x = \pm \frac{1}{\sqrt{3}}}$$

$$(1+x)$$

Second-order-sufficient-condition

$$g''(x) > 0 \quad (\text{Local minima})$$

$$< 0 \quad (\text{maxima})$$

$$g''(x) = \frac{2 \left[ (1+x^2)^4 \left[ (1+x^2)(-6x) + (1-3x^2)(2x) \right] - (1+x^2)(1-3x^2)(4)(1+x^2)^3(2x) \right]}{(1+x^2)^8}$$

$$\begin{aligned} g''(x) &= \frac{-2}{(1+x^2)^8} \left[ (1+x^2)^4 \left[ (1+x^2)(-6x) + (1-3x^2)(2x) \right] - (1-3x^2)(4)(2x) \right] \\ &= \frac{-2}{(1+x^2)^4} \left[ -6x - 6x^3 + (1-3x^2)(-6x) \right] \end{aligned}$$

$$\begin{aligned}
 &= -\frac{2}{(1+x^2)^4} \left( -12x - 6x^3 + 18x^3 \right) \\
 &= -\frac{2}{(1+x^2)^4} (12x^3 - 12x) \\
 &= -\frac{2}{(1+x^2)^4} 12x(x^2 - 1) \\
 &\leq 0
 \end{aligned}$$

$$g'(x^* = -\frac{1}{\sqrt{3}})$$

$x^* = -\frac{1}{\sqrt{3}}$  is a maxima.

$$\begin{aligned}
 f''(x^* = -\frac{1}{\sqrt{3}}) &= \frac{-2x - 10x^3}{(1+x^2)^2} = \frac{-2}{\sqrt{3}} \div \frac{16}{9} = \frac{18}{16\sqrt{3}} \\
 &= \frac{9}{8\sqrt{3}}
 \end{aligned}$$

$$\left| \frac{1}{1+y^2} - \frac{1}{1+x^2} \right| \leq \frac{9}{8\sqrt{3}} (|y-x|)$$

Lipschitz - gradient exists.

Q Tutorial 3 :.

$$-(x_1^2 + x_2^2)$$



Q1. (c)  $f(x) = x_1 x_2 e^{-(x_1^2 + x_2^2)}$

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

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} x_2 \left[ x_1 e^{-(x_1^2 + x_2^2)} (-2x_1) + e^{-(x_1^2 + x_2^2)} (1) \right] \\ x_1 \left[ x_2 e^{-(x_1^2 + x_2^2)} (2x_2) + e^{-(x_1^2 + x_2^2)} (1) \right] \end{bmatrix}$$

$$= \begin{bmatrix} x_2 (1 - 2x_1^2) e^{-(x_1^2 + x_2^2)} \\ (1 - 2x_2^2) x_1 e^{-(x_1^2 + x_2^2)} \end{bmatrix}$$

$$\nabla f = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ (1-2x_1^2)x_2 \\ (1-2x_2^2)x_1 \\ -(x_1^2+x_2^2) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\textcircled{1} \quad x_2(1-2x_1^2) = 0$$

$$x_1(1-2x_2^2) = 0 \quad \textcircled{2}$$

$$x_1 = 0, x_2 = 0$$

$$\left. \begin{aligned} x_1^2 &= 1/2 \\ x_2^2 &= 1/2 \end{aligned} \right\}$$

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix}, \begin{pmatrix} -1/2 \\ -1/2 \end{pmatrix}, \begin{pmatrix} 1/2 \\ -1/2 \end{pmatrix}, \begin{pmatrix} -1/2 \\ 1/2 \end{pmatrix}$$

→ local minima or maxima.

$$x_2(1-2x_1^2) = 0 \rightarrow \textcircled{1}$$

$$x_1(1-2x_2^2) = 0$$

$$x_2 = 0$$

$$x_1^2 = x_2$$

$$x_1 = 0$$

$$x_2^2 = x_2$$

$$\nabla f = \begin{bmatrix} (1-2x_1^2)x_2 e^{-(x_1^2+x_2^2)} \\ (1-2x_2^2)x_1 e^{-(x_1^2+x_2^2)} \end{bmatrix}$$

$$\frac{\partial^2 f}{\partial x_1^2} = x_2 \left[ (1-2x_1^2) e^{-(x_1^2+x_2^2)} (-2x_1) + e^{-(x_1^2+x_2^2)} (-4x_1) \right]$$

$$x_2^2 p = -(x_1^2+x_2^2) / (2x_1^2-3)$$

$$\frac{\partial f}{\partial x_1^2} = e^{-(x_1^2 + x_2^2)}$$

$$\frac{\partial f}{\partial x_2^2} = e^{-(x_1^2 + x_2^2)}$$

$$\frac{\partial^2 f}{\partial x_1 \partial x_2} = (1 - 2x_1^2)(1 - 2x_2^2) e^{-(x_1^2 + x_2^2)}$$

$$\frac{\partial^2 f}{\partial x_2^2} = 2x_1 x_2 e^{-(x_1^2 + x_2^2)} (2x_2^2 - 3)$$

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix}$$

$$\nabla^2 f = \begin{bmatrix} e^{-(x_1^2 + x_2^2)} (2x_1 x_2) (2x_1^2 - 3) & (1 - 2x_1^2)(1 - 2x_2^2) e^{-(x_1^2 + x_2^2)} \\ 2x_1 x_2 e^{-(x_1^2 + x_2^2)} (2x_2^2 - 3) & e^{-(x_1^2 + x_2^2)} (2x_1^2 - 3) \end{bmatrix}$$

$\left( \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$

$$\nabla^2 f = \begin{bmatrix} \frac{2}{e} = \lambda_1 & 0 \\ 0 & \frac{2}{e} = \lambda_2 \end{bmatrix}$$

$$> 0$$

second-order condition,  
{local minima}

Q. Tutorial 3:

Q. 5. (b) convex, strictly convex, strongly  
 $\{x^T A x\}$  convex

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

$$f(x) = (x_1 - 3x_2)^2 + (x_1 - 2x_2)^2$$

$$f(x) = x_1^2 - 6x_1x_2 + 9x_2^2 + x_1^2 - 4x_1x_2 + 4x_2^2$$

$$= 2x_1^2 - 10x_1x_2 + 13x_2^2$$

$$= \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 2 & -5 \\ -5 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$f(x) = x^T A x$$

A is symmetric  
 $A^T = A$

$$\nabla f(x) = 2Ax$$

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

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$$= \begin{bmatrix} 4 & -10 \\ -10 & 26 \end{bmatrix}$$

P.D strictly convex Defn  
 eigen values  $\begin{bmatrix} 4 & -10 \\ -10 & 26 \end{bmatrix} > 0$   
 $\nabla^2 f(x)$   
Strong-convexity

Q. can we find an  $m > 0$

$$\nabla^2 f(x) - mI \geq 0$$

$$\begin{bmatrix} 4 & -10 \\ -10 & 26 \end{bmatrix} - \begin{bmatrix} m & 0 \\ 0 & m \end{bmatrix} \geq 0$$

$\begin{bmatrix} 4-m \\ -10 \end{bmatrix} \quad \begin{bmatrix} -10 \\ 26-m \end{bmatrix} \quad \{m=0.1\}$   
 $f(x)$  is strongly convex with  
parameter  $m=0.1$

Q.) Tutorial 1

(Q2.) verify the parallelogram-Law,

$$\|x+y\|_2^2 + \|x-y\|_2^2 \leq 2(\|x\|_2^2 + \|y\|_2^2)$$

$$\|x\|_2^2 = x^T x$$

L.H.S

$$(x+y)^T(x+y) + (x-y)^T(x-y)$$



$$\begin{aligned}
 &= x^T x + \cancel{x^T y} + \cancel{y^T x} + y^T y \\
 &\quad + x^T x - \cancel{x^T y} - \cancel{y^T x} + y^T y
 \end{aligned}$$

$$\begin{aligned}
 &= 2 \{ x^T x + y^T y \} \\
 &= 2 \{ \|x\|_2^2 + \|y\|_2^2 \}
 \end{aligned}$$

proved.

Q. Tutorial

$$\nabla (x^T A x) = (A + A^T)x,$$

A is not symmetric

$$A = \begin{bmatrix} \leftarrow a_1^T \longrightarrow \\ \leftarrow a_2^T \longrightarrow \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$$

$$\left[ \leftarrow a_n^T \rightarrow \right] \in \mathbb{R}^{n \times n}$$

$$A x = \begin{bmatrix} a_1^T x \\ a_2^T x \\ \vdots \\ a_n^T x \end{bmatrix} \equiv x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

$$x^T A x = x_1 (a_1^T x) + x_2 (a_2^T x) + \dots + x_n (a_n^T x)$$

$$\nabla (x^T A x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix}$$

$$\left[ \begin{aligned} \frac{\partial f}{\partial x_1} &= x_1 [a_{11}] + (a_1^T x) + x_2 (a_{21}) + \dots + x_n (a_{n1}) \\ \frac{\partial f}{\partial x_2} &= x_1 (a_{12}) + x_2 (a_{22}) + (a_2^T x) + \dots + x_n a_{n2} \\ &\vdots \\ \frac{\partial f}{\partial x_n} &= x_1 (a_{1n}) + x_2 a_{n2} + \dots + a_n^T x + x_n a_{nn} \end{aligned} \right]$$

$$= \left[ \begin{aligned} a_{11} x_1 + (a_1^T x) + x_2 a_{21} + \dots + x_n a_{n1} \\ x_1 (a_{12}) + x_2 a_{22} + (a_2^T x) + \dots + x_n a_{n2} \\ x_1 a_{1n} + x_2 a_{n2} + \dots + a_n^T x + x_n a_{nn} \end{aligned} \right] = \nabla f$$

$$= \begin{bmatrix} a_1^T x \\ a_2^T x \\ \vdots \end{bmatrix} + \begin{bmatrix} a_{11} x_1 + x_2 a_{21} + \dots + x_n a_{n1} \\ a_{12} x_1 + x_2 a_{22} + \dots + x_n a_{n2} \\ \dots + x_n a_{nn} \end{bmatrix}$$

$$\begin{aligned}
 & \underbrace{[a_1^T \ x]}_{\text{row vector}} + \underbrace{\begin{bmatrix} a_1^T & a_2^T & \dots & a_n^T \end{bmatrix}}_{\text{matrix of row vectors}} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \\
 & \text{where } a_i^T = (a_{i1}, a_{i2}, \dots, a_{in})
 \end{aligned}$$

$$= Ax + A^T x$$

$$\begin{aligned}
 \forall f &= \underbrace{(A + A^T)}_{A^T = A} x \\
 &= 2Ax
 \end{aligned}$$