

$$1. f(x_1, x_2) = 10 - 2(x_1^4 - 2x_1^2x_2 + x_2^2)$$

$$= -2x_1^4 + 4x_1^2x_2 - 2x_2^2 + 10$$

$$\frac{\partial f(x_1, x_2)}{\partial x_1} = -8x_1^3 + 8x_2x_1, \quad \frac{\partial f(x_1, x_2)}{\partial x_2} = 4x_1^2 - 4x_2$$

$$\frac{\partial^2 f(x_1, x_2)}{\partial x_1^2} = -24x_1^2 + 8x_2, \quad \frac{\partial^2 f(x_1, x_2)}{\partial x_2^2} = -4$$

$$\frac{\partial^2 f(x_1, x_2)}{\partial x_2 \partial x_1} = 8x_1 = \frac{\partial^2 f(x_1, x_2)}{\partial x_1 \partial x_2}$$

Hessian matrix:

$$H = \begin{bmatrix} -24x_1^2 + 8x_2 & 8x_1 \\ 8x_1 & -4 \end{bmatrix}$$

$$|H| = 96x_1^2 - 32x_2 - 64x_1^2 = 32(x_1^2 - x_2)$$

Let,  $x_1 = 0$  and  $x_2 = 1$ , then  $|H| < 0$

So Hessian matrix is not positive semidefinite

Hence,  $f(x_1, x_2)$  is not a convex function over the set  $S$ .

2.(a) By Jensen's inequality:  $f(x) = f\left(\frac{b-x}{b-a} \cdot a + \frac{x-a}{b-a} \cdot b\right) \leq \frac{b-x}{b-a} f(a) + \frac{x-a}{b-a} f(b)$ . ( $\frac{b-x}{b-a} \in [0,1]$ ,  $\frac{x-a}{b-a} \in [0,1]$ ,  $x \in [a,b]$ )

$$(b) f(x) \leq \frac{(b-a)-(x-a)}{b-a} f(a) + \frac{x-a}{b-a} f(b)$$

$$\Rightarrow f(x) - f(a) \leq \frac{x-a}{b-a} (f(b) - f(a))$$

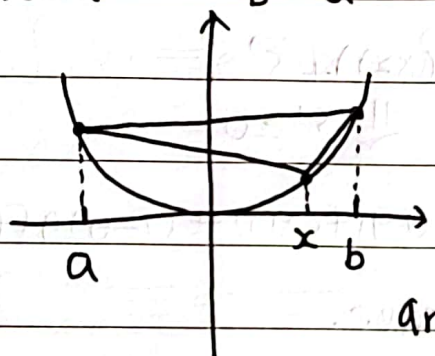
$$\Rightarrow \frac{f(x) - f(a)}{x-a} \leq \frac{f(b) - f(a)}{b-a}$$

$$\text{Similarly, } f(x) \leq \frac{b-x}{b-a} f(a) + \frac{(b-a)-(b-x)}{b-a} f(b)$$

$$\Rightarrow f(b) - f(x) \geq \frac{b-x}{b-a} f(b) - \frac{b-x}{b-a} f(a)$$

$$\Rightarrow \frac{f(b) - f(x)}{b-x} \geq \frac{f(b) - f(a)}{b-a}$$

$$\text{Hence, } \frac{f(x) - f(a)}{x-a} \leq \frac{f(b) - f(a)}{b-a} \leq \frac{f(b) - f(x)}{b-x}$$



the slope of the line ax

is smaller than the line ab's

and the slope of the line ab is

smaller than the line xb's.

(c) let  $x = a + \Delta x$ ,  $\Delta x \rightarrow 0$

then we have  $\frac{f(x) - f(a)}{x-a} = \frac{f(a + \Delta x) - f(a)}{a + \Delta x - a} = f'(a)$ ,

Similarly, we have  $\frac{f(b - \Delta x) - f(b)}{b - \Delta x - b} = f'(b)$

Therefore,  $f'(a) \leq \frac{f(b) - f(a)}{b-a} \leq f'(b)$

(d) let  $b \rightarrow a+$ , then we have:  $f'(a) \leq f'(a+)$

let  $b = a + \Delta x$ ,  $\Delta x \rightarrow 0+$ ,  $\therefore f'(a) \leq f'(a + \Delta x)$

$\Rightarrow \frac{f'(a + \Delta x) - f'(a)}{\Delta x} \geq \frac{0}{\Delta x} = 0$ ,  $\Rightarrow f''(a) \geq 0$

Similarly, we have  $\frac{f'(b + \Delta x) - f'(b)}{\Delta x} \geq 0$ ,  $f''(b) \geq 0$

Therefore,  $f''(a) \geq 0$  and  $f''(b) \geq 0$





### 3. Concave.

we have  $g(f(x)) = x$ , with domain  $(f(a), f(b))$  and  $a < x < b$ .

Consider  $g(\theta f(x_1) + (1-\theta)f(x_2))$ ,  $a < x_1 < x_2 < b$ ,  $0 \leq \theta \leq 1$

Since  $f(x)$  is convex and increasing, we have  $f(\theta x_1 + (1-\theta)x_2) \leq \theta f(x_1) + (1-\theta)f(x_2)$ .

So  $g(\theta f(x_1) + (1-\theta)f(x_2)) = g(f(\theta x_1 + (1-\theta)x_2) + s)$ ,  $s \geq 0$ .

$g(f(\theta x_1 + (1-\theta)x_2) + s) = \theta x_1 + (1-\theta)x_2 + s'$  because  $f(x)$  is increasing,  $f(\theta x_1 + (1-\theta)x_2) + s = f(\theta x_1 + (1-\theta)x_2 + s')$ ,  $s' \geq 0$ .

Finally we get  $g(\theta f(x_1) + (1-\theta)f(x_2)) = \theta x_1 + (1-\theta)x_2 + s'$   
 $= \theta g(f(x_1)) + (1-\theta)g(f(x_2)) + s'$

$$\Downarrow s' \geq 0$$

$g(\theta f(x_1) + (1-\theta)f(x_2)) \geq \theta g(f(x_1)) + (1-\theta)g(f(x_2))$ ,  $0 \leq \theta \leq 1$

Therefore,  $g$  is concave.

4.  $f(v) = \sum_{i=1}^n v_i \log v_i$ ,  $f''(v) = \sum_{i=1}^n (\frac{1}{v_i}) > 0$  because  $u, v \in \mathbb{R}_{++}^n$

So  $f(v)$  is a convex function and  $f(u) \geq f(v) + \nabla f(v)^T (u-v)$   
 (Since it's differentiable).

Therefore, we have  $f(u) - f(v) - \nabla f(v)^T (u-v) \geq 0$ .

Because  $f''(v) > 0$ ,  $f(u) > f(v) + \nabla f(v)^T (u-v)$  when  $u \neq v$ .

$f(u) = f(v) + \nabla f(v)^T (u-v)$  only when  $u = v$ , which means  
 $D_{KL}(u, v) = 0$

Therefore,  $D_{KL} \geq 0$  and  $D_{KL} = 0$  if and only if  $u = v$ .

$$-\frac{1}{x_2} \cdot \frac{1}{x_1} - \frac{1}{x_1} \cdot \frac{1}{x_2} \quad \frac{2}{x_1^2 x_2} \quad \frac{2}{x_2^2 x_1}$$

5. (a)  $f''(x) = e^x > 0$ , convex

(b)  $f''_{x_1 x_1} = f''_{x_2 x_2} = 0$ ,  $f''_{x_1 x_2} = f''_{x_2 x_1} = 1$

Hessian matrix =  $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$|H| < 0$ ,  $\text{Tr}(H) = 0$ , so it's not positive semidefinite nor negative semidefinite

Therefore,  $f$  is neither convex nor concave

(c)  $f''_{x_1 x_1} = \frac{2}{x_1^3 x_2}$ ,  $f''_{x_2 x_2} = \frac{2}{x_2^3 x_1}$ ,  $f''_{x_1 x_2} = f''_{x_2 x_1} = \frac{1}{x_1^2 x_2^2}$

$$H = \begin{bmatrix} \frac{2}{x_1^3 x_2} & \frac{1}{x_1^2 x_2^2} \\ \frac{1}{x_1^2 x_2^2} & \frac{2}{x_2^3 x_1} \end{bmatrix},$$

$|H| = \frac{3}{x_1^4 x_2^4} > 0$ ,  $\text{Tr}(H) > 0$ , positive semidefinite

Therefore,  $f$  is convex.

(d)  $f''_{x_1 x_1} = 0$ ,  $f''_{x_2 x_2} = \frac{2x_1}{x_2^3}$ ,  $f''_{x_1 x_2} = f''_{x_2 x_1} = -\frac{1}{x_2^2}$

$$H = \begin{bmatrix} 0 & -\frac{1}{x_2^2} \\ -\frac{1}{x_2^2} & \frac{2x_1}{x_2^3} \end{bmatrix}, \quad |H| < 0, \text{Tr}(H) > 0$$

$H$  is neither positive nor negative semidefinite

Therefore,  $f$  is neither convex nor concave

(e)  $f''_{x_1 x_1} = \alpha(\alpha-1)x_1^{\alpha-2}x_2^{1-\alpha}$ ,  $f''_{x_2 x_2} = (1-\alpha)(1-\alpha)x_1^\alpha x_2^{-\alpha-1}$

$f''_{x_1 x_2} = f''_{x_2 x_1} = \alpha(1-\alpha)x_1^{\alpha-1}x_2^{-\alpha}$

$|H| = 0$ ,  $\text{Tr}(H) < 0$ ,  $H$  is negative semidefinite

Therefore,  $f$  is concave.