

①

a) Convexity is generally defined as:

$$f(\alpha u + \beta y) \leq \alpha f(u) + \beta f(y)$$

for all  $u, y \in \mathbb{R}^n$  and  $\alpha, \beta \in \mathbb{R}$  with  $\alpha + \beta = 1$  and  $\alpha, \beta \geq 0$ .

One of the benefits of convexity is that it guarantees the existence of local minima. Also, Convex Optimization Problems can be solved efficiently using algorithmic approaches.

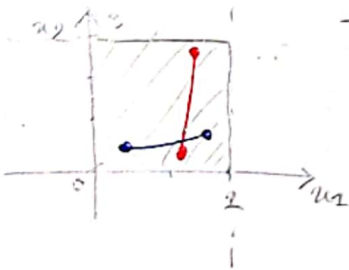
b) A set  $C$  is convex if the line segment between any two points in  $C$  lies in  $C$ . e.g.: for any  $x, y \in C$  and any  $\theta \in [0, 1]$  we have

$$\theta y + (1 - \theta)x \in C$$

A function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex if  $\text{dom } f$  is convex set and if for all  $x, y \in \text{dom } f$  and  $\theta$  with  $0 \leq \theta \leq 1$  we have

$$f(\theta x + (1 - \theta)y) \leq \theta f(x) + (1 - \theta)f(y)$$

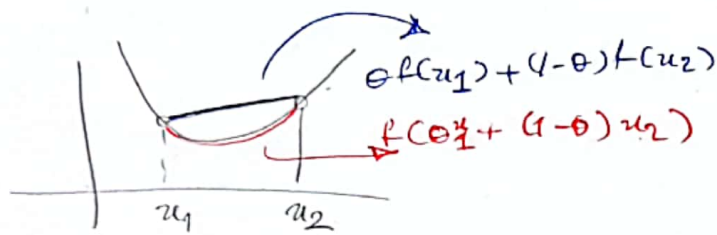
c)



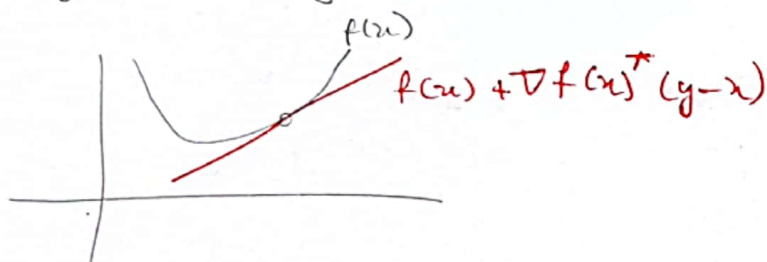
The colored area is the set. As we can see, a line between any two points of  $D$ , is also inside  $D$ .

①

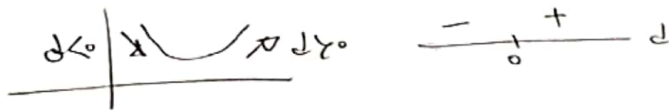
d) i) Jensen's inequality shows that the function is always lower/less than the interpolation (straight line) between two points in the domain of that function.



ii) This shows that any first order Taylor approximation of  $f$  near  $x$ , namely tangent lines in  $x$ , is lesser than the actual function. e.g.: It is a global underestimator.



iii) This shows that the derivative is non-decreasing. This means that the derivative curve is a non-decreasing curve and the function itself, is a curve with this around the minima:



②

i) Theorem 1) a matrix is positive definite if and only if the determinant of each square matrix in its upper left are all positive.

②

$$\frac{df}{dx} = 2ax - 4ay$$

$$\frac{df}{dx} = 2a, \frac{d^2f}{dx^2} = -4a, \frac{d^3f}{dx^3} = 0$$

$$\frac{df}{dy} = 2y - 4ax + 2z$$

$$\frac{d^2f}{dydx} = -4a, \frac{d^2f}{dy^2} = 2, \frac{d^3f}{dydz} = 2$$

$$\frac{df}{dz} = 4z + 2y$$

$$\frac{d^2f}{dzdx} = 0, \frac{d^2f}{dzdy} = 2, \frac{d^3f}{dz^2} = 4$$

$$H_f = \begin{vmatrix} 2a & -4a & 0 \\ -4a & 2 & 2 \\ 0 & 2 & 4 \end{vmatrix}$$

(i) it is symmetric ✓

(ii) by theorem 1:

$$2a > 0 \Rightarrow a > 0 \text{ (I)}$$

$$\det \begin{vmatrix} 2a & -4a \\ -4a & 2 \end{vmatrix} > 0 \Rightarrow 4a - 16a^2 > 0 \Rightarrow a(4 - 16a) > 0 \Rightarrow 0 < a < \frac{1}{4} \text{ (II)}$$

$$\det(H_f) > 0 \Rightarrow 16a + 0 + 0 - 0 - 8a - 64a^2 > 0 \Rightarrow 8a(1 - 8a) > 0$$

$$\Rightarrow 0 < a < \frac{1}{8} \text{ (III)}$$

$$\text{(I), (II), (III)} \Rightarrow 0 < a < \frac{1}{8}$$

$$(ii) \frac{df}{dx} = 8ax + 8y$$

$$\frac{df}{dx} = 8a, \frac{d^2f}{dx^2} = 8$$

$$\frac{df}{dy} = 8x + 2by$$

$$\frac{df}{dydx} = 8, \frac{d^2f}{dy^2} = 2b$$

$$H_f = \begin{pmatrix} 8a & 8 \\ 8 & 2b \end{pmatrix}$$

i) it is symmetric ✓

ii) (theorem 1)

$$8a > 0 \Rightarrow a > 0 \text{ (I)}$$

$$16ab - 64 > 0 \Rightarrow ab > 4 \text{ (II)} \left\{ \begin{array}{l} \text{I, II} \Rightarrow a > 0, b > \frac{4}{a} \end{array} \right.$$

⑥

$$i) \frac{df}{dx} = \frac{1}{\sqrt{xy}} \times \frac{y}{2\sqrt{xy}} = \frac{1}{2x}$$

$$\frac{df}{dx} = \frac{-1}{2x^2}, \quad \frac{df}{dx} = 0$$

$$\frac{df}{dy} = \frac{1}{\sqrt{xy}} \times \frac{x}{2\sqrt{xy}} = \frac{1}{2y}$$

$$\frac{df}{dy} = 0, \quad \frac{df}{dy} = \frac{-1}{2y^2}$$

$$H_f = \begin{pmatrix} -\frac{1}{2x^2} & 0 \\ 0 & -\frac{1}{2y^2} \end{pmatrix}$$

i) symmetric ✓

ii) theorem 2: if the det of square matrixes from left hand side alternate signs, then the matrix is negative definite. ②

$$\det(-\frac{1}{2x^2}) < 0$$

$$\det(H_f) = \frac{1}{4x^2y^2} > 0 \quad \text{①}$$

①, ②  $\rightarrow f$  is concave.

$$ii) \frac{df}{dx} = 2(x-1) + y^2$$

$$\frac{df}{dx} = 2, \quad \frac{df}{dx} = 2y$$

$$\frac{df}{dy} = 2xy$$

$$\frac{df}{dy} = 2y, \quad \frac{df}{dy} = 2x$$

$$H_f = \begin{pmatrix} 2 & 2y \\ 2y & 2x \end{pmatrix}$$

i) it is symmetric ✓

ii) theorem 1):

$2 > 0$  ✓

$$\det(H_f) > 0 \Rightarrow 4x - 4y^2 > 0 \Rightarrow x > y^2 \Rightarrow \text{dom } f = \{(x,y) \in \mathbb{R}^2 \mid x > y^2\}$$

④