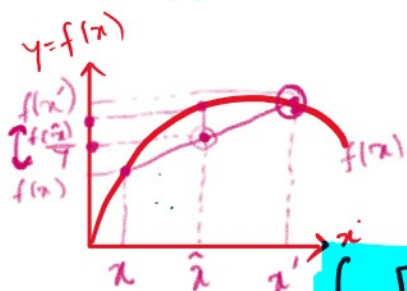


① Concave Function:

Let $f: S \rightarrow \mathbb{R}$ be a function defined on the convex set $S \subset \mathbb{R}^n$.

Then f is concave on the set S if for all $x, x' \in S$ and $\lambda \in [0, 1]$ we have

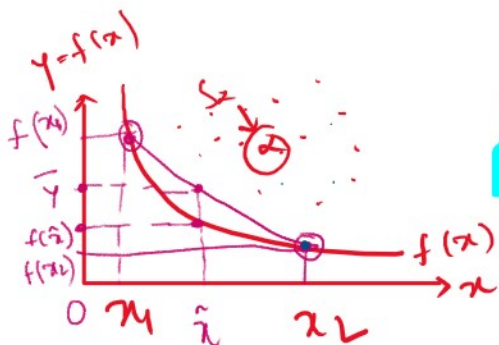
$$f[\lambda x + (1-\lambda)x'] \geq \lambda f(x) + (1-\lambda)f(x')$$



② Convex function:

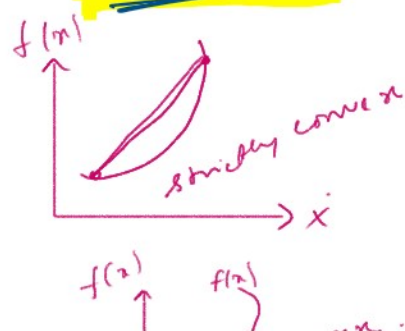
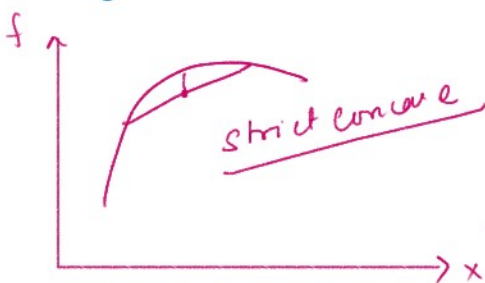
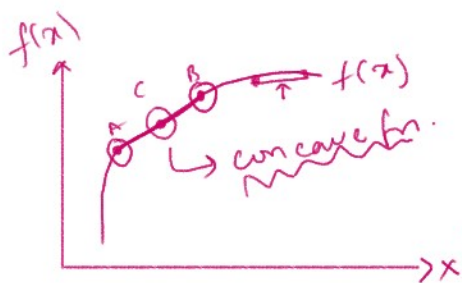
A function $f: S \rightarrow \mathbb{R}$ defined on convex set $S \subset \mathbb{R}^n$; for all x_1, x_2 in its domain and $\lambda \in [0, 1]$ we have

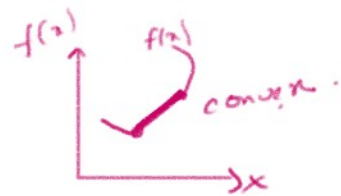
$$f[\lambda x_1 + (1-\lambda)x_2] \leq \lambda f(x_1) + (1-\lambda)f(x_2)$$



Upper bound: Upper bound for a function f is a number ' U ' so that for all x , we have $f(x) \leq U$

Lower bound: Lower bound for a function f is a number ' L ' so that for all x , we have $f(x) \geq L$.





③ Quasiconcave and Quasiconvex fn

Quasiconcave Functions:

Let $f: S \rightarrow \mathbb{R}$ be a fn defined on the convex set $S \subset \mathbb{R}^n$. Then f is quasiconcave if every upper contour set of f is convex i.e.,

$$P_a = \{x \in S : f(x) \geq a\} \text{ is a convex set } \forall a \in \mathbb{R}$$

In other words: $f(x)$ is quasiconcave if $f(\hat{x}) \geq \min[f(x_0), f(x_1)]$

where $\hat{x} = \lambda x_0 + (1-\lambda)x_1$ and $\lambda \in [0, 1]$

Rough explanation:
 $x \in [a, b]$
 $y = f(x)$

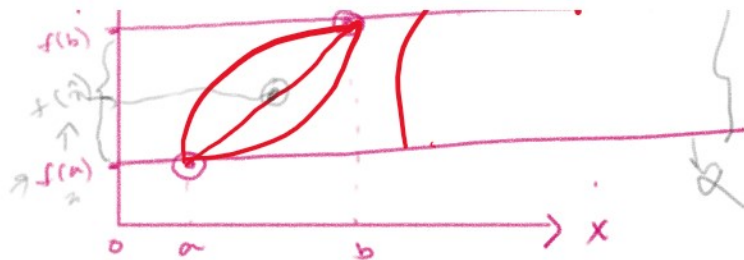
Case (i) $f(a) < f(b)$
 $(\text{or } f(b) < f(a))$

$$f(\lambda a + (1-\lambda)b) \geq \min[f(a), f(b)]$$



Concave \rightarrow quasi-concave
 Convex \rightarrow quasi-concave

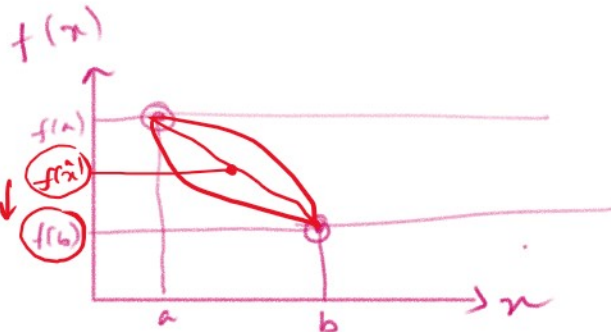
Case (i) $f(a) < f(b)$
 $\min\{f(a), f(b)\}$
 $f(x)$



Case (ii)

$f(b) < f(a)$

$f(x) > f(b)$



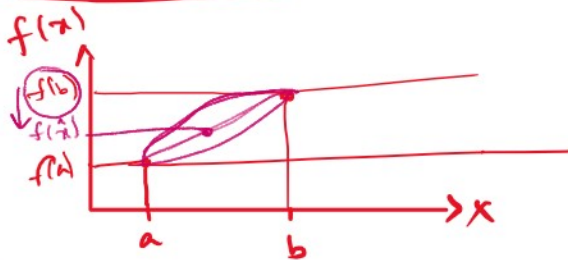
What is quasi convex fn: Let $f: S \rightarrow \mathbb{R}$ where $S \subset \mathbb{R}^n$ is a non empty convex set. The fn f is said to be quasi convex if for each $x_1, x_2 \in S$, we have
 $f(\lambda x_1 + (1-\lambda)x_2) \leq \max\{f(x_1), f(x_2)\}$

ie, iff $S_\alpha = \{x \in S : f(x) \leq \alpha\}$ is convex for each real number.

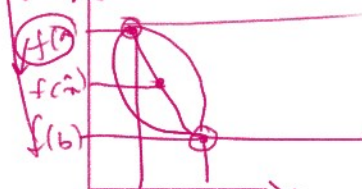
Rough $x \in [a, b] ; f(x) = y$

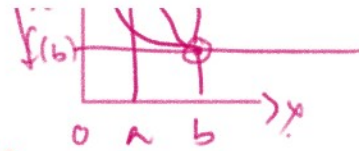
Quasi convex $\Rightarrow f[(1-\lambda)a + \lambda b] \leq \max\{f(a), f(b)\}$

Case 1: $f(a) < f(b)$
 $f((1-\lambda)a + \lambda b) \leq f(b)$



Case (ii)
 $f(a) > f(b)$
 $f(x) \leq f(a)$





Theorem 1: A convex function is always quasi-convex,
but the converse is not true.

[it means a quasi-convex can be concave]

Theorem 2: A concave function is always quasi-concave,
but the converse is not true.

[it means a quasi-concave can be convex].

Theorem 3: $F(g(x))$ is quasi-concave iff $g(x)$ is quasi-concave and F is strictly increasing
(ex: exponential fn)

Theorem 4: $f(g(x))$ is quasi-convex iff $g(x)$ is quasi-convex and f is strictly increasing.

Ex: $z = e^{-x^2}$. Prove whether quasi-concave or not.

let $g(x) = -x^2 \rightarrow g'(x) = -2x$
 $\rightarrow g''(x) = -2 < 0$
 $\rightarrow g(x)$ is concave

~~from diagram that $g(x)$ is concave.~~



we know a concave fn is also quasi-concave
 $\therefore g(x)$ is quasi-concave.

Now $F(g(x)) = F(u) = e^u$

exponential fn is strictly \uparrow sing i.e. F is strictly \uparrow sing.

Hence $z = F(g(x)) = e^{-x^2}$

is quasi-concave.

(Proved)