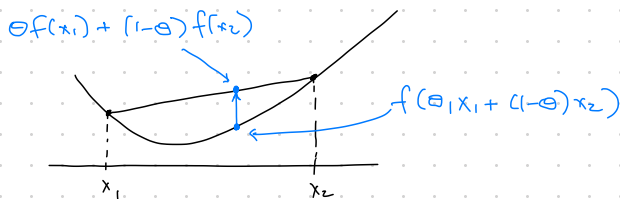


## Convex Function Basics

**Definition**: A function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is **convex** if for  $0 \leq \theta \leq 1$ ,

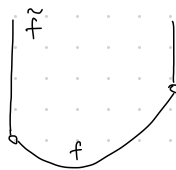
$$f(\theta x_1 + (1-\theta)x_2) \leq \theta f(x_1) + (1-\theta)f(x_2)$$



The function is **strictly convex** if the inequality is strict  $\forall x_1 \neq x_2$ .

**Extended Value Functions**: It's convenient to have  $f$  be defined over all  $\mathbb{R}^n$ . We can extend any convex function  $f$  w/  $\text{dom } f \subset \mathbb{R}^n$  to  $\tilde{f}: \mathbb{R}^n \rightarrow \mathbb{R}$ :

$$\tilde{f}(x) = \begin{cases} f(x) & \text{if } x \in \text{dom } f \\ \infty & \text{else} \end{cases}$$



allows us to just worry about  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ , eg. in this thm.

w/ the convention  $\infty \leq \infty$ ,  $\tilde{f}$  is still convex.

**Thm**: A function is Convex iff it is convex along all lines emanating from any point in the domain, i.e.: "convexity is defined over lines".

$$f: \mathbb{R}^n \rightarrow \mathbb{R} \text{ cvx iff } g(t) = f(x + tv) \text{ cvx } \forall x, v \in \mathbb{R}^n, t \in \mathbb{R}.$$

pf: ( $\Rightarrow$ )  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  cvx.

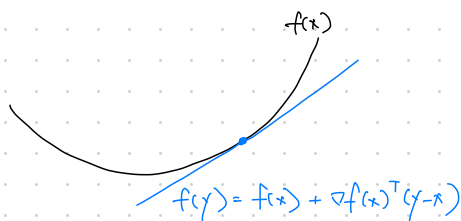
$$\begin{aligned} g(\theta t_1 + (1-\theta)t_2) &= f(x + (\theta t_1 + (1-\theta)t_2)v) \\ &= f(\theta(x + t_1v) + (1-\theta)(x + t_2v)) \\ &\leq \theta f(x + t_1v) + (1-\theta)f(x + t_2v) \\ &= \theta g(t_1) + (1-\theta)g(t_2) \end{aligned}$$

( $\Leftarrow$ )  $g(t) = f(x + tv)$  is cvx.

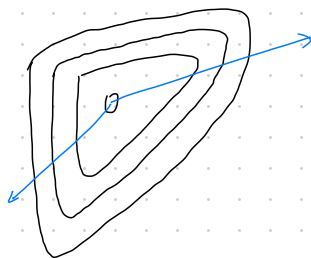
$$\begin{aligned} g(\theta t_1 + (1-\theta)t_2) &, \quad 0 \leq \theta \leq 1 \\ &= f(x_1 + \theta(x_2 - x_1)) \\ &\leq \theta g(t_1) + (1-\theta)g(t_2) \\ &= \theta f(x_2) + (1-\theta)f(x_1) \end{aligned}$$

**First-Order Condition:**  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex iff the 1-st order Taylor is a global underestimator  $\forall x \in \mathbb{R}^n$ .

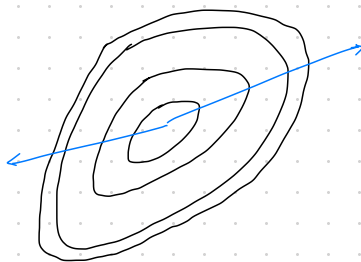
$$f(y) \geq f(x) + \nabla f(x)^T (y-x) \quad \forall x, y$$



**Second-Order Condition:**  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex iff it has positive Curvature everywhere, ie.  $\nabla^2 f(x) \succeq 0 \quad \forall x$ .



cvx

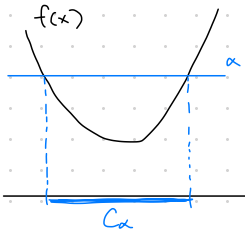


not cvx

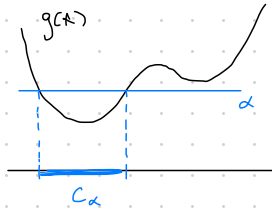
Geometrically, gaps between contour lines in any dir should be monotonically increasing (convex) or decreasing (concave).

**Sublevel Sets**: A sublevel set of  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is defined as:

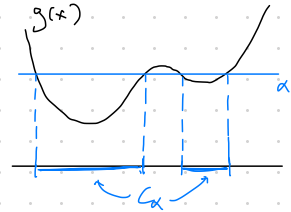
$$C_\alpha = \{x \in \text{dom } f \mid f(x) \leq \alpha\}$$



Convex



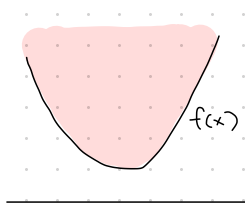
not Convex.



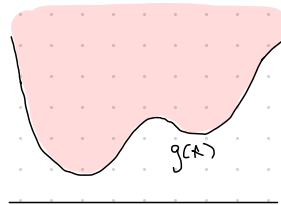
If  $f$  is convex,  $C_\alpha$  is a convex set. The converse is not true.

**Epigraph**: The epigraph of  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is defined as:

$$\text{epi } f = \{(x, t) \in \mathbb{R}^{n+1} \mid t \geq f(x)\}$$



Convex



not Convex.

$f: \mathbb{R}^n \rightarrow \mathbb{R}$  is convex iff  $\text{epi } f$  is a convex set.