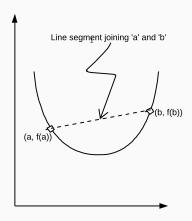
Convex Functions

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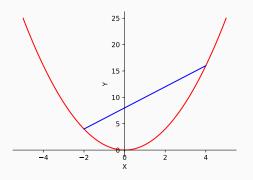
Definition

- · Convexity is defined on an interval $[\alpha, \beta]$
- The line segment joining (a, f(a)) and (b, f(b)) should be above or on the function f for all points in interval $[\alpha, \beta]$.

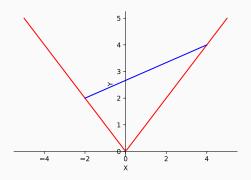


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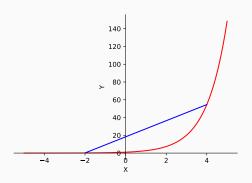
Convex on the entire real line i.e. $(-\infty, \infty)$



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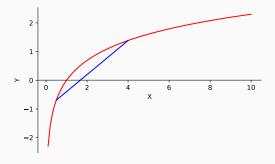


Convex on the entire real line i.e. $(-\infty, \infty)$

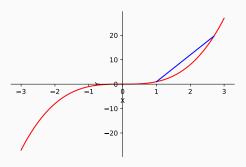


Example: $y = log_e x$

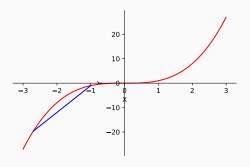
Not convex on the entire real line i.e. $(-\infty, \infty)$



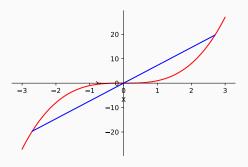
It is convex for the interval $[0, \infty)$



It is concave for the interval $(-\infty, 0]$



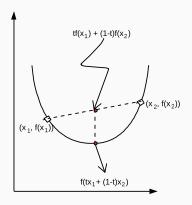
But, it is not convex for the interval $(-\infty, \infty)$



Mathematical Formulation

Function f is convex on set X, if $\forall x_1, x_2 \in X$ and $\forall t \in [0, 1]$

$$f(tx_1 + (1-t)x_2) \le tf(x_1) + (1-t)f(x_2)$$



To prove:

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$$LHS = f(tx_1 + (1-t)x_2) = t^2x_1^2 + (1-t)^2x_2^2 + 2t(1-t)x_1x_2$$

$$RHS = tf(x_1) + (1-t)f(x_2) = tx_1^2 + (1-t)x_2^2$$

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RHS = $tf(x_1) + (1 - t)f(x_2) = tx_1^2 + (1 - t)x_2^2$

Here,

LHS - RHS =
$$(t^2 - t)x_1^2 + [(1 - t)^2 - (1 - t)]x_2^2 + 2t(1 - t)x_1x_2$$

= $(t^2 - t)x_1^2 + (t^2 - t)x_2^2 - 2(t^2 - t)x_1x_2$
= $(t^2 - t)(x_1 - x_2)^2$

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= $(t^2 - t)(x_1 - x_2)^2$

Here, $(t^2-t) \le 0$ since $t \in [0,1]$ and $(x_1-x_2)^2 \ge 0$ Hence, LHS -RHS ≤ 0 Hence LHS \le RHS Hence proved.

The Double-Derivative Test

If f''(x) > 0, the function is convex.

For example,

$$\frac{\partial^2(x^2)}{\partial x^2} = 2 > 0 \Rightarrow x^2 \text{ is a Convex function.}$$

The double derivate test for multi-parameter function is equal to using the Hessian Matrix

A function $f(x_1, x_2, ..., x_n)$ is convex iff its $n \times n$ Hessian Matrix is positive semidefinite for all posible values of $(x_1, x_2, ..., x_n)$

$$\mathbf{H} = \begin{bmatrix} \frac{\partial^2 f}{\partial \mathbf{x}_1^2} & \frac{\partial^2 f}{\partial \mathbf{x}_1 \partial \mathbf{x}_2} & \cdots & \frac{\partial^2 f}{\partial \mathbf{x}_1 \partial \mathbf{x}_n} \\ \frac{\partial^2 f}{\partial \mathbf{x}_2 \partial \mathbf{x}_1} & \frac{\partial^2 f}{\partial \mathbf{x}_2^2} & \cdots & \frac{\partial^2 f}{\partial \mathbf{x}_2 \partial \mathbf{x}_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial \mathbf{x}_n \partial \mathbf{x}_1} & \frac{\partial^2 f}{\partial \mathbf{x}_n \partial \mathbf{x}_2} & \cdots & \frac{\partial^2 f}{\partial \mathbf{x}_n^2} \end{bmatrix}$$

Show that
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 is convex.

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$$H = \begin{bmatrix} \frac{\partial^{2}(x_{1}^{2} + x_{2}^{2})}{\partial x_{1}^{2}} & \frac{\partial^{2}(x_{1}^{2} + x_{2}^{2})}{\partial x_{1}\partial x_{2}} \\ \frac{\partial^{2}(x_{1}^{2} + x_{2}^{2})}{\partial x_{2}\partial x_{1}} & \frac{\partial^{2}(x_{1}^{2} + x_{2}^{2})}{\partial x_{2}^{2}} \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

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Eigen Values of H are 2 and $2 > 0 \Rightarrow$ H is positive semi-definite. Hence, $f(x_1, x_2) = x_1^2 + x_2^2$ is convex.

Properties of Convex Functions

- If f(x) is convex, then kf(x) is also convex, for some constant k
- If f(x) and g(x) are convex, then f(x) + g(x) is also convex.