

VE414 Appendix 9

Jing Liu

UM-SJTU Joint Institute

June 27, 2019

Jensen's Inequality

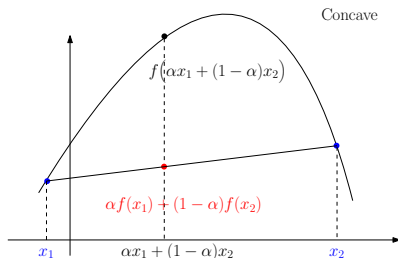
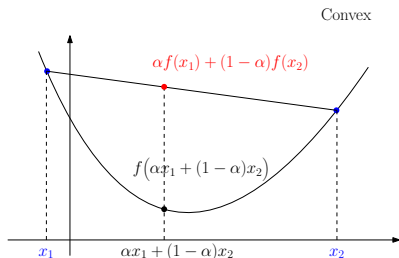
Definition

A function $f: \mathcal{S} \rightarrow \mathbb{R}$ is said to be **convex** if, for every $x_1, x_2 \in \mathcal{S}$, it satisfies

$$f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2) \quad \text{for all } \alpha \in [0, 1]$$

It is said to be **concave** if, for every $x_1, x_2 \in \mathcal{S}$, it satisfies

$$f(\alpha x_1 + (1 - \alpha)x_2) \geq \alpha f(x_1) + (1 - \alpha)f(x_2) \quad \text{for all } \alpha \in [0, 1]$$



Jensen's Inequality

Suppose $g: \mathcal{D} \rightarrow \mathcal{S}$, where \mathcal{D} is the support of a random variable \mathbf{Y} , and $\mathcal{S} \subset \mathbb{R}$.

- If $f: \mathcal{S} \rightarrow \mathbb{R}$ is convex, then

$$\mathbb{E}[(f \circ g)(\mathbf{y})] \geq f\left(\mathbb{E}[g(\mathbf{y})]\right)$$

- If $f: \mathcal{S} \rightarrow \mathbb{R}$ is concave, then

$$\mathbb{E}[(f \circ g)(\mathbf{y})] \leq f\left(\mathbb{E}[g(\mathbf{y})]\right)$$

Proof

Let $\ell(z) = a + b \cdot z$ be a line in \mathbb{R}^2 tangent to $f(z)$ at the point $z = \mathbb{E}[g(\mathbf{y})]$.

$$\mathbb{E}[(f \circ g)(\mathbf{y})] \geq \mathbb{E}[(\ell \circ g)(\mathbf{y})] = a + b\mathbb{E}[g(\mathbf{y})] = \ell(\mathbb{E}[g(\mathbf{y})]) = f\left(\mathbb{E}[g(\mathbf{y})]\right)$$

where the first inequality is due to convexity.

[Back](#)