

CMO: Convex functions

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Definition 1 (Convex set). Let $S \subseteq \mathbb{R}^d$, S is convex iff

$$\forall u, v \in S, \forall \alpha \in (0, 1), (1 - \alpha)u + \alpha v \in S$$

Definition 2 (Convex function). Let $f : S \mapsto \mathbb{R}$, where S is convex. f is convex iff

$$\forall \alpha \in (0, 1), \forall u, v \in S, f((1 - \alpha)u + \alpha v) \leq (1 - \alpha)f(u) + \alpha f(v)$$

Theorem 1. f is convex $\iff (\forall u, v \in S, \nabla_f(u)^T(v - u) \leq f(v) - f(u))$

Proof.

f is convex

$$\begin{aligned} &\Rightarrow f((1 - \alpha)u + \alpha v) \leq (1 - \alpha)f(u) + \alpha f(v) \\ &\Rightarrow f(u + \alpha(v - u)) \leq f(u) + \alpha(f(v) - f(u)) \\ &\Rightarrow \frac{1}{\alpha}(f(u + \alpha(v - u)) - f(u)) \leq f(v) - f(u) \end{aligned}$$

Let $g(\alpha) = f(u + \alpha(v - u))$.

f is convex

$$\begin{aligned} &\Rightarrow \frac{g(\alpha) - g(0)}{\alpha} \leq g(1) - g(0) \\ &\Rightarrow \lim_{\alpha \rightarrow 0} \frac{g(\alpha) - g(0)}{\alpha} \leq \lim_{\alpha \rightarrow 0} (g(1) - g(0)) \\ &\Rightarrow g'(0) \leq g(1) - g(0) \\ &\Rightarrow \nabla_f(u)^T(v - u) \leq f(v) - f(u) \end{aligned}$$

Suppose $\forall u, v \in S, \nabla_f(u)^T(v - u) \leq f(v) - f(u)$.

For any arbitrarily chosen x_1 and x_2 ($x_1 \neq x_2$), let $x = (1 - \alpha)x_1 + \alpha x_2$. Then $x_1 - x = \alpha(x_1 - x_2)$ and $x_2 - x = (1 - \alpha)(x_2 - x_1)$.

Setting $u = x$ and $v = x_1$, we get

$$\nabla_f(x)^T \alpha(x_1 - x_2) \leq f(x_1) - f(x)$$

Setting $u = x$ and $v = x_2$, we get

$$-\nabla_f(x)^T (1 - \alpha)(x_1 - x_2) \leq f(x_2) - f(x)$$

Adding these equations with weights $1 - \alpha$ and α , we get

$$\begin{aligned} & (1 - \alpha) \nabla_f(x)^T \alpha(x_1 - x_2) - \alpha \nabla_f(x)^T (1 - \alpha)(x_1 - x_2) \\ & \leq (1 - \alpha)(f(x_1) - f(x)) + \alpha(f(x_2) - f(x)) \\ & \Rightarrow 0 \leq (1 - \alpha)f(x_1) + \alpha f(x_2) - f(x) \\ & \Rightarrow f((1 - \alpha)x_1 + \alpha x_2) \leq (1 - \alpha)f(x_1) + \alpha f(x_2) \\ & \Rightarrow f \text{ is convex} \end{aligned}$$

□

Theorem 2. *If f is convex, and x^* is a local minimum, then x^* is also a global minimum.*

Proof. For all $x \in \mathbb{R}^d$,

$$0 = f(x^*)^T(x - x^*) \leq f(x) - f(x^*)$$

□

Theorem 3 (Proof omitted). *Let $f : \mathbb{R}^d \mapsto \mathbb{R}$ and $f \in C^2$. Then f is convex iff H_f is positive semi-definite.*