

1. Consider a differentiable function $f : \mathbb{R} \rightarrow \mathbb{R}$ with $\mathbb{R}_+ \subseteq \text{dom } f$, and its running average, defined as

$$F(x) = \frac{1}{x} \int_0^x f(t) dt \quad (1)$$

with $\text{dom } F = \mathbb{R}_{++}$. Show that $F(x)$ is convex if $f(x)$ is convex.

Solution: It is given that $f(x)$ is convex. Therefore, it must hold that the affine transform $f(sx)$ is convex. Hence,

$$w(x) = \int_0^1 f(sx) ds \quad (2)$$

is also convex, as it is the limit of a sum of convex functions. Next, let us substitute $t = sx$ so that $dt = xds$. Hence

$$w(x) = \frac{1}{x} \int_0^x f(t) dt \quad (3)$$

is also convex. Note that this approach allows us to generalize the problem statement by following the steps. For instance, since $f(x)$ is convex, we also have that $h(s)f(sx)$ is convex for any $h(s) > 0$, or any other transformation. Similarly each step can be generalized.

2. Show that the running average of the non-differentiable convex function $f(x)$ is also convex. Since f is not differentiable, you must use the zeroth order condition to prove the convexity of F .

Solution: Running average of the function f is

$$F(\mathbf{x}) = \frac{1}{x} \int_0^x f(\tau) d\tau.$$

By changing variable as $\tau = xt$, we write

$$F(\mathbf{x}) = \int_0^1 f(xt) dt.$$

For any $0 < \theta < 1$,

$$\begin{aligned} F(\theta x + (1 - \theta)y) &= \int_0^1 f(\theta xt + (1 - \theta)yt) dt \\ &\leq \int_0^1 (\theta f(xt) + (1 - \theta)f(yt)) dt \\ &= \theta \int_0^1 f(xt) dt + (1 - \theta) \int_0^1 f(yt) dt \\ &= \theta F(x) + (1 - \theta)F(y) \end{aligned}$$

3. Show that the function $f(\mathbf{x})$ is convex if and only if the function $f(\mathbf{a} + t\mathbf{b})$ is convex for all $\mathbf{a} + t\mathbf{b} \in \text{dom } f$ and t .

Solution: If $f(\mathbf{x})$ is convex then, $f(\mathbf{a} + t\mathbf{b})$ is convex because affine transformation preserves convexity. Now assume $g(t) = f(\mathbf{a} + t\mathbf{b})$ is convex in t , i.e., it holds that for any $\theta \in [0, 1]$:

$$g(\theta t_1 + (1 - \theta)t_2) \leq \theta g(t_1) + (1 - \theta)g(t_2) \quad (4)$$

$$\Rightarrow f(\mathbf{a} + (\theta t_1 + (1 - \theta)t_2)\mathbf{b}) \leq \theta f(\mathbf{a} + t_1\mathbf{b}) + (1 - \theta)f(\mathbf{a} + t_2\mathbf{b}) \quad (5)$$

$$\Rightarrow f(\theta(\mathbf{a} + t_1\mathbf{b}) + (1 - \theta)(\mathbf{a} + t_2\mathbf{b})) \leq \theta f(\mathbf{a} + t_1\mathbf{b}) + (1 - \theta)f(\mathbf{a} + t_2\mathbf{b}) \quad (6)$$

for all \mathbf{a} and \mathbf{b} and $t_1, t_2 \in \mathbb{R}$.

Since the statement is true for all \mathbf{a} , \mathbf{b} , t_1 , and t_2 . Let us consider two arbitrary points $\mathbf{x}, \mathbf{y} \in \text{dom}(f)$. Choose $\mathbf{a} = \mathbf{x}$, $\mathbf{b} = \mathbf{y} - \mathbf{x}$, $t_1 = 0$, and $t_2 = 1$, then it can be seen that $\mathbf{x} = \mathbf{a} + t_1\mathbf{b}$ and $\mathbf{y} = \mathbf{a} + t_2\mathbf{b}$, so that we get:

$$f(\theta\mathbf{x} + (1 - \theta)\mathbf{y}) \leq \theta f(\mathbf{x}) + (1 - \theta)f(\mathbf{y}) \quad (7)$$

This statement also holds for all \mathbf{x}, \mathbf{y} , and $\theta \in [0, 1]$, implying that f is convex.

4. Show that the harmonic function: $f(\mathbf{x}) = (\sum_{i=1}^n x_i^a)^{1/a}$ for $a < 1$, and $a \neq 0$ is concave.

Solution: Denote $w = \sum_{i=1}^n x_i^a$. Elements of Hessian matrix are,

$$[\nabla^2 f(\mathbf{x})]_{ii} = (1 - a)w^{1/a-2}x_i^{2a-2}(1 - x_i^{-a}w)$$

$$[\nabla^2 f(\mathbf{x})]_{ij} = (1 - a)w^{1/a-2}x_i^{a-1}x_j^{a-1}.$$

Now for any vector $\mathbf{u} \in \mathbb{R}^n$,

$$\mathbf{u}^T \nabla^2 f(\mathbf{x}) \mathbf{u} = \frac{(1 - a)}{w^{2-1/a}} \left[\left(\sum_{i=1}^n \frac{u_i}{x_i^{1-a}} \right)^2 - w \sum_{i=1}^n \frac{u_i^2}{x_i^{2-a}} \right] \quad (8)$$

Now set $g_i = x_i^{a/2}$ and $b_i = u_i x_i^{a/2-1} \forall 1 \leq i \leq n$. Hence the expression in (8) can be written as

$$\mathbf{u}^T \nabla^2 f(\mathbf{x}) \mathbf{u} = \frac{(1 - a)}{w^{2-1/a}} \left[(\mathbf{a}^T \mathbf{b})^2 - \|\mathbf{a}\|_2^2 \|\mathbf{b}\|_2^2 \right] \leq 0.$$

The above result follows from Cauchy-Schwarz inequality.

5. Show that the entropy function $H(\mathbf{x}) = -\sum_{i=1}^n x_i \log(x_i)$ with $\text{dom } H = \{\mathbf{x} \in \mathbb{R}_{++}^n \mid \mathbf{1}^T \mathbf{x} = 1\}$ is concave.

Solution: The Hessian of $H(\mathbf{x})$ is $\nabla^2 H(\mathbf{x}) = -\text{diag}\{1/x_1, \dots, 1/x_n\} \prec 0$ for $x_i > 0$.