

Chapter-3

Monday, May 3, 2021 11:04 AM

3.3.

$$\text{dom } g = (f(a), f(b)).$$

$$\text{and } g(f(x)) = x, \quad a < x < b.$$

Since f is convex and increasing,
 $f' \geq 0$ and $f' > 0$.

The function g can be realized as a composition,

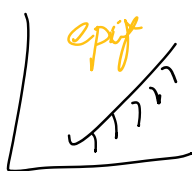
$$g \circ f = g(f(x)), \text{ dom } g = \{f(x) \mid x \in \text{dom } f\}.$$

Since $f(x)$ is convex, g may be convex or concave depending on $\text{dom } g$. Since $f(x)$ is increasing, $g(f(x))$ must be nondecreasing. Thus g is convex.

3.6.

Epigraph is a halfspace when function is affine linear

$$\text{epi } f = \{x \mid a^T x \leq b\}$$



Epigraph is a convex cone when $f(x)$ is convex.

$$\text{epi } f = \{x \mid f(x) \leq t\}. \quad f(x) \text{ is positively homogeneous}$$

Epigraph is a polyhedron when $f(x)$ represents a finite number of inequalities,

$$\text{epi } f = \{x \mid f_1(x) \leq b_1, f_2(x) \leq b_2, \dots\} \quad \text{is piecewise-affine}$$

3.9.

Consider f to be convex and $x, y \in \text{dom } f$. Since $\text{dom } f$ is convex, we conclude that $\forall \theta \in [0, 1], x + \theta(y-x) \in \text{dom } f$,

$$f(x + \theta(y-x)) \leq (1-\theta)f(x) + \theta f(y).$$

Div. by θ ,

$$\begin{aligned} f(x + \theta(y-x)) &\leq \frac{1}{\theta} f(x) - f(x) + f(y) \\ \Rightarrow f(y) &\geq \frac{f(x + \theta(y-x)) - f(x)}{\theta} + f(x). \end{aligned}$$

as $\theta \rightarrow 0$,

$$f(y) \geq f(x) + f'(x)(y-x).$$

Taking deriv. w.r. to x ,

$$\begin{aligned} 0 &\geq f'(x) + f''(x)(y-x) - f'(x), \\ \Rightarrow f''(x) &\leq 0. \end{aligned}$$

Now, assume that $\forall f'(x) \geq 0 \quad \forall x \in \text{dom } f$,

Let $x \neq y$ and $0 \leq \theta \leq 1$, $z = \theta x + (1-\theta)y$,

$$f'(x) \geq 0, \quad f'(y) \geq 0,$$

$$\text{and, } f'(z) = f'(\theta x + (1-\theta)y),$$

$$\geq \theta f'(x) + (1-\theta)f'(y),$$

$$\Rightarrow f'(z) \geq 0.$$

Thus, $f'(x) \geq 0$ implies that f is convex

3.11.

Since $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is convex,

f satisfies the 1st order condition

$$f(y) \geq f(x) + f'(x)(y-x),$$

$$\Rightarrow \frac{f(y) - f(x)}{(y-x)} \geq f'(x),$$

$$\text{and, } f'(y) \geq f'(x),$$

$$\Rightarrow f'(y) - f'(x) \geq 0,$$

$$\Rightarrow \frac{(f'(y) - f'(x))^T (y-x)}{(y-x)} \geq 0$$

$$\Rightarrow (f'(y) - f'(x))^T (y-x) \geq 0.$$

Thus, ∇f is monotone.

Consider the converse by taking $(y(x) - y(x))^T (x-y) \geq 0$

We need to show that $y(x)$ is the gradient of convex function s.t. $x \in \mathbb{R}^n$.

However, this is not true as the 0th condition for $f = (f_1, f_2, \dots, f_n)$ does not hold,

$$\Rightarrow \frac{\partial f_i}{\partial x_j} = \frac{\partial f_j}{\partial x_i}, \quad 1 \leq i \leq j \leq n.$$

3.15.

$$u_2(x) = \frac{x^2-1}{2},$$

$$\lim_{x \rightarrow 0} u_2(x) = \lim_{x \rightarrow 0} \frac{x^2-1}{2} = -\frac{1}{2}$$

(i)

$$\text{Firstly, } u_2(1) = \frac{1^2-1}{2} = 0,$$

$$u_2(x) = \frac{x^2-1}{2} > 0 \quad \forall \quad 0 < x < 1.$$

$$\text{and, } u_2'(x) = (x-1)x^{2-1} < 0 \quad \forall \quad 0 < x < 1.$$

$\Rightarrow u_2'(x)$ is concave increasing and has $u_2(1) = 0$.

3.17.

$$f(x) = \left(\sum_{i=1}^n x_i^p \right)^{1/p} \quad \left| \quad f(x) = \left(\sum_{i=1}^n x_i^q \right)^{1/q} \right.$$

$$f'(x) = \frac{1}{p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} f^{(p-1)} \quad \left| \quad f'(x) = \left(\sum_{i=1}^n \frac{1}{x_i} \right)^{-1} \right.$$

$$f''(x) = (1-p) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} + (p-1) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1}$$

$$\nabla_{x_i}^T f(x) = (1-p) x_i^{p-1} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1}$$

$$\text{Thus, } \nabla^2 f(x) = (1-p) \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} + (p-1) \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} + (1-p) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1}$$

$$(1-p) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} \left[\frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{-1} - \frac{1}{x_i^2} + \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{-1} \right] \quad \alpha^2 \beta^2 \gamma^2$$

We need to show that $\nabla^T \nabla^2 f(x) \nabla \leq 0$,

$$-\nabla^T (p-1) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} \left[\frac{1}{x_i^2} - \frac{1}{x_i^2} + \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{-1} \right]$$

$$= -\nabla^T (p-1) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} \left[\frac{1}{x_i^2} - \frac{1}{x_i^2} + \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{-1} \right]$$

$$= -\nabla^T (p-1) \frac{1}{x_i^p} \left(\sum_{i=1}^n x_i^p \right)^{\frac{1}{p}-1} \left[\frac{1}{x_i^2} - \frac{1}{x_i^2} + \frac{1}{x_i^2} \left(\sum_{i=1}^n x_i^p \right)^{-1} \right] \geq 0$$

Using Cauchy-Schwarz inequality for $\nabla^2 f$.

3.20.

(a)

$$f(\theta x + (1-\theta)y) \leq \theta f(x) + (1-\theta)f(y),$$

$$\leq (\theta f(x) + (1-\theta)f(y)) (\theta g(x) + (1-\theta)g(y)).$$

$$= \theta f(x)g(x) + (1-\theta)f(y)g(y) + \underbrace{\theta(1-\theta)(f(y)-f(x))(g(x)-g(y))}_{\leq 0}$$

$$\Rightarrow f(\theta x + (1-\theta)y)g(\theta x + (1-\theta)y) \leq \theta f(x)g(x) + (1-\theta)f(y)g(y)$$

(b) Same as above but with reversed inequality

(c) Note that f is convex, positive and increasing.
 Thus, $\frac{1}{f}$ is convex as per (a).

x-----x-----x