Problem Set 1

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Warm-Up: Subgradients and More

- 1. Recall that a function $f: \mathbb{R}^n \to \mathbb{R}$ is convex for all $x, y \in \mathbb{R}^n$ and $\lambda \in [0, 1], if$ $\lambda f(x) + (1 \lambda)f(y) \ge f(\lambda x + (1 \lambda)y)$. Using this definition, show that
 - (a) $f(x) = wf_1(x)$ is a convex function for $x \in \mathbb{R}^n$ whenever $f_1 : \mathbb{R}^n \to \mathbb{R}$ is a convex function and $w \ge 0$

Ans:

$$\lambda f(x) + (1 - \lambda)f(y)$$

$$= \lambda w f_1(x) + (1 - \lambda)w f_1(y) \quad [\because f(x) = w f_1(x)]$$

$$= w(\lambda f_1(x) + (1 - \lambda)f_1(y))$$

$$\geq w f_1(\lambda x + (1 - \lambda)y)$$

$$[\because w \geq 0 \quad \text{and}$$

$$\because f_1 \text{ is a convex function}$$

$$\iff \lambda f_1(x) + (1 - \lambda)f_1(y) \geq f_1(\lambda x + (1 - \lambda)y)]$$

$$= f(\lambda x + (1 - \lambda)y)$$

$$\therefore \lambda f(x) + (1 - \lambda)f(y) \geq f(\lambda x + (1 - \lambda)y)$$

$$\therefore f \text{ is a convex function}$$

(b) $f(x) = f_1(x) + f_2(x)$ is a convex function for $x \in \mathbb{R}^n$ whenever $f_1 : \mathbb{R}^n \to \mathbb{R}$ and $f_2 : \mathbb{R}^n \to \mathbb{R}$ are convex functions

Ans:

$$\lambda f(x) + (1 - \lambda)f(y)$$

$$= \lambda (f_1(x) + f_2(x)) + (1 - \lambda)(f_1(y) + f_2(y))$$

$$= \lambda f_1(x) + (1 - \lambda)f_1(y) + \lambda f_2(x) + (1 - \lambda)f_2(y)$$

$$\geq f_1(\lambda x + (1 - \lambda)y) + f_2(\lambda x + (1 - \lambda)y)$$

$$[\because f_1 \text{ and } f_2 \text{ are convex functions }]$$

$$= f(\lambda x + (1 - \lambda)y)$$

$$\therefore \lambda f(x) + (1 - \lambda)f(y) \geq f(\lambda x + (1 - \lambda)y$$

$$\therefore f \text{ is a convex function}$$

(c) $f(x) = max\{f_1(x), f_2(x)\}\$ is a convex function for $x \in \mathbb{R}^n$ whenever $f_1 : \mathbb{R}^n \to \mathbb{R}$ and $f_2 : \mathbb{R}^n \to \mathbb{R}$ are convex functions

Ans:

$$\lambda f(x) + (1 - \lambda)f(y)$$

$$= \lambda \max\{f_1(x), f_2(x)\} + (1 - \lambda) \max\{f_1(y), f_2(y)\}$$

$$\geq \lambda f_1(x) + (1 - \lambda)f_1(y) \quad \text{and} \quad \geq \lambda f_2(x) + (1 - \lambda)f_2(y)$$

$$[\because \max\{f_1(x), f_2(x)\} \geq f_1(x)] \quad [\because \max\{f_1(x), f_2(x)\} \geq f_2(x)]$$

$$\geq f_1(\lambda x + (1 - \lambda)y) \quad \text{and} \quad \geq f_2(\lambda x + (1 - \lambda)y)$$

$$[\because f_1 \text{ is a convex function }] \quad [\because f_2 \text{ is a convex function }]$$

$$\Leftrightarrow \lambda f(x) + (1 - \lambda)f(y) \geq \max\{f_1(\lambda x + (1 - \lambda)y), f_2(\lambda x + (1 - \lambda)y)\}$$

$$\Leftrightarrow \lambda f(x) + (1 - \lambda)f(y) \geq f(\lambda x + (1 - \lambda)y)$$

$$\therefore f \text{ is a convex function}$$