

Problem 1

a. $f(x) = x^2$

$$f'(x) = \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} = 2x \quad f''(x) = 2$$

for $\forall x \in \mathbb{R}$, $f'(x) \geq 0$, $f(x)$ is monotonic increasing function in $x \in \mathbb{R}$.

$\therefore f(x)$ is convex function

b. $f(x) = \ln x$

$$f'(x) = \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} = \frac{1}{x}$$

$$f''(x) = -\frac{1}{x^2}$$

for $\forall x \in \mathbb{R}$, $f''(x) < 0$, $f(x)$ is concave function in $x \in \mathbb{R}$

c. $f(x) = \frac{1}{1+e^{-x}}$

$$f'(x) = \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} = \frac{e^{-x}}{(1+e^{-x})^2}$$

$$f''(x) = \lim_{\Delta x \rightarrow 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \rightarrow 0} \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} = \frac{e^{-x}(e^{-x} - 1)}{(1+e^{-x})^3}$$

\therefore for $x < 0$, $f''(x) > 0$

$x > 0$, $f''(x) < 0$

$\therefore f(x)$ is convex when $x < 0$, $f(x)$ is concave when $x > 0$.

Problem 2

Since X is drawn from uniform distribution between 0 and θ

$$f(x) = \frac{1}{\theta} \quad 0 \leq x \leq \theta$$

a. $E_x[X]$

$$\begin{aligned} E(X) &= \int_{-\infty}^{\infty} x \cdot f(x) dx \\ &= \int_0^{\theta} x \cdot \frac{1}{\theta} dx \\ &= \frac{x^2}{2\theta} \Big|_0^{\theta} \\ &= \frac{\theta}{2} \end{aligned}$$

b. $Var(X)$

$$\begin{aligned} Var(X) &= E(X^2) - [E(X)]^2 \\ &= \int_0^{\theta} x^2 \frac{1}{\theta} dx - \left(\frac{\theta}{2}\right)^2 \\ &= \frac{\theta^2}{12} \end{aligned}$$

c. $H(X)$

$$\begin{aligned} H(X) &= -\int_{-\infty}^{\infty} f(x) \cdot \log f(x) dx \\ &= -\int_0^{\theta} \frac{1}{\theta} \log\left(\frac{1}{\theta}\right) \cdot dx \\ &= -\frac{\log(\theta)}{\theta} x \Big|_0^{\theta} = \frac{\log(\theta)}{\theta} \cdot \theta = \log(\theta) \end{aligned}$$

Problem 3

We know $X \sim U(0, \theta)$, M is drawn from X , $x_1, x_2 \dots x_m$ is from M .

Let $x_{(1)} = \min \{x_1, x_2, x_3 \dots x_m\}$; $x_{(m)} = \max \{x_1, x_2 \dots x_m\}$

$$L(\theta | X) = \prod_{i=1}^m \frac{1}{\theta} = \frac{1}{\theta^m}$$

Find a $\hat{\theta}$ makes $L(\theta | X)$ maximum.

$$\therefore \hat{\theta} = \underset{\theta}{\text{any max}} \frac{1}{\theta^m}$$
$$= \underset{\theta}{\text{any min}} \theta^m$$

$$L(\theta) = \begin{cases} \frac{1}{\theta^m} & , 0 \leq x_1, x_2, \dots x_m \leq \theta \\ 0 & , \text{else} \end{cases}$$

$\therefore 0 \leq x_1, x_2 \dots x_m \leq \theta$ equals to $0 \leq x_{(1)}, x_{(m)} \leq \theta$

$$\therefore L(\theta) = \begin{cases} \frac{1}{\theta^m} & , 0 \leq x_{(1)}, x_{(m)} \leq \theta \\ 0 & , \text{else} \end{cases}$$

$$\text{So } L(\theta) = \frac{1}{\theta^m} \leq \frac{1}{(x_{(m)} - x_{(1)})^m}$$

Only when $\theta = x_{(m)}$, $L(\theta)$ has max value $\frac{1}{x_{(m)}^m}$

$$\therefore \hat{\theta} = \text{Max} \{x_1, x_2, x_3 \dots x_m\}$$

To make sure $0 \leq x_1, x_2 \dots x_m \leq \theta$

$$\therefore \hat{\theta} \neq 0$$

Problem 4

Let R be the event "1M cash behind the door I initially choose", and W be the event "I win the 1M cash by switching doors."

$$P(W|R) = 0$$

$$P(W|R^c) = 1$$

based on Bayes' Rule

$$\begin{aligned} P(W) &= P(W|R) \cdot P(R) + P(W|R^c) \cdot P(R^c) \\ &= 0 \times \frac{1}{3} + 1 \times \frac{2}{3} = \frac{2}{3} \end{aligned}$$

∴ The probability of winning by switching doors is $\frac{2}{3}$ and the probability of winning without switching doors is $\frac{1}{3}$

∴ I should switch doors to maximize the winning probability.

Problem 5

$$\Sigma = E[(X - E[X])(X - E[X])^T]$$

To illustrate Σ is positive semi-definite, we need to prove

$$u \Sigma u^T \geq 0$$

Use definition of Σ

$$\begin{aligned} & u E[(X - E[X])(X - E[X])^T] u^T \\ &= E[(u(X - E[X]))(u(X - E[X]))^T] \\ &= E[u(X - E[X])^2] \end{aligned}$$

An expectation of a positive number is positive.

$\therefore \Sigma$ is positive semidefinite