## Problem Set 3

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## Problem 1.

1

We have:

$$\ell_n = \sum_{i=1}^n (\log \theta + \theta \log \tau - (\theta + 1) \log X_i) \mathbf{1}(X_i \ge \tau)$$

$$\Rightarrow \frac{\partial \ell_n}{\partial \theta} = \sum_{i=1}^n (\frac{1}{\theta} + \log \tau - \log X_i) \mathbf{1}(X_i \ge \tau) = 0$$

$$\Leftrightarrow \theta = \frac{\sum_{i=1}^n (\mathbf{1}(X_i \ge \tau))}{\sum_{i=1}^n (\log X_i - \log \tau) \mathbf{1}(X_i \ge \tau)}$$

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$$\ell_n = \sum_{i=1}^n \left(\frac{1}{2}\log\theta + (\sqrt{\theta} - 1)\log X_i\right) \mathbf{1}(0 \leqslant X_i \leqslant 1)$$

$$\Rightarrow \frac{\partial \ell_n}{\partial \theta} = \left(\frac{1}{2\theta} + \frac{\log X_i}{2\sqrt{\theta}}\right) \mathbf{1}(0 \leqslant X_i \leqslant 1) = 0$$

$$\Leftrightarrow \theta = \left(\frac{\sum_{i=1}^n 1}{\sum_{i=1}^n \log X_i}\right)^2$$

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$$\ell_n = \sum_{i=1}^n (\log X_i - 2\log \theta - \frac{{X_i}^2}{2\theta^2}) \mathbf{1}(X_i \geqslant 0)$$

$$\Leftrightarrow \frac{\partial \ell_n}{\partial \theta} = \sum_{i=1}^n (-\frac{2}{\theta} + \frac{{X_i}^2}{\theta^3}) \mathbf{1}(X_i \geqslant 0) = 0$$

$$\Leftrightarrow \theta = \frac{\sum_{i=1}^n (X_i) \mathbf{1}(X_i \geqslant 0)}{\sum_{i=1}^n \sqrt{2} \mathbf{1}(X_i \geqslant 0)}$$

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$$\ell_{n} = \sum_{i=1}^{n} (\log \theta + \log \tau + (\tau - 1) \log X_{i} - \theta X_{i}^{\tau}) \mathbf{1}(X_{i} \ge 0)$$

$$\Leftrightarrow \frac{\partial \ell_{n}}{\partial \theta} = \sum_{i=1}^{n} (\frac{1}{\theta} - X_{i}^{\tau}) \mathbf{1}(X_{i} \ge 0)$$

$$\Leftrightarrow \theta = \frac{\sum_{i=1}^{n} \mathbf{1} \mathbf{1}(X_{i} \ge 0)}{\sum_{i=1}^{n} (X_{i}^{\tau}) \mathbf{1}(X_{i} \ge 0)}$$

## **Problem 2.** We have :

$$\mathcal{L}(\mu, \sigma) = \prod_{i=1}^{n} \left( \frac{1}{\sigma} \exp(-\frac{1}{2\sigma^{2}} (X_{i} - \mu)^{2}) \right)$$

$$\ell_{n} = \log \mathcal{L}(\mu, \sigma) = \sum_{i=1}^{n} \left( -\log \sigma - \frac{1}{2\sigma^{2}} (X_{i} - \mu)^{2} \right)$$

$$= \sum_{i=1}^{n} \left( -\log \sigma - \frac{1}{2\sigma^{2}} (X_{i} - \bar{X} + \bar{X} - \mu)^{2} \right)$$

$$= \sum_{i=1}^{n} \left( -\log \sigma - \frac{1}{2\sigma^{2}} \left( (X_{i} - \bar{X})^{2} + 2(X_{i} - \bar{X})(\bar{X} - \mu) + (\bar{X} - \mu)^{2} \right) \right)$$

$$-n \log \sigma - \frac{1}{2\sigma^{2}} \left( \sum_{i=1}^{n} (X_{i} - \bar{X})^{2} + 2(\bar{X} - \mu) \sum_{i=1}^{n} (X_{i} - \bar{X}) + n(\bar{X} - \mu)^{2} \right)$$

Because  $\sum_{i=1}^{n} (X_i - \bar{X} = 0)$ , we get :

$$\begin{split} \ell_n &= -n\log\sigma - \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{2\sigma^2} - \frac{n(\bar{X} - \mu)}{2\sigma^2} \\ \Rightarrow & \begin{cases} \frac{\partial \ell_n}{\partial \mu} &= \frac{n(\bar{X} - \mu)}{\sigma^2} = 0 \\ \frac{\partial \ell_n}{\partial \sigma} &= \frac{-n}{\sigma} + \frac{1}{\sigma^3} \left( \sum_{i=n}^n (X_i - \bar{X})^2 + n(\bar{X} - \mu)^2 \right) = 0 \end{cases} \\ \Leftrightarrow & \begin{cases} \mu &= \bar{X} \\ \sigma^2 &= \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n} \end{cases} \end{split}$$

Prove this is consitent:

To do this, we need to prove bias  $\to 0$  and  $\mathbf{V}(\mu, \sigma^2) \to 0$  as  $n \to \infty$ , which means that prove  $\mathbb{E}[\bar{X}] = \mu$  and  $\mathbb{E}\left[\frac{\sum_{i=1}^{n}(X_i - \bar{X})^2}{n}\right] = 0$ 

$$\mathbb{E}[\bar{X}] = \mathbb{E}\left[\frac{\sum_{i=1}^{n} X_{i}}{n}\right] = \frac{1}{n} \sum_{i=1}^{n} \mathbb{E}[X_{i}] = \frac{n\mu}{n} = \mu$$

$$\mathbb{E}\left[\frac{\sum_{i=1}^{n} (X_{i} - \bar{X})^{2}}{n}\right] = \mathbb{E}\left[\frac{\sum_{i=1}^{n} (X_{i} - \mu + \mu - \bar{X})^{2}}{n}\right] = \mathbb{E}\left[\frac{\sum_{i=1}^{n} (X_{i} - \mu)^{2} + 2(\mu - \bar{X})n(\bar{X} - \mu) + n(\mu - \bar{X})^{2}}{n}\right]$$

$$= \mathbb{E}\left[\frac{\sum_{i=1}^{n} (X_{i} - \mu)^{2} - n(\mu - \bar{X})^{2}}{n}\right] = \frac{\sigma^{2} - n\frac{\sigma^{2}}{n}}{n} = 0$$