

1st 56

3.) $X_1, \dots, X_n \dots$ iid uniform $(0, \theta)$ $\theta > 0$

a) MME of θ

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i = \mu(\hat{\theta}) = \int_{\mathbb{R}} x f_{\hat{\theta}}(x) dx = \int_0^{\hat{\theta}} x \frac{1}{\hat{\theta}} dx = \frac{1}{\hat{\theta}} \left(\frac{x^2}{2} \right) \Big|_0^{\hat{\theta}} = \frac{1}{\hat{\theta}} \frac{\hat{\theta}^2}{2} = \frac{\hat{\theta}}{2}$$

$$\Rightarrow \hat{\theta} = 2\bar{X}$$

MLE of θ

$$L(\hat{\theta}) = \begin{cases} 0 & \text{if } \exists i \in \{1, \dots, n\} : X_i > \hat{\theta} \\ \left(\frac{1}{\hat{\theta}}\right)^n & \text{otherwise} \end{cases}$$

$\frac{1}{\hat{\theta}^n}$ gets bigger if $\hat{\theta}$ is smaller \Rightarrow we want the smallest $\hat{\theta}$ with

$$\hat{\theta} \leq X_i \forall i \Rightarrow \hat{\theta} = \max \{X_i : i \in \{1, \dots, n\}\}$$

b) MSE of both estimators

$$MSE = W + b^2$$

$$MME: b(\hat{\theta}) = E(2\bar{X}) - \theta = 2E(\bar{X}) - \theta = \theta - \theta = 0$$

$$V(\hat{\theta}) = V(2\bar{X}) = 4V(\bar{X}) = 4 \frac{\theta^2}{12n} = \frac{\theta^2}{3n}$$

$$\Rightarrow MSE = \frac{\theta^2}{3n} + 0 = \frac{\theta^2}{3n}$$

$$MLE: b(\hat{\theta}) = E(\max_{1 \leq i \leq n} X_i) - \theta = \frac{n}{n+1} \theta - \theta = -\theta \frac{1}{n+1}$$

$$V(\hat{\theta}) = V(\max_{1 \leq i \leq n} X_i) = \frac{n \theta^2}{(n+1)^2 (n+2)}$$

$$\Rightarrow MSE = \frac{n \theta^2}{(n+1)^2 (n+2)} + \frac{\theta^2}{(n+1)^2} = \frac{n \theta^2 + \theta^2 (n+2)}{(n+1)^2} = \frac{2n \theta^2 + 2 \theta^2}{(n+1)^2} = 2 \theta^2 \frac{n+1}{(n+1)^2}$$

$$= 2 \theta^2 \frac{1}{n+1}$$

$$\theta^2 \frac{1}{3n} = \theta^2 \frac{2}{n+1} \Leftrightarrow n+1 = 6n \Leftrightarrow n = \frac{1}{5} \Rightarrow \forall n > 1: \frac{2 \theta^2}{n+1} > \frac{\theta^2}{3n}$$

therefore MME has a better MSE than MLE.