

# ACM 157 Set 4

2.

$$X_1, \dots, X_n \sim U(\alpha, \beta) \quad \alpha < \beta$$

a)

$$\eta_0(\alpha, \beta) = \hat{m}_0 \quad \hat{m}_1 = \frac{1}{n} \sum_{i=1}^n X_i \quad \hat{m}_2 = \frac{1}{n} \sum_{i=1}^n X_i^2$$

$$\eta_1(\alpha, \beta) = E[X] = \int_{\alpha}^{\beta} x \frac{1}{\beta - \alpha} dx = \left( \frac{x^2}{2} \frac{1}{\beta - \alpha} \right) \Big|_{\alpha}^{\beta} = \frac{\beta^2 - \alpha^2}{2(\beta - \alpha)} = \frac{\beta + \alpha}{2}$$

$$\eta_2(\alpha, \beta) = E[X^2] = \int_{\alpha}^{\beta} x^2 \frac{1}{\beta - \alpha} dx = \left( \frac{x^3}{3} \frac{1}{\beta - \alpha} \right) \Big|_{\alpha}^{\beta} = \frac{\beta^3 - \alpha^3}{3(\beta - \alpha)} = \frac{\alpha^2 + \alpha\beta + \beta^2}{3}$$

$$\hat{m}_1 = \frac{\alpha + \beta}{2} \quad \hat{m}_2 = \frac{\alpha^2 + \alpha\beta + \beta^2}{3}$$

$$\alpha = \hat{m}_1 - \sqrt{3(\hat{m}_2 - \hat{m}_1^2)} \quad \beta = \hat{m}_1 + \sqrt{3(\hat{m}_2 - \hat{m}_1^2)}$$

b)

$$L(\alpha, \beta | X_1, \dots, X_n) = \prod_{i=1}^n \left( \frac{1}{\beta - \alpha} \right) = \left( \frac{1}{\beta - \alpha} \right)^n$$

$$\hat{\alpha}_{MLE} = \arg \max_{\alpha \in [X_{(1)}, \dots, X_n]} \left( \frac{1}{\beta - \alpha} \right)^n = \min(X_1, \dots, X_n) = X_{(1)}$$

$$\hat{\beta}_{MLE} = \arg \max_{\beta \in [X_1, \dots, X_n]} \left( \frac{1}{\beta - \alpha} \right)^n = \max(X_1, \dots, X_n) = X_{(n)}$$

3.

$$X_1, \dots, X_n \sim U(\alpha, \beta) \quad \mu = \int x dF(x)$$

a)

$$\mu = \frac{\alpha + \beta}{2}$$

$$\hat{\mu}_{MLE} = \mu(\hat{\alpha}_{MLE}, \hat{\beta}_{MLE}) = \frac{\hat{\alpha}_{MLE} + \hat{\beta}_{MLE}}{2} = \frac{X_{(1)} + X_{(n)}}{2}$$

b)

$$\hat{\mu}_n = \bar{X}_n \quad n \geq 10 \quad \alpha = 1 \quad \beta = 3$$

$$MSE[\hat{\mu}_n] = \text{bias}[\hat{\mu}_n]^2 + \text{se}[\hat{\mu}_n]^2$$

$$= (E[\hat{\mu}_n] - \mu)^2 + V(\hat{\mu}_n)/n$$

$$= (E[\bar{X}_n] - \mu)^2 + \sigma^2/10$$

$$= (\mu - \mu)^2 + \frac{1}{10} (E[X^2] - E[X]^2)$$

$$= 0 + \frac{1}{10} \left( \int_{\alpha}^{\beta} x^2 \frac{1}{\beta - \alpha} dx - \left( \int_{\alpha}^{\beta} x \frac{1}{\beta - \alpha} dx \right)^2 \right)$$

$$= \frac{1}{10} \left( \frac{x^3}{3} \frac{1}{3-1} \Big|_1^3 - \left( \frac{x^2}{2} \frac{1}{3-1} \Big|_1^3 \right)^2 \right)$$

$$= \frac{1}{10} \left( \frac{27}{3} \frac{1}{2} - \frac{1}{3} \frac{1}{2} - \left( \frac{9}{2} \frac{1}{2} - \frac{1}{2} \frac{1}{2} \right)^2 \right)$$

$$= \frac{1}{10} \left( \frac{13}{3} - (2)^2 \right)$$

$$= \frac{1}{10} \cdot \frac{1}{3} = \frac{1}{30}$$

4.  $X_1, \dots, X_n \sim N(\theta, 1)$   $Y_i = \begin{cases} 1, & \text{if } X_i > 0 \\ 0, & \text{if } X_i \leq 0 \end{cases}$   $\Psi = E[Y_i]$

a)

$$\Psi = E[Y_i] = \int_{-\infty}^{\infty} y N(\theta, 1) dy = \int_{-\infty}^0 0 \cdot N(\theta, 1) dy + \int_0^{\infty} 1 \cdot N(\theta, 1) dy$$

$$= \int_0^{\infty} N(\theta, 1) dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\theta)^2}{2}} dx$$

$\hat{\theta}_{MLE} = \bar{X}_n$  from Maximum Likelihood notes eq. 12 with  $\mu = \theta$

$$\hat{\Psi}_{MLE} = \Psi(\hat{\theta}_{MLE}) = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\hat{\theta}_{MLE})^2}{2}} dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\bar{X}_n)^2}{2}} dx$$

b)

$$I_{\theta} = \hat{\theta}_n \pm z_{1-\alpha/2} \hat{SE} = \bar{X}_n \pm z_{0.975} \frac{\sigma}{\sqrt{n}} = \bar{X}_n \pm \frac{1.96}{\sqrt{n}}$$

$$I_{\Psi} = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-I_{\theta})^2}{2}} dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-(\bar{X}_n \pm \frac{1.96}{\sqrt{n}}))^2}{2}} dx$$

5.

$X_1, \dots, X_n \sim U[0, \theta]$   $\hat{\theta}_{MLE} = X_{(n)}$

a)

$$\text{CDF} = P(X_{(n)} \leq x) = P(X_1 \leq x) \cdot \dots \cdot P(X_n \leq x)$$

$$= \frac{x}{\theta} \cdot \dots \cdot \frac{x}{\theta} = \left(\frac{x}{\theta}\right)^n$$

$$\text{PDF} = \frac{nx^{n-1}}{\theta^n}$$