

$$1. \quad P(X_1 = \lambda_1, X_2 = \lambda_2 \mid X_1 + X_2 = y)$$

→ does not depend on y if $y = 0, 1, 3, \dots$

$$\text{But, when } y = 2, \quad P(X_1 + X_2 = 2) = P(X_1 = 1, X_2 = 1) + 2P(X_1 = 0, X_2 = 2) \\ = 2(1 - e^{-\theta} - \theta e^{-\theta})e^{-\theta} + \theta^2 e^{-2\theta}$$

$$\frac{P(X_1 = \lambda_1, X_2 = \lambda_2, X_1 + X_2 = 2)}{P(X_1 + X_2 = 2)}$$

$$= \frac{P(X_1 = \lambda_1, X_2 = 2 - \lambda_1)}{2(1 - e^{-\theta} - \theta e^{-\theta})e^{-\theta} + \theta^2 e^{-2\theta}}$$

$$= \frac{P(X_1 = \lambda_1) P(X_2 = 2 - \lambda_1)}{2(1 - e^{-\theta} - \theta e^{-\theta})e^{-\theta} + \theta^2 e^{-2\theta}}$$

$$(i) \quad \lambda_1 = 0, \lambda_2 = 2$$

$$P(X_1 = \lambda_1, X_2 = \lambda_2 \mid X_1 + X_2 = y) = \frac{e^{-\theta}(1 - e^{-\theta} - \theta e^{-\theta})}{2(1 - e^{-\theta} - \theta e^{-\theta})e^{-\theta} + \theta^2 e^{-2\theta}} \Rightarrow \text{depends on } \theta$$

$$(ii) \quad \lambda_1 = 2, \lambda_2 = 2 \Rightarrow \text{Similar to (i), } P(X_1 = \lambda_1, X_2 = \lambda_2 \mid X_1 + X_2 = y) \text{ depends on } \theta$$

$$(iii) \quad \lambda_1 = 1, \lambda_2 = 1$$

$$P(X_1 = \lambda_1, X_2 = \lambda_2 \mid X_1 + X_2 = y) = \frac{\theta^2 e^{-2\theta}}{2(1 - e^{-\theta} - \theta e^{-\theta})e^{-\theta} + \theta^2 e^{-2\theta}} \Rightarrow \text{depends on } \theta$$

$\therefore X_1 + X_2$ is NOT SS for θ .

2.

$$(a) L(\lambda) = 9\lambda^6 x_1^2 x_2^2 e^{-\lambda^3 x_1^3 - \lambda^3 x_2^3}$$

$\Rightarrow x_1^3 + x_2^3$: s.s. by factorization Theorem

$$(b) \ell(\lambda) \propto 6 \log \lambda - \lambda^3 (x_1^3 + x_2^3)$$

$$\ell'(\lambda) = \frac{6}{\lambda} - 3\lambda^2 (x_1^3 + x_2^3) = 0$$

$$\Rightarrow \frac{2}{\lambda^3} = x_1^3 + x_2^3 \Rightarrow \lambda^3 = \frac{2}{x_1^3 + x_2^3}$$

$$\hat{\lambda} = \left(\frac{2}{x_1^3 + x_2^3} \right)^{\frac{1}{3}} = \left(\frac{1}{14} \right)^{\frac{1}{3}}$$

3.

$$(a) L(\theta) = \theta^n 2^{n\theta} \left(\prod_{i=1}^n X_i \right)^{-(\theta n)}, \quad \ell(\theta) = n \log \theta + n\theta \log 2 - (\theta n) \sum_{i=1}^n \log X_i$$

$$\ell'(\theta) = \frac{n}{\theta} + n \log 2 - \sum \log X_i = 0 \Rightarrow \frac{n}{\theta} = \sum \log \left(\frac{X_i}{2} \right)$$

$$\therefore \hat{\theta}_{MLE} = \frac{n}{\sum \left(\log \frac{X_i}{2} \right)}$$

$$(b) \log f(x|\theta) = \log \theta + \theta \log 2 - (\theta n) \log x$$

$$\frac{d}{d\theta} \log f(x|\theta) = \frac{1}{\theta} - \log x$$

$$\frac{d^2}{d\theta^2} \log f(x|\theta) = -\frac{1}{\theta^2} \Rightarrow I(\theta) = \frac{1}{\theta^2}$$

$$(c) \sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} N(0, \theta^2)$$

$$4. f_X(x) = \theta(1-x)^{\theta-1} \Rightarrow E(X) = \frac{1}{1+\theta}, V(X) = \frac{\theta}{(1+\theta)^2(2+\theta)}$$

$$\text{Let } Y = -\log(1-X) \Rightarrow f_Y(y) = \theta e^{-\theta y} : Y \sim \text{Gamma}(1, \frac{1}{\theta})$$

$$(a) \hat{\theta}_1 = \frac{1}{\frac{1}{n} \sum -\log(1-X_i)} = \frac{1}{\bar{Y}}$$

$$\text{From CLT, } \sqrt{n}(\bar{Y} - \frac{1}{\theta}) \xrightarrow{d} N(0, \frac{1}{\theta^2})$$

$$\text{Let } s(y) = \frac{1}{y} \Rightarrow s'(y) = -\frac{1}{y^2}$$

$$\text{From Delta method, } \sqrt{n}(\frac{1}{\bar{Y}} - \theta) \xrightarrow{d} N(0, \theta^2)$$

$$\hat{\theta}_1 = \frac{1}{\bar{Y}}$$

$$\text{From CLT, } \sqrt{n}(\bar{X} - \frac{1}{1+\theta}) \xrightarrow{d} N(0, \frac{\theta}{(1+\theta)^2(2+\theta)})$$

$$\text{Let } s(x) = \frac{1}{1-x} \Rightarrow s'(x) = \frac{1}{(1-x)^2}$$

$$\text{From Delta method, } \sqrt{n}(\frac{1}{1-\bar{X}} - \frac{\theta+1}{\theta}) \xrightarrow{d} N(0, \frac{(1+\theta)^4}{\theta^4} \frac{\theta}{(1+\theta)^2(2+\theta)})$$

$$\sqrt{n}(\hat{\theta}_2 - \frac{\theta+1}{\theta}) \xrightarrow{d} N(0, \frac{(1+\theta)^2}{\theta^3(2+\theta)})$$

(b) There is a mistake in my problem (a).

$\hat{\theta}_2$ should be $\frac{1-\bar{X}}{\bar{X}}$! So computing ARE is not relevant..

I gave you full points for (b) regardless of your answer.

Sorry about that.

5. (a)

$$L(\theta) = \frac{2^n}{\theta^{2n}} \prod_{i=1}^n I(0 < X_i \leq \theta) = \frac{2^n}{\theta^{2n}} I(0 < \max_{1 \leq i \leq n} X_i \leq \theta)$$

$$\Rightarrow \hat{\theta} = Y = \max_{1 \leq i \leq n} X_i$$

$$\therefore F(x; \theta) = \frac{x^2}{\theta^2} = \frac{1}{2} \Rightarrow \text{median} = \frac{\theta}{\sqrt{2}}$$

$$\text{MLE for median} = \frac{Y}{\sqrt{2}}$$

$$(b) f_Y(y) = \frac{2n y^{2n-1}}{\theta^{2n}}, \quad 0 < y \leq \theta, \quad F_Y(y) = \left(\frac{y}{\theta}\right)^{2n}$$

$$E(Y) = \int_0^{\theta} \frac{2n y^{2n}}{\theta^{2n}} dy = \frac{2n}{2n+1} \theta$$

$$\Rightarrow E\left(\frac{Y}{\sqrt{2}}\right) = \frac{2n \theta}{(2n+1)\sqrt{2}} \Rightarrow \frac{(2n+1)Y}{2n\sqrt{2}} \text{ is unbiased estimator for the median}$$

(c)

$$\lim_{n \rightarrow \infty} P\left(\left|\frac{Y}{\sqrt{2}} - \frac{\theta}{\sqrt{2}}\right| > \varepsilon\right)$$

$$= \lim_{n \rightarrow \infty} P(|Y - \theta| > \sqrt{2} \varepsilon)$$

$$= \lim_{n \rightarrow \infty} [P(Y > \theta + \sqrt{2} \varepsilon) + P(Y < \theta - \sqrt{2} \varepsilon)]$$

$$= \lim_{n \rightarrow \infty} P(Y < \theta - \sqrt{2} \varepsilon) = \lim_{n \rightarrow \infty} \left(\frac{\theta - \sqrt{2} \varepsilon}{\theta}\right)^{2n} = \lim_{n \rightarrow \infty} \left(1 - \frac{\sqrt{2} \varepsilon}{\theta}\right)^{2n} = 0$$