

(a)

$$L(\theta) = \prod_{i=1}^n \left(\frac{2x_i}{\theta^2} \right) \quad (0 \leq x \leq \theta)$$
$$= \left(\frac{2}{\theta^2} \right)^n \prod_{i=1}^n x_i \quad (0 \leq x \leq \theta)$$

we can see when θ increasing, $L(\theta)$ would become smaller
so θ should be equal to its minimum value to make sure $L(\theta)$ get its maximum value. then $\hat{\theta}_{ml} = \max(x_1, \dots, x_n) = x_{(n)}$ (nth order statistics.)

(b)

$$E(x) = \int_0^{\theta} \frac{2x}{\theta^2} \cdot x dx = \int_0^{\theta} \frac{2x^2}{\theta^2} dx = \left. \frac{2}{3} \frac{x^3}{\theta^2} \right|_0^{\theta} = \frac{2}{3} \theta$$

$$\frac{2}{3} \theta = \frac{1}{n} \sum_{i=1}^n x_i = \bar{x} \quad \hat{\theta}_{mm} = \frac{3}{2} \bar{x}$$

$$E(x^2) = \int_0^{\theta} \frac{2x}{\theta^2} \cdot x^2 dx = \int_0^{\theta} \frac{2x^3}{\theta^2} dx = \left. \frac{2}{4} \frac{x^4}{\theta^2} \right|_0^{\theta} = \frac{1}{2} \theta^2$$

$$\text{Var}(x) = E(x^2) - E^2(x) = \frac{1}{2} \theta^2 - \frac{4}{9} \theta^2 = \frac{1}{18} \theta^2$$

\therefore based on CLT: $\bar{x} \sim N\left(\frac{2}{3}\theta, \frac{\theta^2}{18n}\right)$

$$E(\hat{\theta}_{mm}) = \frac{3}{2} E(\bar{x}) = \theta \quad \text{Var}(\hat{\theta}_{mm}) = \text{Var}\left(\frac{3}{2} \bar{x}\right) = \frac{9}{4} \text{Var}(\bar{x}) = \frac{9}{4} \times \frac{\theta^2}{18n} = \frac{\theta^2}{8n}$$

$$\therefore \hat{\theta}_{mm} \sim N\left(\theta, \frac{\theta^2}{8n}\right)$$

$$\frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0,1) \quad \text{when } n \text{ is large enough}$$

$$\therefore \frac{\hat{\theta}_{mm} - \theta}{\frac{\theta}{\sqrt{8n}}} \sim N(0,1) \Rightarrow P\left(Z_{0.025} < \frac{\hat{\theta}_{mm} - \theta}{\frac{\theta}{\sqrt{8n}}} < Z_{0.975}\right) = 0.95$$

$$\Rightarrow P\left(Z_{0.025} < \frac{\hat{\theta}_{mm} - \theta}{\frac{\hat{\theta}_{mm}}{\sqrt{8n}}} < Z_{0.975}\right) = 0.95$$

$$\Rightarrow P\left(\hat{\theta}_{mm} Z_{0.975} \frac{\hat{\theta}_{mm}}{\sqrt{8n}} < \theta < \hat{\theta}_{mm} - Z_{0.025} \frac{\hat{\theta}_{mm}}{\sqrt{8n}}\right)$$

$$\therefore 95\% \text{ CI for } \theta \text{ is } \left(\hat{\theta}_{mm} \left(1 - \frac{1.96}{\sqrt{8n}}\right), \hat{\theta}_{mm} \left(1 + \frac{1.96}{\sqrt{8n}}\right) \right)$$

$$\left(\frac{3}{2} \bar{x} \left(1 - \frac{1.96}{\sqrt{8n}}\right), \frac{3}{2} \bar{x} \left(1 + \frac{1.96}{\sqrt{8n}}\right) \right)$$

$$c) T = \hat{\theta}_{ML} = \max(X_1, X_2, \dots, X_n) = X_{(n)}$$

$$P(T \leq t) = P(\max(X_1, X_2, \dots, X_n) \leq t) = \prod_{i=1}^n P(X_i \leq t)$$

$$P(X_i \leq t) = \int_0^t \frac{2x}{\theta^2} dx \quad (0 \leq x \leq \theta) = \frac{x^2}{\theta^2} \Big|_0^t = \frac{t^2}{\theta^2}$$

CDF of $\hat{\theta}_{ML}$:

$$P(T \leq t) = \begin{cases} 0 & t \leq 0 \\ \left(\frac{t^2}{\theta^2}\right)^n & 0 < t < \theta \\ 1 & t \geq \theta \end{cases}$$

$$F_n(y) = P(W_n \leq y) = P(n(\theta - \hat{\theta}_{ML}) \leq y) = P(\theta - \hat{\theta}_{ML} \leq \frac{y}{n}) = P(\hat{\theta}_{ML} \geq \theta - \frac{y}{n})$$

$$F_n(y) = 1 - P(\hat{\theta}_{ML} \leq \theta - \frac{y}{n}) = 1 - \left[\frac{(\theta - \frac{y}{n})^2}{\theta^2}\right]^n = 1 - \left(1 - \frac{y}{\theta}\right)^{2n} = 1 - \left(1 - \frac{2y}{\theta}\right)^n$$

$$\lim_{n \rightarrow \infty} F_n(y) = 1 - e^{-2y/\theta}$$

The assumption of asymptotic normality is violated here.

① the range of RV is depend on parameter θ

② $\hat{\theta}_{ML}$ is not differentiable

cd)

$$P\left(\frac{W_n}{\theta} \leq y\right) = P(W_n \leq y\theta) = 1 - e^{-2y}$$

so $\frac{W_n}{\theta} \sim \text{Exp}(\lambda=2)$ a: 2.5% quantile of $\text{Exp}(2)$. b: 97.5% quantile of $\text{Exp}(2)$

$$P\left\{a \leq \frac{W_n}{\theta} \leq b\right\} = 0.95$$

$$P\left\{a \leq \frac{n(\theta - \hat{\theta}_{ML})}{\theta} \leq b\right\} = 0.95 \Rightarrow P\left\{a \leq \frac{n(\theta - \hat{\theta}_{ML})}{\hat{\theta}_{ML}} \leq b\right\} = 0.95$$

$$\Rightarrow P\left\{\frac{a\hat{\theta}_{ML}}{n} \leq \theta - \hat{\theta}_{ML} \leq \frac{b\hat{\theta}_{ML}}{n}\right\} = 0.95 \Rightarrow P\left\{\hat{\theta}_{ML}\left(1 + \frac{a}{n}\right) \leq \theta \leq \hat{\theta}_{ML}\left(1 + \frac{b}{n}\right)\right\} = 0.95$$

in R command: a = qexp(0.025, rate=2) \approx 0.0127 . b = qexp(0.975, rate=2) \approx 1.844

so 95% CI of θ is $(X_{(n)}(1 + \frac{0.0127}{n}), X_{(n)}(1 + \frac{1.844}{n}))$

(e)

$$n=2 \quad \text{CDF of } \hat{\theta}_{ML}: P(T \leq t) = \frac{t^3}{\theta^3} \quad (0 < t < \theta)$$

$$\alpha = P(\hat{\theta}_{ML} > c | \theta=1) = 1 - P(\hat{\theta}_{ML} \leq c | \theta=1) = 1 - c^3$$

$$\beta = P(\hat{\theta}_{ML} \leq c | \theta=2) = \frac{c^3}{2^3}$$

$$d^2 + \beta^2 = (1 - c^3)^2 + \left(\frac{c^3}{8}\right)^2 \quad \text{let } c^3 = u$$

$$d^2 + \beta^2 = (1 - u)^2 + \frac{u^2}{64} = 1 - 2u + u^2 + \frac{u^2}{64}$$

$$\text{when } u = -\frac{-2}{2(1 + \frac{1}{64})}, d^2 + \beta^2 \text{ gets its minimum value}$$

$$c^3 = \frac{16^2}{16^2 + 1} \quad c \approx 0.999$$

assuming H_0 is true, and observed value $\hat{\theta}_{ML}$ is 0.99

$$\therefore \text{P-value: } 1 - P\left(\frac{0.99^3}{\theta^3} \leq t | H_0\right) = 1 - 0.99^3 \approx 0.039 < 5\%$$

So we have strong confidence to reject H_0 .