

Rubric Q2 Mid-sem

Q2 Y_1, Y_2, \dots, Y_n are iid samples from the pdf $f_0(y_i)$

$$f_0(y_i) = \frac{1}{2\beta^3} y_i^2 e^{-y_i/\beta} \quad y_i > 0, \beta > 0$$

a) MoM estimator for β

$$\left. \begin{aligned} m_1 &= \frac{1}{n} \sum Y_i \\ E[X] &= \int_0^\infty y f(y) \cdot dy \\ \Rightarrow \int y f(y) \cdot dy &= \int_0^\infty y \cdot \frac{1}{2\beta^3} y^2 e^{-y/\beta} \cdot dy \\ &\Rightarrow \frac{1}{2\beta^3} \int_0^\infty y^3 \cdot e^{-y/\beta} \cdot dy \end{aligned} \right\} +0.5$$

applying integration by parts $\int u \cdot v dy = u \int v dy - \int u' v dy$

$$\Rightarrow \frac{1}{2\beta^3} \left[y^3 \int e^{-y/\beta} \cdot dy - \int (3y^2 \int e^{-y/\beta} \cdot dy) \cdot dy \right]$$

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} - \int -3y^2 \beta e^{-y/\beta} \cdot dy \right]$$

again applying by parts

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} + 3\beta \int y^2 e^{-y/\beta} \cdot dy \right]$$

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} + 3\beta \left[y^2 \int e^{-y/\beta} \cdot dy - \int 2y \int e^{-y/\beta} \cdot dy \right] \right]$$

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} + 3\beta \left[-\beta y^2 e^{-y/\beta} + 2\beta \int y e^{-y/\beta} \cdot dy \right] \right]$$

again applying by parts,

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} - 3\beta^2 y^2 e^{-y/\beta} + 6\beta^2 \left[y \int e^{-y/\beta} dy - \int \int e^{-y/\beta} dy dy \right] \right]$$

$$\Rightarrow \frac{1}{2\beta^3} \left[-\beta y^3 e^{-y/\beta} - 3\beta^2 y^2 e^{-y/\beta} + 6\beta^3 y e^{-y/\beta} - \cancel{6\beta^4 e^{-y/\beta}} - 6\beta^4 e^{-y/\beta} \right] \Big|_0^{\infty}$$

$$\Rightarrow -\frac{1}{2\beta^2} \left[y^3 e^{-y/\beta} + 3\beta y^2 e^{-y/\beta} + 6\beta^2 y e^{-y/\beta} + 6\beta^3 e^{-y/\beta} \right] \Big|_0^{\infty}$$

$$\Rightarrow 0 - \left(-\frac{1}{2\beta^2} \right) (6\beta^3) + 1 \text{ depending on correctness of integration steps.}$$

$$\Rightarrow \left. \begin{aligned} &3\beta \\ \therefore, E[y] &= 3\beta \\ m_1 = E[y] &= 3\beta \\ \frac{\sum y_i}{n} &= 3\beta \end{aligned} \right\} + 0.5$$

$$\Rightarrow \boxed{\hat{\beta} = \frac{\sum y_i}{3n}} \rightarrow \text{mom estimate}$$

b) MLE estimator for β

$$L(\theta) = \prod_{i=1}^n f_{\theta}(y_i)$$

$$= \prod_{i=1}^n \frac{1}{2\beta^3} y_i^2 e^{-y_i/\beta} \quad y_i > 0, \beta > 0$$

$$= \left(\frac{1}{2\beta^3}\right)^n \left(\prod_{i=1}^n y_i\right)^2 e^{-\sum y_i/\beta}$$

$$\log L(\theta) = \log \left[\left(\frac{1}{2\beta^3}\right)^n \left(\prod_{i=1}^n y_i\right)^2 e^{-\sum y_i/\beta} \right]$$

$$= -n \log(2\beta^3) + 2 \log \left(\prod_{i=1}^n y_i \right) - \sum y_i/\beta$$

$$\frac{d \log L(\theta)}{d\theta} = \frac{-n}{2\beta^3} \times 6\beta^2 + 0 + \frac{\sum y_i}{\beta^2}$$

$$\frac{d \log L(\theta)}{d\theta} = 0$$

$$\frac{6\beta^2 n}{2\beta^3} = \frac{\sum y_i}{\beta^2}$$

$$\Rightarrow \hat{\beta} = \frac{\sum y_i}{3n}$$

check $\frac{d^2L}{d\theta^2}$ at $\hat{\beta} = \frac{\sum y_i}{3n}$

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$$\frac{d^2L}{d\theta^2} = \frac{3n}{\beta^2} + 0 - \frac{2\sum y_i}{\beta^3}$$

$$\left. \frac{d^2L}{d\theta^2} \right|_{\hat{\beta} = \frac{\sum y_i}{3n}}$$

$$\Rightarrow \frac{27n^3}{(\sum y_i)^2} - \frac{2\sum y_i}{(\sum y_i)^3} \times 27n^3$$

$$\Rightarrow \frac{-27n^3}{(\sum y_i)^2}$$

we can see that $(\sum y_i)^2 > 0$ as $y_i > 0$

$$\therefore, \left. \frac{d^2L}{d\theta^2} \right|_{\hat{\beta}} < 0$$

since 2nd order derivative is $-ve$, $\hat{\beta} = \frac{\sum y_i}{3n}$

is the maxima.

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c) $Y \sim f_Y(y) = \frac{1}{2\beta^2} y^2 e^{-y/\beta} \quad y > 0, \beta > 0$

let $Z = cy + d$

$\Rightarrow y = \frac{Z-d}{c} \Rightarrow \frac{dy}{dz} = \frac{1}{c}$

By transformation theorem distribution/pdf of Z is given by

$f(z) = \left| \frac{dy}{dz} \right| \cdot f(y)$

$f(z) = \frac{1}{c} \cdot \frac{1}{2\beta^2} y^2 e^{-y/\beta}$

$= \frac{1}{c} \cdot \frac{1}{2\beta^2} \left(\frac{z-d}{c} \right)^2 e^{-\frac{(z-d)}{\beta c}}$

$= \frac{1}{c} \cdot \frac{1}{2\beta^2} \left(\frac{z-d}{c} \right)^2 e^{-\frac{1}{\beta} \left(\frac{z-d}{c} \right)}$

$= \frac{1}{c} \cdot f_Y \left(\frac{z-d}{c} \right)$

Thus, Y belongs to the loc-scale family