

# IE241 final Practice solution (even number)

2. Let  $V = Y_1 Y_2$ ,  $U = -\ln^{(V)}$

$$\begin{aligned}
 F_V(v) &= P(Y_1 Y_2 \leq v) = \int_0^1 P(Y_1 \leq \frac{v}{y_2} \mid Y_2 = y_2) \cdot f_{Y_2}(y_2) dy_2 \\
 &= \int_0^v \underbrace{f_{Y_2}(y_2)}_{=1} dy_2 + \int_v^1 \frac{v}{y_2} \cdot \underbrace{f_{Y_2}(y_2)}_{=1} dy_2 \\
 &= \int_0^v 1 \cdot dy_2 + \int_v^1 \frac{v}{y_2} dy_2 = v - v \ln^{(v)}
 \end{aligned}$$

$P(Y_1 \leq \frac{v}{y_2} \mid Y_2 = y_2) = \begin{cases} \frac{v}{y_2} & (y_2 \geq v) \\ 1 & (y_2 < v) \end{cases}$

$$f_V(v) = \frac{d}{dv} F_V(v) = -\ln^{(v)}$$

$$\therefore f_U(u) = f_V(e^{-u}) \cdot \left| \frac{d}{du} e^{-u} \right| = u \cdot e^{-u} \quad (u > 0)$$

4. (a)  $\frac{(n-1)S_x^2}{\sigma^2} \sim \chi^2(n-1)$ ,  $\frac{(m-1)S_r^2}{\sigma^2} \sim \chi^2(m-1)$

and they are independent.  $\therefore W \sim \chi^2_{n+m-2}$

(b)  $\bar{X} - \bar{Y} \sim N(\mu_x - \mu_y, \sigma^2(\frac{1}{n} + \frac{1}{m}))$

$$\therefore \frac{\bar{X} - \bar{Y} - (\mu_x - \mu_y)}{\sigma \sqrt{\frac{1}{n} + \frac{1}{m}}} \sim N(0, 1)$$

(c) Remember  $T = \frac{Z}{\sqrt{V/r}}$  ( $Z \sim N(0, 1)$ ,  $V \sim \chi^2(r)$ ,  $Z$  &  $V$  are independent)

$$\bar{X} \perp S_x^2, \bar{Y} \perp S_r^2, \bar{X} - \bar{Y} \perp S_x^2, \bar{X} - \bar{Y} \perp S_r^2$$

$$\Rightarrow T = \frac{Z}{\sqrt{W/(n+m-2)}} \sim t_{n+m-2}$$

6. 
$$F_Y(y) = \begin{cases} 0 & (y \leq 0) \\ \frac{1}{2}y^2 & (0 < y \leq 1) \\ y - \frac{1}{2} & (1 < y \leq 1.5) \\ 1 & (y > 1.5) \end{cases}, \quad F_U(u) = P(U \leq u) = P(Y \leq \frac{u+4}{10})$$

$$F_U(u) = \begin{cases} 0 & (u \leq -4) \\ \frac{1}{2} \left( \frac{u+4}{10} \right)^2 & (-4 < u \leq 6) \\ \frac{u}{10} - \frac{1}{10} & (6 \leq u \leq 11) \\ 1 & (u > 11) \end{cases} \Rightarrow f(u) = \begin{cases} \frac{u+4}{100} & (-4 < u \leq 6) \\ \frac{1}{10} & (6 < u \leq 11) \\ 0 & \text{elsewhere} \end{cases}$$

8. (a) 
$$F_{Y_{(n)}}(y) = P(Y_{(n)} \leq y) = 1 - P(Y_{(n)} > y) = 1 - [P(Y_i > y)]^n = 1 - [1 - F(y)]^n$$

$$\therefore f_{(n)}(y) = n [1 - F(y)]^{n-1} \cdot f(y)$$

$$F(y) = 1 - e^{-\frac{(y-2)}{4}}, \quad f_{(n)}(y) = n \cdot e^{-\frac{(y-2)(n-1)}{4}} \cdot \frac{1}{4} e^{-\frac{(y-2)}{4}}$$

$$\therefore f_{(n)}(y) = \frac{n}{4} \cdot e^{-\frac{n}{4} \cdot (y-2)} \quad (y \geq 2)$$

(b) 
$$\int_2^{\infty} \frac{ny}{4} \cdot e^{-\frac{n}{4}(y-2)} dy = \frac{1}{n} + 2$$

10. 
$$W = \frac{mv^2}{2}, \quad v = \sqrt{\frac{2W}{m}}, \quad \left| \frac{dv}{dW} \right| = \frac{1}{\sqrt{2mW}}$$

$$f_W(w) = f(v) \cdot \left| \frac{dv}{dW} \right| = a(2w/m) \cdot e^{-w/kT} \cdot \frac{1}{\sqrt{2mW}} \quad (w \geq 0)$$

12.  $Y_1, Y_2, \dots \sim U(-\frac{1}{2}, \frac{1}{2})$ , let  $C = \sum Y_i \sim N(0, \frac{n}{12})$  <sup>48</sup>  $= N(0, 4)$

$$P(|C| > 4) = 2 \cdot P(C > 4) = 2 \cdot P(Z > 2) \leq 0.05$$

$$14. (a) f(y_1, \dots, y_n | \theta) = \frac{1}{\beta^{2n}} \cdot \prod_{i=1}^n y_i \cdot e^{-y_i/\beta} = \frac{1}{\beta^{2n}} \cdot e^{-\frac{1}{\beta} \sum y_i} \cdot \left( \prod_{i=1}^n y_i \right)$$

by factorization thm,  $\sum y_i$  is SS for  $\beta$

As we know,  $\sum y_i \sim \text{gamma}(2n, \beta)$ ,  $E(\sum y_i) = 2n\beta$ .

$$(b) \text{ by (a), MVUE of } \beta : \frac{\sum y_i}{2n}, \quad \text{Var}\left(\frac{\sum y_i}{2n}\right) = \frac{\beta^2}{2n}$$

(c) done in class

(d) let  $\tilde{\beta} = \frac{Y_1}{2}$ ,  $E(\tilde{\beta}) = \beta$ , but does not converge if  $n \rightarrow \infty$

So  $\tilde{\beta}$  not consistent estimator.

$$16. (a) f(y | \theta) = \frac{3y^2}{\theta^3} \quad (0 \leq y \leq \theta) = \frac{3y^2}{\theta^3} \cdot I_{(0, \theta)}(y), \quad I_{(0, \theta)}(y) \begin{cases} 1 & (0 < y < \theta) \\ 0 & \text{o.w.} \end{cases}$$

$$L(\theta) = \frac{3^n \cdot \prod_{i=1}^n y_i^2}{\theta^{3n}} \cdot \prod_{i=1}^n I_{(0, \theta)}(y_i) = \frac{3^n \cdot \prod_{i=1}^n y_i^2}{\theta^{3n}} \cdot I_{(0, \theta)}(Y_{(n)})$$

By factorization thm,  $Y_{(n)}$  is SS for  $\theta$ .

$$(b) E(Y_{(n)}) = \frac{3n}{3n+1} \theta \quad \leftarrow \text{check by yourself. (using PDF / CDF of } Y_{(n)})$$

Since  $Y_{(n)}$  is SS for  $\theta$ ,

$\frac{3n+1}{3n} Y_{(n)}$  is MVUE for  $\theta$ . (by Rao-Blackwell thm)

$$18. P(\bar{Y}-1 < U < \bar{Y}+1) = P(-\bar{Y}-1 < -U < -\bar{Y}+1)$$

$$= P\left(\frac{-1}{\frac{3}{\sqrt{n}}} < \frac{\bar{Y}-U}{\frac{3}{\sqrt{n}}} = Z < \frac{1}{\frac{3}{\sqrt{n}}}\right) = P\left(|Z| < \frac{\sqrt{n}}{3}\right) = 0.95$$

$$\frac{\sqrt{n}}{3} \approx 1.96 \Rightarrow n \approx 35$$

20. See HWq 8.93

22. (a), (b)  $f(y_1, \dots, y_n | \theta) = (\theta+1)^n \left( \prod_{i=1}^n y_i \right)^\theta$

$$\log f(y_1, \dots, y_n | \theta) = n \log(\theta+1) + \theta \sum_{i=1}^n \log y_i$$

By factorization thm,  $W_n = \sum \log y_i$  is SS for  $\theta$

(c) Let  $X = -\log Y$ ,  $f_X(x) = (\theta+1) e^{-(\theta+1)x} \cdot | -e^{-x} | = (\theta+1) e^{-(\theta+1)x}$  ( $x > 0$ )

then  $X \sim \exp(\theta+1)$

$$E(X) = \frac{1}{\theta+1} = \eta, \quad E(\bar{X}) = E\left(\frac{1}{n} \sum -\log y_i\right) = \frac{1}{\theta+1}$$

Since  $\sum \log y_i$  is M.S.S.,  $-\frac{1}{n} \sum \log y_i$  is MVUE of  $\frac{1}{\theta+1}$

(d)  $\text{Var}\left(-\frac{1}{n} \sum \log y_i\right) = \frac{1}{n(\theta+1)^2} \xrightarrow{n \rightarrow \infty} 0$  ; consistent.

(e) by (d), I suggest  $\hat{\theta} = \frac{1}{-\frac{1}{n} \sum \log y_i} - 1$