Examples for 10/19/2020 (3) (continued)

5. The Weibull distribution has many applications in reliability engineering, survival analysis, and general insurance. Let  $\beta > 0$ ,  $\delta > 0$ . Consider the probability density function

$$f(x; \beta, \delta) = \beta \delta x^{\delta-1} e^{-\beta x^{\delta}}, \quad x > 0,$$
 zero otherwise.

Recall:  $W = X^{\delta}$  has an Exponential  $(\theta = \frac{1}{\beta}) = Gamma(\alpha = 1, \theta = \frac{1}{\beta})$  distribution.

Let  $X_1, X_2, \dots, X_n$  be a random sample from the above probability distribution.

$$\Rightarrow Y = \sum_{i=1}^{n} X_{i}^{\delta} = \sum_{i=1}^{n} W_{i} \text{ has a Gamma} (\alpha = n, \theta = \frac{1}{\beta}) \text{ distribution.}$$

Suppose  $\delta$  is known.

- k) Suggest a confidence interval for  $\beta$  with  $(1 \alpha)100\%$  confidence level.
- 1) Suppose  $\delta = 3$ , n = 5, and  $x_1 = 0.2$ ,  $x_2 = 1.2$ ,  $x_3 = 0.2$ ,  $x_4 = 0.9$ ,  $x_5 = 0.3$ . Use part (k) to construct a 90% confidence interval for  $\beta$ .
- m) Find a sufficient statistic  $Y = u(X_1, X_2, ..., X_n)$  for  $\beta$ .
- n) Find the Fisher information  $I(\beta)$ .
- o) Recall that  $\hat{\beta} = \frac{n-1}{\sum_{i=1}^{n} X_{i}^{\delta}}$  is an unbiased estimator of β.

Is  $\hat{\beta}$  an efficient estimator of  $\beta$ ? If  $\hat{\beta}$  is not an efficient estimator of  $\beta$ , find its efficiency. "Hint": Recall Examples for 10/19/2020 (3)  $\mathbf{2}$  (g).

## **Answers:**

5. The Weibull distribution has many applications in reliability engineering, survival analysis, and general insurance. Let  $\beta > 0$ ,  $\delta > 0$ . Consider the probability density function

$$f(x; \beta, \delta) = \beta \delta x^{\delta-1} e^{-\beta x^{\delta}}, \quad x > 0,$$
 zero otherwise.

Recall:  $W = X^{\delta}$  has an Exponential  $(\theta = \frac{1}{\beta}) = Gamma(\alpha = 1, \theta = \frac{1}{\beta})$  distribution.

Let  $X_1, X_2, \dots, X_n$  be a random sample from the above probability distribution.

$$\Rightarrow$$
  $Y = \sum_{i=1}^{n} X_{i}^{\delta} = \sum_{i=1}^{n} W_{i}$  has a Gamma  $(\alpha = n, \theta = \frac{1}{\beta})$  distribution.

Suppose  $\delta$  is known.

k) Suggest a confidence interval for  $\beta$  with  $(1 - \alpha)100\%$  confidence level.

$$Y = \sum_{i=1}^{n} X_{i}^{\delta} = \sum_{i=1}^{n} W_{i}$$
 has a Gamma distribution with  $\alpha = n$  and  $\theta = \frac{1}{\beta}$  ( $\lambda = \beta$ ).

If  $T_{\alpha}$  has a Gamma  $(\alpha, \theta = 1/\lambda)$  distribution, where  $\alpha$  is an integer, then  ${}^{2}T_{\alpha}/_{\theta} = 2\lambda T_{\alpha}$  has a  $\chi^{2}(2\alpha)$  distribution (a chi-square distribution with  $2\alpha$  degrees of freedom).

$${}^{2}Y/_{\theta} = 2 \beta \sum_{i=1}^{n} X_{i}^{\delta}$$
 has a chi-square distribution with  $r = 2 \alpha = 2 n$  d.f.

$$\Rightarrow \qquad \mathrm{P} \left( \ \chi_{1-\alpha/2}^{\, 2} \big( 2n \, \right) \, \leq \, 2 \, \beta \, \sum_{i=1}^{n} \mathrm{X}_{\, i}^{\, \delta} \ \, \leq \, \chi_{\, \alpha/2}^{\, 2} \big( 2n \, \big) \, \right) \, = \, 1 - \alpha.$$

$$\Rightarrow P\left(\frac{\chi_{1-\alpha/2}^{2}(2n)}{2\sum_{i=1}^{n}X_{i}^{\delta}} < \beta < \frac{\chi_{\alpha/2}^{2}(2n)}{2\sum_{i=1}^{n}X_{i}^{\delta}}\right) = 1 - \alpha.$$

A 
$$(1-\alpha)$$
 100 % confidence interval for  $\beta$ :
$$\left(\begin{array}{c} \chi_{1-\alpha/2}^2(2n), & \chi_{\alpha/2}^2(2n) \\ 2\sum\limits_{i=1}^n X_i^{\delta}, & 2\sum\limits_{i=1}^n X_i^{\delta} \end{array}\right).$$

1) Suppose  $\delta = 3$ , n = 5, and  $x_1 = 0.2$ ,  $x_2 = 1.2$ ,  $x_3 = 0.2$ ,  $x_4 = 0.9$ ,  $x_5 = 0.3$ . Use part (k) to construct a 90% confidence interval for  $\beta$ .

$$\sum_{i=1}^{n} x_i^3 = 2.5.$$

$$\chi^{\,2}_{\,0.95}(\,10\,\,) \,=\, 3.940, \qquad \chi^{\,2}_{\,0.05}(\,10\,\,) \,=\, 18.31.$$

$$\left(\frac{3.940}{2 \cdot 2.5}, \frac{18.31}{2 \cdot 2.5}\right)$$
 (0.788, 3.662)

m) Find a sufficient statistic  $Y = u(X_1, X_2, ..., X_n)$  for  $\beta$ .

$$\begin{split} f(x_1, x_2, \dots x_n; \beta) &= f(x_1; \beta) \ f(x_2; \beta) \ \dots \ f(x_n; \beta) \\ &= \prod_{i=1}^n \left( \beta \delta x_i^{\delta - 1} e^{-\beta x_i^{\delta}} \right) = \left[ \beta^n e^{-\beta \sum x_i^{\delta}} \right] \left( \delta^n \prod_{i=1}^n x_i^{\delta - 1} \right). \end{split}$$

By Factorization Theorem,  $Y = \sum_{i=1}^{n} X_{i}^{\delta}$  is a sufficient statistic for  $\beta$ .

OR

$$f(x; \beta) = \exp \{-\beta x^{\delta} + \ln \beta + \ln \delta + (\delta - 1) \ln x\}. \qquad \Rightarrow \qquad K(x) = x^{\delta}.$$

$$\Rightarrow \qquad Y = \sum_{i=1}^{n} X_{i}^{\delta} \text{ is a sufficient statistic for } \beta.$$

n) Find the Fisher information  $I(\beta)$ .

$$\ln f(x;\beta) = -\beta x^{\delta} + \ln \beta + \ln \delta + (\delta - 1) \ln x.$$

$$\frac{\partial}{\partial \beta} \ln f(x; \beta) = -x^{\delta} + \frac{1}{\beta}. \qquad \qquad \frac{\partial^2}{\partial \beta^2} \ln f(x; \beta) = -\frac{1}{\beta^2}.$$

$$I(\beta) = \operatorname{Var}\left[\frac{\partial}{\partial \beta} \ln f(X; \beta)\right]$$

$$= \operatorname{Var}\left[-X^{\delta} + \frac{1}{\beta}\right]$$

$$= \operatorname{Var}(W)$$

$$= \alpha \theta^{2} = \frac{1}{\beta^{2}}.$$

$$I(\beta) = -\operatorname{E}\left[\frac{\partial^{2}}{\partial \beta^{2}} \ln f(X; \beta)\right]$$

$$= -\operatorname{E}\left[-\frac{1}{\beta^{2}}\right]$$

$$= \frac{1}{\beta^{2}}.$$

o) Recall that 
$$\hat{\beta} = \frac{n-1}{\sum_{i=1}^{n} X_i^{\delta}}$$
 is an unbiased estimator of β.

Is  $\hat{\beta}$  an efficient estimator of  $\beta$ ? If  $\hat{\beta}$  is not an efficient estimator of  $\beta$ , find its efficiency. "Hint": Recall Examples for 10/19/2020 (3)  $\mathbf{2}$  (g).

Recall: 
$$\operatorname{Var}\left(\frac{1}{Y}\right) = \frac{\beta^2}{(n-1)(n-2)} - \left(\frac{\beta}{n-1}\right)^2 = \frac{\beta^2}{(n-1)^2(n-2)}.$$

$$\operatorname{Var}(\hat{\hat{\beta}}) = \operatorname{Var}(\frac{n-1}{Y}) = (n-1)^2 \operatorname{Var}(\frac{1}{Y}) = \frac{\beta^2}{n-2}.$$

Rao-Cramer lower bound = 
$$\frac{1}{n \cdot I(\beta)} = \frac{\beta^2}{n}$$
.

$$\operatorname{Var}(\hat{\hat{\beta}}) = \frac{\beta^2}{n-2} > \frac{\beta^2}{n}.$$

 $\text{Var}\big(\,\hat{\hat{\beta}}\,\big)\,$  does NOT attain its Rao-Cramer lower bound.

$$\Rightarrow \qquad \hat{\hat{\beta}} \ \ \text{is NOT an efficient estimator of} \ \ \beta \, ,$$

its efficiency = 
$$\frac{n-2}{n} \to 1$$
 as  $n \to \infty$ .