

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

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

Recall (Examples for 08/28/2020 (3)):

$W = X^\delta$ 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. !!!

- 1.** Suppose δ is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

- a) Find a closed-form expression for $E(X^k)$, $k > -\delta$.

“Hint” 1: $u = \beta x^\delta$ “Hint” 2: $\Gamma(a) = \int_0^\infty u^{a-1} e^{-u} du, \quad a > 0.$

- b) Obtain a method of moments estimator for $\beta, \tilde{\beta}$.

- c) Suppose $\delta = 3, n = 5$, and $x_1 = 0.2, x_2 = 1.2, x_3 = 0.2, x_4 = 0.9, x_5 = 0.3$. Obtain a method of moments estimate for $\beta, \tilde{\beta}$.

“Hint”: $\Gamma(a)$ R: $> \text{gamma}(a)$
Excel: $=\text{GAMMA}(a)$

2. Suppose δ is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

d) Obtain the maximum likelihood estimator for β , $\hat{\beta}$.

e) Suppose $\delta = 3$, $n = 5$, and $x_1 = 0.2$, $x_2 = 1.2$, $x_3 = 0.2$, $x_4 = 0.9$, $x_5 = 0.3$. Obtain the maximum likelihood estimate for β , $\hat{\beta}$.

f) Is the maximum likelihood estimator $\hat{\beta}$ an unbiased estimator of β ?
If $\hat{\beta}$ is not an unbiased estimator of β , construct an unbiased estimator of β based on $\hat{\beta}$.

“Hint” 1: $E(a \odot) = a E(\odot)$. “Hint” 2: $\frac{1}{\heartsuit} = \heartsuit^{-1}$.

“Hint” 3: If T_α has a Gamma($\alpha, \theta = 1/\lambda$) distribution, then

$$E(T_\alpha^k) = \frac{\theta^k \Gamma(\alpha + k)}{\Gamma(\alpha)} = \frac{\Gamma(\alpha + k)}{\lambda^k \Gamma(\alpha)}, \quad k > -\alpha.$$

g) Find $MSE(\hat{\beta})$, where $\hat{\beta}$ the maximum likelihood estimator of β .

“Hint” 1: $\text{bias}(\hat{\beta}) = E(\hat{\beta}) - \beta$. You have $E(\hat{\beta})$ from part (f).

“Hint” 2: $\text{Var}(a \odot) = a^2 \text{Var}(\odot)$. $\text{Var}(\odot) = E(\odot^2) - [E(\odot)]^2$.

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“Hint” 4: $MSE(\hat{\beta}) = (\text{bias}(\hat{\beta}))^2 + \text{Var}(\hat{\beta})$.

3. Suppose β is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

h) Obtain an equation for the maximum likelihood estimator for $\delta, \hat{\delta}$.

“Hint”: $\frac{d}{d\delta} \ln L(\delta) = 0.$

4. Suppose $\delta = 3, \beta = 4, n = 5$.

Let X_1, X_2, X_3, X_4, X_5 be a random sample of size $n = 5$ from the distribution with the probability density function

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

i) Find the probability $P\left(\sum_{i=1}^5 X_i^3 > 2.5\right) = P\left(\sum_{i=1}^n X_i^\delta > 2.5\right).$

“Hint”: If T_α has a $\text{Gamma}(\alpha, \theta = 1/\lambda)$ distribution, where α is an integer, then

$$P(T_\alpha \leq t) = P(X_t \geq \alpha) \quad \text{and} \quad P(T_\alpha > t) = P(X_t \leq \alpha - 1),$$

where X_t has a $\text{Poisson}(\lambda t)$ distribution.

j) Find a such that $P\left(\sum_{i=1}^5 X_i^3 > a\right) = P\left(\sum_{i=1}^n X_i^\delta > a\right) = 0.10.$

“Hint”: If T_α has a $\text{Gamma}(\alpha, \theta = 1/\lambda)$ distribution, where α is an integer, then

$2T_\alpha/\theta = 2\lambda T_\alpha$ has a $\chi^2(2\alpha)$ distribution (a chi-square distribution with 2α degrees of freedom).

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

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Recall (Examples for 08/28/2020 (3)):

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

1. Suppose δ is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

a) Find a closed-form expression for $E(X^k)$, $k > -\delta$.

“Hint” 1: $u = \beta x^\delta$ “Hint” 2: $\Gamma(a) = \int_0^\infty u^{a-1} e^{-u} du, \quad a > 0.$

$$\begin{aligned} E(X^k) &= \int_0^\infty x^k \beta \delta x^{\delta-1} e^{-\beta x^\delta} dx & u = \beta x^\delta & \quad du = \beta \delta x^{\delta-1} dx \\ &= \int_0^\infty \left(\frac{u}{\beta}\right)^{k/\delta} e^{-u} du = \frac{1}{\beta^{k/\delta}} \Gamma\left(\frac{k}{\delta} + 1\right), & k > -\delta. \end{aligned}$$

OR

$$\begin{aligned}
 E(X^k) &= \int_0^{\infty} x^k \beta \delta x^{\delta-1} e^{-\beta x^{\delta}} dx & u &= x^{\delta} & du &= \delta x^{\delta-1} dx \\
 &= \int_0^{\infty} u^{k/\delta} \beta e^{-\beta u} du = \frac{\Gamma\left(\frac{k}{\delta}+1\right)}{\beta^{k/\delta}} \int_0^{\infty} \frac{\beta^{k/\delta+1}}{\Gamma\left(\frac{k}{\delta}+1\right)} u^{k/\delta+1-1} \beta e^{-\beta u} du \\
 &= \frac{1}{\beta^{k/\delta}} \Gamma\left(\frac{k}{\delta}+1\right),
 \end{aligned}$$

since $\frac{\beta^{k/\delta+1}}{\Gamma\left(\frac{k}{\delta}+1\right)} u^{k/\delta+1-1} \beta e^{-\beta u}$ is the p.d.f. of

Gamma($\alpha = \frac{k}{\delta}+1, \theta = \frac{1}{\beta}$) distribution.

OR

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

If T_{α} has a Gamma($\alpha, \theta = 1/\lambda$) distribution, then

$$E(T_{\alpha}^k) = \frac{\theta^k \Gamma(\alpha+k)}{\Gamma(\alpha)} = \frac{\Gamma(\alpha+k)}{\lambda^k \Gamma(\alpha)}, \quad k > -\alpha.$$

$$E(X^k) = E(W^{k/\delta}) = \frac{\Gamma\left(1+\frac{k}{\delta}\right)}{\beta^{k/\delta} \Gamma(1)} = \frac{1}{\beta^{k/\delta}} \Gamma\left(\frac{k}{\delta}+1\right), \quad \frac{k}{\delta} > -1.$$

- b) Obtain a method of moments estimator for β , $\tilde{\beta}$.

$$E(X) = \frac{1}{\beta^{1/\delta}} \Gamma\left(\frac{1}{\delta} + 1\right). \quad \bar{X} = \frac{1}{\tilde{\beta}^{1/\delta}} \Gamma\left(\frac{1}{\delta} + 1\right).$$

$$\tilde{\beta} = \left(\frac{\Gamma\left(\frac{1}{\delta} + 1\right)}{\bar{X}} \right)^{\delta}.$$

- c) Suppose $\delta = 3$, $n = 5$, and $x_1 = 0.2$, $x_2 = 1.2$, $x_3 = 0.2$, $x_4 = 0.9$, $x_5 = 0.3$.
Obtain a method of moments estimate for β , $\tilde{\beta}$.

“Hint”: $\Gamma(a)$ R: `> gamma(a)`
Excel: `=GAMMA(a)`

$$\bar{x} = \frac{2.8}{5} = 0.56.$$

$$\Gamma\left(\frac{4}{3}\right) \approx 0.89298.$$

$$\tilde{\beta} = \left(\frac{0.89298}{0.56} \right)^3 \approx \mathbf{4.0547}.$$

OR

$$\text{b) } E(X^2) = \frac{1}{\beta^{2/\delta}} \Gamma\left(\frac{2}{\delta} + 1\right). \quad \overline{X^2} = \frac{1}{\tilde{\beta}^{2/\delta}} \Gamma\left(\frac{2}{\delta} + 1\right).$$

$$\tilde{\beta} = \left(\frac{\Gamma\left(\frac{2}{\delta} + 1\right)}{\overline{X^2}} \right)^{\delta/2}.$$

$$\text{c) } \overline{x^2} = \frac{2.42}{5} = 0.484. \quad \Gamma\left(\frac{5}{3}\right) \approx 0.902745.$$

$$\tilde{\beta}_2 = \left(\frac{0.902745}{0.484} \right)^{1.5} \approx \mathbf{2.5473}.$$

OR

$$\text{b) } E(X^3) = \frac{1}{\beta^{3/\delta}} \Gamma\left(\frac{3}{\delta} + 1\right). \quad \overline{X^3} = \frac{1}{\tilde{\beta}^{3/\delta}} \Gamma\left(\frac{3}{\delta} + 1\right).$$

$$\tilde{\beta} = \left(\frac{\Gamma\left(\frac{3}{\delta} + 1\right)}{\overline{X^3}} \right)^{\delta/3}.$$

$$\text{c) } \overline{x^3} = \frac{2.5}{5} = 0.50. \quad \Gamma(2) = 1.$$

$$\tilde{\beta}_3 = \left(\frac{1}{0.50} \right)^1 = \mathbf{2}.$$

OR

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2. Suppose δ is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

d) Obtain the maximum likelihood estimator for β , $\hat{\beta}$.

$$L(\beta) = \prod_{i=1}^n \left(\beta \delta x_i^{\delta-1} e^{-\beta x_i^\delta} \right).$$

$$\ln L(\beta) = n \cdot \ln \beta + n \cdot \ln \delta + (\delta - 1) \cdot \sum_{i=1}^n \ln x_i - \beta \cdot \sum_{i=1}^n x_i^\delta.$$

$$\frac{d}{d\beta} \ln L(\beta) = \frac{n}{\beta} - \sum_{i=1}^n x_i^\delta = 0. \quad \Rightarrow \quad \hat{\beta} = \frac{n}{\sum_{i=1}^n x_i^\delta}.$$

e) Suppose $\delta = 3$, $n = 5$, and $x_1 = 0.2$, $x_2 = 1.2$, $x_3 = 0.2$, $x_4 = 0.9$, $x_5 = 0.3$.
Obtain the maximum likelihood estimate for β , $\hat{\beta}$.

$$\sum_{i=1}^n x_i^3 = 2.5. \quad \hat{\beta} = \frac{5}{2.5} = 2.$$

- f) Is the maximum likelihood estimator $\hat{\beta}$ an unbiased estimator of β ?
 If $\hat{\beta}$ is not an unbiased estimator of β , construct an unbiased estimator of β based on $\hat{\beta}$.

“Hint” 1: $E(a \odot) = a E(\odot)$. “Hint” 2: $\frac{1}{\heartsuit} = \heartsuit^{-1}$.

“Hint” 3: If T_α has a $\text{Gamma}(\alpha, \theta = 1/\lambda)$ distribution, then

$$E(T_\alpha^k) = \frac{\theta^k \Gamma(\alpha + k)}{\Gamma(\alpha)} = \frac{\Gamma(\alpha + k)}{\lambda^k \Gamma(\alpha)}, \quad k > -\alpha.$$

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

$$\hat{\beta} = \frac{n}{\sum_{i=1}^n X_i^\delta} = \frac{n}{Y}. \quad a = n, \odot = \frac{1}{Y}, \heartsuit = Y.$$

$$E\left(\frac{1}{Y}\right) = E(Y^{-1}) = \frac{\Gamma(n-1)}{\beta^{-1} \Gamma(n)} = \frac{\beta}{n-1}.$$

$$E(\hat{\beta}) = E\left(\frac{n}{Y}\right) = n E\left(\frac{1}{Y}\right) = n \cdot \frac{\beta}{n-1} = \frac{n}{n-1} \cdot \beta = \beta + \frac{\beta}{n-1} \neq \beta.$$

$$\hat{\beta} \text{ is NOT an unbiased estimator of } \beta. \quad \text{bias}(\hat{\beta}) = E(\hat{\beta}) - \beta = \frac{\beta}{n-1}.$$

$$\text{Consider } \hat{\tilde{\beta}} = \frac{n-1}{n} \cdot \hat{\beta} = \frac{n-1}{n} \cdot \frac{n}{\sum_{i=1}^n X_i^\delta}. \quad \text{Then } E(\hat{\tilde{\beta}}) = \frac{n-1}{n} \cdot E(\hat{\beta}) = \beta.$$

$\hat{\tilde{\beta}}$ is an unbiased estimator of β .

g) Find $\text{MSE}(\hat{\beta})$, where $\hat{\beta}$ the maximum likelihood estimator of β .

“Hint” 1: $\text{bias}(\hat{\beta}) = E(\hat{\beta}) - \beta$. You have $E(\hat{\beta})$ from part (f).

“Hint” 2: $\text{Var}(a \odot) = a^2 \text{Var}(\odot)$. $\text{Var}(\odot) = E(\odot^2) - [E(\odot)]^2$.

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$$E(T_\alpha^k) = \frac{\theta^k \Gamma(\alpha + k)}{\Gamma(\alpha)} = \frac{\Gamma(\alpha + k)}{\lambda^k \Gamma(\alpha)}, \quad k > -\alpha.$$

“Hint” 4: $\text{MSE}(\hat{\beta}) = (\text{bias}(\hat{\beta}))^2 + \text{Var}(\hat{\beta})$.

$$\text{bias}(\hat{\beta}) = E(\hat{\beta}) - \beta = \frac{\beta}{n-1}.$$

$$E\left(\frac{1}{Y^2}\right) = E(Y^{-2}) = \frac{\Gamma(n-2)}{\beta^{-2} \Gamma(n)} = \frac{\beta^2}{(n-1)(n-2)}.$$

$$\begin{aligned} \text{Var}(\hat{\beta}) &= \text{Var}\left(\frac{n}{Y}\right) = n^2 \text{Var}\left(\frac{1}{Y}\right) = n^2 \left[\frac{\beta^2}{(n-1)(n-2)} - \frac{\beta^2}{(n-1)^2} \right] \\ &= \frac{n^2 \beta^2}{(n-1)^2 (n-2)}. \end{aligned}$$

$$\begin{aligned} \text{MSE}(\hat{\beta}) &= (\text{bias}(\hat{\beta}))^2 + \text{Var}(\hat{\beta}) = \frac{\beta^2}{(n-1)^2} + \frac{n^2 \beta^2}{(n-1)^2 (n-2)} \\ &= \frac{(n^2 + n - 2) \beta^2}{(n-1)^2 (n-2)} = \frac{(n+2) \beta^2}{(n-1)(n-2)}. \end{aligned}$$

3. Suppose β is known.

Let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

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

h) Obtain an equation for the maximum likelihood estimator for δ , $\hat{\delta}$.

“Hint”: $\frac{d}{d\delta} \ln L(\delta) = 0$.

$$L(\delta) = \prod_{i=1}^n \left(\beta \delta x_i^{\delta-1} e^{-\beta x_i^\delta} \right).$$

$$\ln L(\delta) = n \cdot \ln \beta + n \cdot \ln \delta + (\delta - 1) \cdot \sum_{i=1}^n \ln x_i - \beta \cdot \sum_{i=1}^n x_i^\delta.$$

$$\frac{d}{d\delta} \ln L(\delta) = \frac{n}{\delta} + \sum_{i=1}^n \ln x_i - \beta \cdot \sum_{i=1}^n x_i^\delta \ln x_i = 0.$$

This equation cannot be solved algebraically for δ in closed form.

The solution could be approximated for given x_1, x_2, \dots, x_n by using iterative numerical procedures.

4. Suppose $\delta = 3$, $\beta = 4$, $n = 5$.

Let X_1, X_2, X_3, X_4, X_5 be a random sample of size $n = 5$ from the distribution with the probability density function

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

i) Find the probability $P\left(\sum_{i=1}^5 X_i^3 > 2.5\right) = P\left(\sum_{i=1}^n X_i^\delta > 2.5\right)$.

“Hint”: If T_α has a Gamma($\alpha, \theta = 1/\lambda$) distribution, where α is an integer, then

$$P(T_\alpha \leq t) = P(X_t \geq \alpha) \quad \text{and} \quad P(T_\alpha > t) = P(X_t \leq \alpha - 1),$$

where X_t has a Poisson(λt) distribution.

$Y = \sum_{i=1}^5 X_i^3 = \sum_{i=1}^n X_i^\delta$ has a Gamma($\alpha = n = 5, \theta = \frac{1}{\beta} = \frac{1}{4}$) distribution.

$$\begin{aligned} P\left(\sum_{i=1}^5 X_i^3 > 2.5\right) &= P(T_5 > 2.5) = P(\text{Poisson}(\beta \cdot 2.5) \leq 5 - 1) \\ &= P(\text{Poisson}(4 \cdot 2.5) \leq 4) = P(\text{Poisson}(10) \leq 4) = \mathbf{0.029}. \end{aligned}$$

OR

$$P(T_5 > 2.5) = \int_{2.5}^{\infty} \frac{4^5}{\Gamma(5)} t^{5-1} e^{-4t} dt = \int_{2.5}^{\infty} \frac{4^5}{4!} t^4 e^{-4t} dt = \dots \approx 0.029253.$$

$$\text{OR} \quad P(T_5 > 2.5) = P(2\beta T_5 > 2\beta \cdot 2.5) = P(\chi^2(10) > 20) \approx 0.029253.$$

j) Find a such that $P\left(\sum_{i=1}^5 X_i^3 > a\right) = P\left(\sum_{i=1}^n X_i^\delta > a\right) = 0.10$.

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

$$Y = \sum_{i=1}^5 X_i^3 = \sum_{i=1}^n X_i^\delta \text{ has a } \text{Gamma}(\alpha = n = 5, \theta = \frac{1}{\beta} = \frac{1}{4}) \text{ distribution.}$$

$$2Y/\theta = 2\beta \sum_{i=1}^n X_i^3 \text{ has a chi-square distribution with } r = 2\alpha = 2n = 10 \text{ d.f.}$$

$$\Rightarrow P\left(2\beta \sum_{i=1}^5 X_i^3 > \chi_{0.10}^2(10)\right) = 0.10.$$

$$\chi_{0.10}^2(10) = 15.99. \quad \Rightarrow \quad 2\beta a = 8a = 15.99.$$

$$\Rightarrow \quad a = \mathbf{1.99875}.$$

OR

$$P\left(\sum_{i=1}^5 X_i^3 > a\right) = P(T_5 > a) = P(\text{Poisson}(\beta \cdot a) \leq 5 - 1) = P(\text{Poisson}(4 \cdot a) \leq 4).$$

$$P(\text{Poisson}(8) \leq 4) = 0.10. \quad \Rightarrow \quad 4a = 8.$$

$$\Rightarrow \quad a = \mathbf{2}.$$

$$a \approx \frac{15.98718}{8} \approx 1.9984.$$