

1. The Weibull distribution has many applications in reliability engineering, survival analysis, and general insurance. Let $\beta > 0, \delta > 0$. Let X_1, X_2, \dots, X_n be a random sample of size n from the distribution with the probability density function

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

Suppose δ is known.

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

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

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

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

- b) 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 , $\hat{\beta}$.

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

Suppose β is known.

- c) 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.

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

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

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

Suppose δ is known.

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

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

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

- f) 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”: $\Gamma(x)$ R: `> gamma(x)`
 Excel: `=GAMMA(x)`

$$\bar{x} = \frac{2.8}{5} = 0.56. \quad \Gamma\left(\frac{4}{3}\right) \approx 0.89298.$$

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

More on the Method of Moments:

$$E(X) = h(\theta). \quad \text{Set } \bar{X} = h(\tilde{\theta}). \quad \text{Solve for } \tilde{\theta}.$$

OR

$$E(X^k) = h(\theta). \quad \text{Set } \overline{X^k} = h(\tilde{\theta}), \quad \text{where } \overline{X^k} = \frac{1}{n} \sum_{i=1}^n X_i^k.$$

Solve for $\tilde{\theta}$.

OR

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

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

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

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

OR

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

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

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

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

OR

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

$$f_X(x) = f_X(x; \theta) = (\theta^2 + \theta) x^{\theta-1} (1-x), \quad 0 < x < 1, \quad \theta > 0.$$

- a) Obtain the method of moments estimator of θ , $\tilde{\theta}$.

$$\begin{aligned} E(X) &= \int_0^1 x \cdot (\theta^2 + \theta) x^{\theta-1} (1-x) dx = (\theta^2 + \theta) \cdot \int_0^1 (x^\theta - x^{\theta+1}) dx \\ &= \theta \cdot (\theta+1) \cdot \left(\frac{1}{\theta+1} x^{\theta+1} - \frac{1}{\theta+2} x^{\theta+2} \right) \Big|_0^1 = \frac{\theta \cdot (\theta+1)}{(\theta+1) \cdot (\theta+2)} = \frac{\theta}{\theta+2}. \end{aligned}$$

OR

$$\text{Beta distribution, } \alpha = \theta, \beta = 2. \quad \Rightarrow \quad E(X) = \frac{\theta}{\theta+2}.$$

$$\frac{\tilde{\theta}}{\tilde{\theta}+2} = \bar{X} \quad \tilde{\theta} = \bar{X} \cdot (\tilde{\theta} + 2) \quad \tilde{\theta} - \tilde{\theta} \bar{X} = 2 \bar{X}$$

$$\Rightarrow \quad \tilde{\theta} = \frac{2\bar{X}}{1-\bar{X}}, \quad \text{where } \bar{X} = \frac{1}{n} \cdot \sum_{i=1}^n X_i.$$

- b) Suppose $n = 6$, and $x_1 = 0.3, x_2 = 0.5, x_3 = 0.6, x_4 = 0.65, x_5 = 0.75, x_6 = 0.8$. Find a method of moments estimate of θ .

$$x_1 = 0.3, x_2 = 0.5, x_3 = 0.6, x_4 = 0.65, x_5 = 0.75, x_6 = 0.8.$$

$$\bar{x} = 0.6. \quad \tilde{\theta} = \frac{2\bar{x}}{1-\bar{x}} = \mathbf{3}.$$

c) Is $\tilde{\theta}$ an unbiased estimator of θ ? *Justify your answer.*

Consider $g(x) = \frac{2x}{1-x}$. Then $g(\bar{X}) = \tilde{\theta}$, $g\left(\frac{\theta}{\theta+2}\right) = \theta$.

Also $g''(x) = \frac{4}{(1-x)^3} > 0$ for $0 < x < 1$, i.e., $g(x)$ is strictly convex.

By Jensen's Inequality,

$$E(\tilde{\theta}) = E[g(\bar{X})] > g(E(\bar{X})) = g(\mu_X) = g\left(\frac{\theta}{\theta+2}\right) = \theta.$$

Therefore, $\tilde{\theta}$ is NOT an unbiased estimator of θ .

d) Obtain the maximum likelihood estimator of θ , $\hat{\theta}$.

That is, find $\hat{\theta} = \arg \max L(\theta) = \arg \max \ln L(\theta)$,

$$\text{where } L(\theta) = \prod_{i=1}^n f(x_i; \theta).$$

“Hint”: ① $\frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$;

② $\theta > 0$;

③ Since $0 < x < 1$, $\ln x < 0$.

$$L(\theta) = \prod_{i=1}^n f(x_i; \theta) = f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta)$$

$$= (\theta^2 + \theta)^n \left(\prod_{i=1}^n x_i \right)^{\theta-1} \prod_{i=1}^n (1-x_i).$$

$$\ln L(\theta) = n \ln(\theta^2 + \theta) + (\theta - 1) \sum_{i=1}^n \ln x_i + \sum_{i=1}^n \ln(1 - x_i)$$

$$(\ln L(\theta))' = \frac{2n\theta + n}{\theta^2 + \theta} + \sum_{i=1}^n \ln x_i = 0.$$

$$\Rightarrow \quad \Sigma \hat{\theta}^2 + (2n + \Sigma) \hat{\theta} + n = 0, \quad \text{where } \Sigma = \sum_{i=1}^n \ln X_i.$$

$$\Rightarrow \quad \hat{\theta} = \frac{-2n - \Sigma \pm \sqrt{(2n + \Sigma)^2 - 4\Sigma n}}{2\Sigma} = \frac{2n + \Sigma \pm \sqrt{4n^2 + \Sigma^2}}{-2\Sigma}.$$

$$\text{Since } 0 < x < 1, \ln x < 0. \quad \Rightarrow \quad \Sigma < 0.$$

$$\Rightarrow \quad (2n + \Sigma)^2 = 4n^2 + 4n\Sigma + \Sigma^2 < 4n^2 + \Sigma^2.$$

$$\Rightarrow \quad |2n + \Sigma| < \sqrt{4n^2 + \Sigma^2}.$$

$$\Rightarrow \quad \text{Since } \theta > 0, \quad \hat{\theta} = \frac{2n + \Sigma + \sqrt{4n^2 + \Sigma^2}}{-2\Sigma},$$

$$\text{where } \Sigma = \sum_{i=1}^n \ln X_i.$$

- e) Suppose $n = 6$, and $x_1 = 0.3$, $x_2 = 0.5$, $x_3 = 0.6$, $x_4 = 0.65$, $x_5 = 0.75$, $x_6 = 0.8$.
Find the maximum likelihood estimate of θ .

$$\Sigma = \sum_{i=1}^n \ln X_i \approx -3.349554. \quad \hat{\theta} \approx \mathbf{3.151}.$$

3. Let $\lambda > 0$ and let X_1, X_2, \dots, X_n be a random sample of size n from a double exponential distribution. That is,

$$f(x; \lambda) = \frac{\lambda}{2} e^{-\lambda|x|}, \quad -\infty < x < \infty.$$

- a) Find $E(X^k)$ for positive integer k .

$$E(X^k) = \int_{-\infty}^{\infty} x^k \cdot \frac{\lambda}{2} e^{-\lambda|x|} dx = \int_{-\infty}^0 x^k \cdot \frac{\lambda}{2} e^{\lambda x} dx + \int_0^{\infty} x^k \cdot \frac{\lambda}{2} e^{-\lambda x} dx = \dots$$

$$k \text{ odd} \quad \dots = 0.$$

$$\begin{aligned} k \text{ even} \quad \dots &= 2 \cdot \int_0^{\infty} x^k \cdot \frac{\lambda}{2} e^{-\lambda x} dx = \int_0^{\infty} \lambda \cdot x^k e^{-\lambda x} dx \\ &= \frac{\Gamma(k+1)}{\lambda^k} \cdot \int_0^{\infty} \frac{\lambda^{k+1}}{\Gamma(k+1)} x^{k+1-1} e^{-\lambda x} dx = \frac{\Gamma(k+1)}{\lambda^k} = \frac{k!}{\lambda^k}. \end{aligned}$$

- b) Obtain the maximum likelihood estimator of λ , $\hat{\lambda}$.

$$L(\lambda) = \frac{\lambda^n}{2^n} \exp\left\{-\lambda \cdot \sum_{i=1}^n |x_i|\right\}. \quad \ln L(\lambda) = n \ln \lambda - n \ln 2 - \lambda \cdot \sum_{i=1}^n |x_i|.$$

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

4. Let $\lambda > 0$ and let X_1, X_2, \dots, X_n be a random sample from the distribution with the probability density function

$$f(x; \lambda) = \lambda^3 x^5 e^{-\lambda x^2}, \quad x > 0.$$

- a) Find $E(X^k)$, $k > -6$. "Hint": Consider $u = \lambda x^2$.

$$\begin{aligned} E(X^k) &= \int_0^{\infty} x^k \cdot \lambda^3 x^5 e^{-\lambda x^2} dx & u &= \lambda x^2 & du &= 2\lambda x \\ &= \frac{\lambda^2}{2} \cdot \int_0^{\infty} \left(\frac{u}{\lambda}\right)^{2+\frac{k}{2}} e^{-u} du = \frac{1}{2} \lambda^{-k/2} \cdot \int_0^{\infty} u^{2+k/2} e^{-u} du \\ &= \frac{1}{2} \lambda^{-k/2} \Gamma\left(3 + \frac{k}{2}\right). \end{aligned}$$

- b) Obtain the method of moments estimator of λ , $\tilde{\lambda}$. Suppose $n = 4$, and $x_1 = 4$, $x_2 = 2$, $x_3 = 4$, $x_4 = 3$. Find the method of moments estimate of λ .

$$\begin{aligned} E(X) &= \frac{1}{2} \lambda^{-1/2} \Gamma\left(3 + \frac{1}{2}\right) = \frac{1}{2} \lambda^{-1/2} \cdot \frac{5}{2} \cdot \Gamma\left(2 + \frac{1}{2}\right) = \frac{1}{2} \lambda^{-1/2} \cdot \frac{5}{2} \cdot \frac{3}{2} \cdot \Gamma\left(1 + \frac{1}{2}\right) \\ &= \frac{1}{2} \lambda^{-1/2} \cdot \frac{5}{2} \cdot \frac{3}{2} \cdot \frac{1}{2} \cdot \Gamma\left(\frac{1}{2}\right) = \frac{1}{2} \lambda^{-1/2} \cdot \frac{5}{2} \cdot \frac{3}{2} \cdot \frac{1}{2} \cdot \sqrt{\pi} = \frac{15}{16} \cdot \sqrt{\frac{\pi}{\lambda}}. \end{aligned}$$

$$\frac{15}{16} \cdot \sqrt{\frac{\pi}{\lambda}} = \bar{X} \quad \Rightarrow \quad \tilde{\lambda} = \frac{225 \pi}{256 (\bar{X})^2}.$$

$$x_1 = 4, \quad x_2 = 2, \quad x_3 = 4, \quad x_4 = 3. \quad \bar{x} = \frac{13}{4} = 3.25.$$

$$\tilde{\lambda} = \frac{225 \pi}{256 (\bar{X})^2} = \frac{225 \pi}{2704} \approx \mathbf{0.2614}.$$

OR

$$E(X^2) = \frac{1}{2} \lambda^{-2/2} \Gamma\left(3 + \frac{2}{2}\right) = \frac{1}{2} \lambda^{-1} \cdot \Gamma(4) = \frac{3!}{2\lambda} = \frac{3}{\lambda}.$$

$$\frac{3}{\lambda} = \overline{X^2} = \frac{1}{n} \cdot \sum_{i=1}^n X_i^2 \quad \Rightarrow \quad \tilde{\lambda}_2 = \frac{3}{\overline{X^2}} = \frac{3n}{\sum_{i=1}^n X_i^2}.$$

$$x_1 = 4, \quad x_2 = 2, \quad x_3 = 4, \quad x_4 = 3. \quad \sum_{i=1}^n x_i^2 = 45.$$

$$\tilde{\lambda}_2 = \frac{3n}{\sum_{i=1}^n x_i^2} = \frac{12}{45} = \frac{4}{15} \approx \mathbf{0.2667}.$$

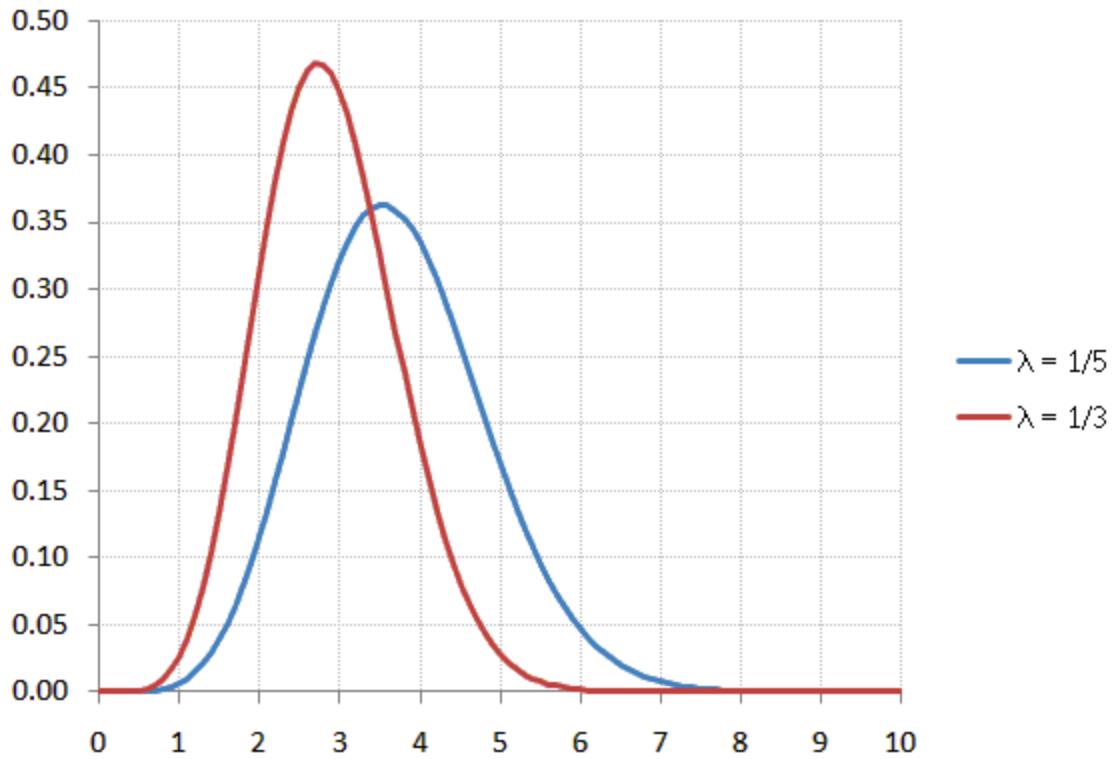
- c) Obtain the maximum likelihood estimator of λ , $\hat{\lambda}$. Suppose $n = 4$, and $x_1 = 4$, $x_2 = 2$, $x_3 = 4$, $x_4 = 3$. Find the maximum likelihood estimate of λ .

$$L(\lambda) = \prod_{i=1}^n \left(\lambda^3 x_i^5 e^{-\lambda x_i^2} \right). \quad \ln L(\lambda) = 3n \cdot \ln \lambda + \sum_{i=1}^n \ln(x_i^5) - \lambda \cdot \sum_{i=1}^n x_i^2.$$

$$(\ln L(\lambda))' = \frac{3n}{\lambda} - \sum_{i=1}^n x_i^2 = 0. \quad \Rightarrow \quad \hat{\lambda} = \frac{3n}{\sum_{i=1}^n X_i^2}.$$

$$x_1 = 4, \quad x_2 = 2, \quad x_3 = 4, \quad x_4 = 3. \quad \sum_{i=1}^n x_i^2 = 45.$$

$$\hat{\lambda} = \frac{3n}{\sum_{i=1}^n x_i^2} = \frac{12}{45} = \frac{4}{15} \approx \mathbf{0.2667}.$$



Useful facts: **Def** $\Gamma(x) = \int_0^{\infty} u^{x-1} e^{-u} du, \quad x > 0.$

$$\Gamma(1) = 1.$$

$$\Gamma(x) = (x-1)\Gamma(x-1).$$

$$\Rightarrow \Gamma(n) = (n-1)! \quad \text{if } n \text{ is an integer.}$$

$$\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}.$$