

1. Let X_1, X_2, \dots, X_n be a random sample of size n from the distribution with probability density function

$$f_X(x) = f_X(x; \theta) = (\theta - 1)^2 \cdot \frac{\ln x}{x^\theta}, \quad x > 1, \quad \theta > 1.$$

Recall: If $\theta > 2$, the method of moments estimator of θ is $\tilde{\theta} = \frac{2\sqrt{\bar{X}} - 1}{\sqrt{\bar{X}} - 1}$.

The maximum likelihood estimator of θ is $\hat{\theta} = 1 + \frac{2n}{\sum_{i=1}^n \ln X_i}$.

$Y = \sum_{i=1}^n \ln X_i$ has Gamma($\alpha = 2n$, “usual $\theta = \frac{1}{\theta - 1}$ ”) distribution.

- Is $\hat{\theta}$ a consistent estimator for θ ?
- Show that $\hat{\theta}$ is asymptotically normally distributed (as $n \rightarrow \infty$). Find the parameters.
- Suppose $\theta > 3$. Is $\tilde{\theta}$ a consistent estimator for θ ?
- Suppose $\theta > 3$. Show that $\tilde{\theta}$ is asymptotically normally distributed (as $n \rightarrow \infty$). Find the parameters.
- Suggest a $(1 - \alpha) 100\%$ confidence interval for θ based on $\sum_{i=1}^n \ln X_i$.
- Suppose $n = 5$, and

$$x_1 = 5, \quad x_2 = 1.2, \quad x_3 = 2, \quad x_4 = 12, \quad x_5 = 1.5.$$
 Construct a 95% confidence interval for θ .
- Find the sufficient statistic $Y = u(X_1, X_2, \dots, X_n)$ for θ .

h) Determine the Fisher information $I(\theta)$.

i) Recall: $\hat{\theta} = 1 + \frac{2n-1}{\sum_{i=1}^n \ln X_i}$ is an unbiased estimator for θ .

Is $\hat{\theta}$ an efficient estimator of θ ? If $\hat{\theta}$ is not an efficient estimator of θ , find its efficiency.

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

$$f(x) = 4\theta x^3 e^{-\theta x^4} \quad x > 0 \quad \theta > 0.$$

Recall: $Y = \sum_{i=1}^n X_i^4$ has Gamma($\alpha = n$, “usual $\theta = \frac{1}{\theta}$ ”) distribution.

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

b) Find the sufficient statistic $Y = u(X_1, X_2, \dots, X_n)$ for θ .

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

$$P(X_i = 1) = \frac{\theta}{3 + \theta}, \quad P(X_i = 2) = \frac{2}{3 + \theta}, \quad P(X_i = 3) = \frac{1}{3 + \theta}, \quad \theta > 0.$$

a) Find a sufficient statistic for θ .

b) Obtain the method of moments estimator $\tilde{\theta}$ of θ .

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

4. Let $X_1, X_2, \dots, X_n, \dots$ be independent random variables, each with the probability density function

$$f(x) = \begin{cases} 2(1-x) & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}.$$

For each of the following sequences, state whether the sequence converges in probability or/and in distribution, and identify the limits (limiting distributions).

a) $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i;$

b) $T_n = \frac{1}{n} (X_1^2 + X_2^2 + \dots + X_n^2);$

c) $U_n = \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right);$

d) $V_n = n \left(\bar{X}_n - \frac{1}{3} \right)^2;$

e) $W_n = \sqrt{n} \left(\bar{X}_n^2 - \frac{1}{9} \right).$

Answers:

1. Let X_1, X_2, \dots, X_n be a random sample of size n from the distribution with probability density function

$$f_X(x) = f_X(x; \theta) = (\theta - 1)^2 \cdot \frac{\ln x}{x^\theta}, \quad x > 1, \quad \theta > 1.$$

Recall: If $\theta > 2$, the method of moments estimator of θ is $\tilde{\theta} = \frac{2\sqrt{\bar{X}} - 1}{\sqrt{\bar{X}} - 1}$.

The maximum likelihood estimator of θ is $\hat{\theta} = 1 + \frac{2n}{\sum_{i=1}^n \ln X_i}$.

$Y = \sum_{i=1}^n \ln X_i$ has Gamma($\alpha = 2n$, “usual $\theta = \frac{1}{\theta - 1}$ ”) distribution.

- a) Is $\hat{\theta}$ a consistent estimator for θ ?

By WLLN, $\frac{1}{n} \sum_{i=1}^n \ln X_i \xrightarrow{P} E(\ln X) = \frac{2}{(\theta - 1)}$.

$g(x) = 1 + \frac{2}{x}$ is continuous at $\frac{2}{(\theta - 1)}$.

$$g\left(\frac{1}{n} \sum_{i=1}^n \ln X_i\right) = \hat{\theta}. \quad g\left(\frac{2}{(\theta - 1)}\right) = \theta. \quad \Rightarrow \quad \hat{\theta} \xrightarrow{P} \theta.$$

- b) Show that $\hat{\theta}$ is asymptotically normally distributed (as $n \rightarrow \infty$). Find the parameters.

$$\text{Var}(\ln X) = \frac{2}{(\theta - 1)^2}.$$

By CLT, $\sqrt{n} \left(\frac{1}{n} \sum_{i=1}^n \ln X_i - \frac{2}{(\theta-1)} \right) \xrightarrow{D} N(0, \frac{2}{(\theta-1)^2})$.

$$g(x) = 1 + \frac{2}{x}, \quad g'(x) = -\frac{2}{x^2}.$$

$$g\left(\frac{1}{n} \sum_{i=1}^n \ln X_i\right) = \hat{\theta}. \quad g\left(\frac{2}{(\theta-1)}\right) = \theta. \quad g'\left(\frac{2}{(\theta-1)}\right) = -\frac{(\theta-1)^2}{2}.$$

$$\sqrt{n} \left(g\left(\frac{1}{n} \sum_{i=1}^n \ln X_i\right) - g\left(\frac{2}{(\theta-1)}\right) \right) = \sqrt{n} (\tilde{\theta} - \theta) \xrightarrow{D} N(0, \frac{(\theta-1)^2}{2}).$$

For large n , $\tilde{\theta} \sim N(\theta, \frac{(\theta-1)^2}{2n})$.

c) Suppose $\theta > 3$. Is $\tilde{\theta}$ a consistent estimator for θ ?

By WLLN, $\bar{X} \xrightarrow{P} \mu = E(X) = \frac{(\theta-1)^2}{(\theta-2)^2}$.

$$g(x) = \frac{2\sqrt{x}-1}{\sqrt{x}-1} \text{ is continuous at } \frac{(\theta-1)^2}{(\theta-2)^2}.$$

$$g(\bar{X}) = \tilde{\theta}. \quad g\left(\frac{(\theta-1)^2}{(\theta-2)^2}\right) = \theta. \quad \Rightarrow \quad \tilde{\theta} \xrightarrow{P} \theta.$$

d) Suppose $\theta > 3$. Show that $\tilde{\theta}$ is asymptotically normally distributed (as $n \rightarrow \infty$). Find the parameters.

By CLT, $\sqrt{n} (\bar{X} - \mu) \xrightarrow{D} N(0, \sigma^2)$.

$$\sigma^2 = \frac{(\theta-1)^2}{(\theta-3)^2} - \left[\frac{(\theta-1)^2}{(\theta-2)^2} \right]^2 = \frac{(\theta-1)^2 (2\theta^2 - 8\theta + 7)}{(\theta-2)^4 (\theta-3)^2}.$$

$$g(x) = \frac{2\sqrt{x}-1}{\sqrt{x}-1}. \quad g'(x) = \frac{1}{2\sqrt{x}(\sqrt{x}-1)^2}.$$

$$g(\bar{X}) = \tilde{\theta} \quad g\left(\frac{(\theta-1)^2}{(\theta-2)^2}\right) = \theta. \quad g'\left(\frac{(\theta-1)^2}{(\theta-2)^2}\right) = \frac{(\theta-2)^3}{2(\theta-1)}.$$

$$\sqrt{n} \left(g(\bar{X}) - g(\mu) \right) = \sqrt{n} (\tilde{\theta} - \theta) \xrightarrow{D} N\left(0, \frac{(\theta-2)^2(2\theta^2-8\theta+7)}{4(\theta-3)^2}\right).$$

$$\text{For large } n, \quad \tilde{\theta} \sim N\left(\theta, \frac{(\theta-2)^2(2\theta^2-8\theta+7)}{4n(\theta-3)^2}\right).$$

e) Suggest a $(1 - \alpha) 100\%$ confidence interval for θ based on $\sum_{i=1}^n \ln X_i$.

$\sum_{i=1}^n \ln X_i$ has Gamma($\alpha = 2n$, “usual $\theta = \frac{1}{\theta-1}$ ”) distribution.

Then $2(\theta-1) \sum_{i=1}^n \ln X_i$ has a $\chi^2(2\alpha = 4n)$ distribution.

$$\Rightarrow P\left(\chi_{1-\alpha/2}^2(4n) < 2(\theta-1) \sum_{i=1}^n \ln X_i < \chi_{\alpha/2}^2(4n)\right) = 1 - \alpha.$$

$$\Rightarrow P\left(1 + \frac{\chi_{1-\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln X_i} < \theta < 1 + \frac{\chi_{\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln X_i}\right) = 1 - \alpha.$$

A $(1 - \alpha) 100\%$ confidence interval for θ

$$\left(1 + \frac{\chi_{1-\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln x_i}, 1 + \frac{\chi_{\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln x_i} \right).$$

f) Suppose $n = 5$, and

$$x_1 = 5, \quad x_2 = 1.2, \quad x_3 = 2, \quad x_4 = 12, \quad x_5 = 1.5.$$

Construct a 95% confidence interval for θ .

$$\sum_{i=1}^5 \ln x_i = \ln 216 \approx 5.3753. \quad \chi_{0.975}^2(20) = 9.591. \quad \chi_{0.025}^2(20) = 34.17.$$

A $(1 - \alpha) 100\%$ confidence interval for θ

$$\left(1 + \frac{\chi_{1-\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln x_i}, 1 + \frac{\chi_{\alpha/2}^2(4n)}{2 \sum_{i=1}^n \ln x_i} \right) = \left(1 + \frac{9.591}{2 \cdot 5.3753}, 1 + \frac{34.17}{2 \cdot 5.3753} \right) \\ = (1.89, 4.18)$$

g) Find the sufficient statistic $Y = u(X_1, X_2, \dots, X_n)$ for θ .

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \prod_{i=1}^n (\theta - 1)^2 \cdot \frac{\ln x_i}{x_i^\theta} \\ = (\theta - 1)^{2n} \cdot \left(\prod_{i=1}^n x_i \right)^{-\theta} \cdot \prod_{i=1}^n \ln x_i.$$

$$\Rightarrow Y_1 = \prod_{i=1}^n X_i \text{ is a sufficient statistic for } \theta.$$

$$\Rightarrow Y_2 = \ln \prod_{i=1}^n X_i = \sum_{i=1}^n \ln X_i \text{ is also a sufficient statistic for } \theta.$$

OR

$$f_X(x; \theta) = (\theta - 1)^2 \cdot \frac{\ln x}{x^\theta} = \exp \{ -\theta \ln x + \ln \ln x + 2 \ln(\theta - 1) \}$$

$$\Rightarrow K(x) = \ln x.$$

$$\Rightarrow Y_2 = \sum_{i=1}^n \ln X_i \text{ is a sufficient statistic for } \theta.$$

$$\Rightarrow Y_1 = e^{Y_2} = \prod_{i=1}^n X_i \text{ is also a sufficient statistic for } \theta.$$

h) Determine the Fisher information $I(\theta)$.

$$\ln f(x; \theta) = 2 \cdot \ln(\theta - 1) + \ln \ln x - \theta \cdot \ln x$$

$$\frac{\partial}{\partial \theta} \ln f(x; \theta) = \frac{2}{\theta - 1} - \ln x$$

$$\frac{\partial^2}{\partial \theta^2} \ln f(x; \theta) = -\frac{2}{(\theta - 1)^2}$$

$$I(\theta) = -E \left[\frac{\partial^2}{\partial \theta^2} \ln f(X; \theta) \right] = \frac{2}{(\theta - 1)^2}$$

i) Recall: $\hat{\theta} = 1 + \frac{2n-1}{\sum_{i=1}^n \ln X_i}$ is an unbiased estimator for θ .

Is $\hat{\theta}$ an efficient estimator of θ ? If $\hat{\theta}$ is not an efficient estimator of θ , find its efficiency.

$$\text{Recall: } \text{Var}(\hat{\theta}) = \frac{(2n-1)^2}{(2n)^2} \text{Var}(\hat{\theta}) = \frac{(\theta-1)^2}{(2n-2)}.$$

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

$\Rightarrow \hat{\theta}$ is NOT an efficient estimator of θ ,

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

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

$$f(x) = 4\theta x^3 e^{-\theta x^4} \quad x > 0 \quad \theta > 0.$$

Recall: $Y = \sum_{i=1}^n X_i^4$ has Gamma($\alpha = n$, “usual $\theta = \frac{1}{\theta}$ ”) distribution.

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

$2Y/\text{“usual } \theta\text{”} = 2\theta \sum_{i=1}^n X_i^4$ has a chi-square distribution with $r = 2\alpha = 2n$ d.f.

$$\Rightarrow P(\chi_{1-\alpha/2}^2(2n) < 2\theta \sum_{i=1}^n X_i^4 < \chi_{\alpha/2}^2(2n)) = 1 - \alpha.$$

$$\Rightarrow P\left(\frac{\chi_{1-\alpha/2}^2(2n)}{2 \sum_{i=1}^n X_i^4} < \theta < \frac{\chi_{\alpha/2}^2(2n)}{2 \sum_{i=1}^n X_i^4}\right) = 1 - \alpha.$$

A $(1 - \alpha) 100\%$ confidence interval for θ :

$$\left(\frac{\chi_{1-\alpha/2}^2(2n)}{2 \sum_{i=1}^n X_i^4}, \frac{\chi_{\alpha/2}^2(2n)}{2 \sum_{i=1}^n X_i^4} \right).$$

- b) Find the sufficient statistic $Y = u(X_1, X_2, \dots, X_n)$ for θ .

$$f(x_1, x_2, \dots, x_n; \theta) = f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta)$$

$$= \left[4^n \theta^n e^{-\theta \sum_{i=1}^n x_i^4} \right] \left(\prod_{i=1}^n x_i^3 \right).$$

By Factorization Theorem, $Y = \sum_{i=1}^n X_i^4$ is a sufficient statistic for θ .

OR

$$f(x; \theta) = \exp\{-\theta \cdot x^4 + \ln \theta + \ln 4 + 3 \ln x\}.$$

$$\Rightarrow K(x) = x^4.$$

$$\Rightarrow Y = \sum_{i=1}^n K(X_i) = \sum_{i=1}^n X_i^4 \text{ is a sufficient statistic for } \theta.$$

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

$$P(X_i = 1) = \frac{\theta}{3+\theta}, \quad P(X_i = 2) = \frac{2}{3+\theta}, \quad P(X_i = 3) = \frac{1}{3+\theta}, \quad \theta > 0.$$

- a) Find a sufficient statistic for θ .

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \frac{1}{(3+\theta)^n} \cdot \theta^{(\# \text{ of 1's})} \cdot 2^{(\# \text{ of 2's})} \cdot 1^{(\# \text{ of 3's})}.$$

$\Rightarrow Y = (\# \text{ of 1's})$ is a sufficient statistic for θ .

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

$$E(X) = 1 \times \frac{\theta}{3+\theta} + 2 \times \frac{2}{3+\theta} + 3 \times \frac{1}{3+\theta} = \frac{\theta+7}{3+\theta}.$$

$$\frac{1}{n} \cdot \sum_{i=1}^n x_i = \bar{x} = \frac{\tilde{\theta}+7}{3+\tilde{\theta}}. \quad 3\bar{x} + \tilde{\theta}\bar{x} = \tilde{\theta}+7.$$

$$\Rightarrow \tilde{\theta} = \frac{7-3\bar{x}}{\bar{x}-1}.$$

- c) Obtain the maximum likelihood estimator $\hat{\theta}$ of θ .

$$L(\theta) = \frac{1}{(3+\theta)^n} \cdot \theta^{(\# \text{ of 1's})} \cdot 2^{(\# \text{ of 2's})} \cdot 1^{(\# \text{ of 3's})}.$$

$$\ln L(\theta) = -n \ln(3+\theta) + (\# \text{ of 1's}) \ln(\theta) + (\# \text{ of 2's}) \ln(2) + (\# \text{ of 3's}) \ln(1).$$

$$(\ln L(\theta))' = -\frac{n}{3+\theta} + \frac{(\# \text{ of 1's})}{\theta} = 0 \quad \Rightarrow \quad \hat{\theta} = \frac{3 \cdot (\# \text{ of 1's})}{n - (\# \text{ of 1's})}.$$

4. Let $X_1, X_2, \dots, X_n, \dots$ be independent random variables, each with the probability density function

$$f(x) = \begin{cases} 2(1-x) & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}.$$

For each of the following sequences, state whether the sequence converges in probability or/and in distribution, and identify the limits (limiting distributions).

a) $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i;$

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i;$$

By WLLN, $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \xrightarrow{P} \mu = E(X) = \frac{1}{3}.$

b) $T_n = \frac{1}{n} (X_1^2 + X_2^2 + \dots + X_n^2);$

$$T_n = \frac{1}{n} (X_1^2 + X_2^2 + \dots + X_n^2);$$

By WLLN, $T_n = \frac{1}{n} (X_1^2 + X_2^2 + \dots + X_n^2) \xrightarrow{P} E(X^2) = \frac{1}{6}.$

c) $U_n = \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right);$

$$U_n = \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right);$$

By CLT, $U_n = \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right) \xrightarrow{D} N(0, \sigma^2) = N(0, \frac{1}{18}).$

$$d) \quad V_n = n \left(\bar{X}_n - \frac{1}{3} \right)^2;$$

$$V_n = n \left(\bar{X}_n - \frac{1}{3} \right)^2;$$

$$\text{From part (c),} \quad \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right) \xrightarrow{D} \frac{1}{\sqrt{18}} N(0, 1).$$

Since $g(x) = x^2$ is continuous,

$$g \left(\sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right) \right) = n \left(\bar{X}_n - \frac{1}{3} \right)^2 \xrightarrow{D} \left[\frac{1}{\sqrt{18}} N(0, 1) \right]^2 = \frac{1}{18} \chi^2(1).$$

$$e) \quad W_n = \sqrt{n} \left(\bar{X}_n^2 - \frac{1}{9} \right).$$

$$W_n = \sqrt{n} \left(\bar{X}_n^2 - \frac{1}{9} \right).$$

$$\text{From part (c),} \quad \sqrt{n} \left(\bar{X}_n - \frac{1}{3} \right) \xrightarrow{D} N\left(0, \frac{1}{18}\right).$$

Since $g(x) = x^2$ is differentiable and $g'(\frac{1}{3}) = \frac{2}{3} \neq 0$,

$$\sqrt{n} \left(g(\bar{X}_n) - g\left(\frac{1}{3}\right) \right) = \sqrt{n} \left(\bar{X}_n^2 - \frac{1}{9} \right) \xrightarrow{D} N\left(0, \left(\frac{2}{3}\right)^2 \times \frac{1}{18}\right) = N\left(0, \frac{2}{81}\right).$$