

SOLUTIONS

The following are a number of practice problems that may be *helpful* for completing the homework, and will likely be **very useful** for studying for exams.

1. (Capture – Recapture) To estimate the populations size of unicorns in Neverland, first N_1 unicorns were captured and tagged. The captured unicorns were then released. One month later, n unicorns were captured. Let X denote the number of tagged unicorns among the ones in the second sample.

a) Construct an estimator for the population size N . Hint: Ask MoM.

X has Hypergeometric distribution. $E(X) = n \cdot \frac{N_1}{N}$.

$$X = n \cdot \frac{N_1}{\tilde{N}} \Rightarrow \tilde{N} = \frac{n \cdot N_1}{X}$$

b) Suppose $N_1 = 12$, $n = 10$, and $x = 3$. Obtain \tilde{N} , an estimate for the population size N .

$$\tilde{N} = \frac{n \cdot N_1}{x} = \frac{10 \cdot 12}{3} = \mathbf{40}.$$

c) Suppose $N = 33$, $N_1 = 12$, and $n = 10$. Find the probability that \tilde{N} is within 10 of N .
That is, find the probability $P(23 \leq \tilde{N} \leq 43)$.

x	0	1	2	3	4	5	6	7	8	9	10
\tilde{N}	∞	120	60	40	30	24	20	~ 17	15	~ 13	12

$$P(23 \leq \tilde{N} \leq 43) = P(3 \leq X \leq 5) = \frac{\binom{12}{3} \binom{21}{7}}{\binom{33}{10}} + \frac{\binom{12}{4} \binom{21}{6}}{\binom{33}{10}} + \frac{\binom{12}{5} \binom{21}{5}}{\binom{33}{10}}$$

$$\approx 0.2764 + 0.2902 + 0.1741 = \mathbf{0.7407}.$$

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

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

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

$$L(\lambda) = \prod_{i=1}^n \left(\frac{2\sqrt{\lambda}}{\sqrt{\pi}} e^{-\lambda x_i^2} \right).$$

$$\ln L(\lambda) = n \cdot \ln 2 + \frac{n}{2} \cdot \ln \lambda - \frac{n}{2} \cdot \ln \pi - \lambda \cdot \sum_{i=1}^n x_i^2.$$

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

- d) Suppose $n = 4$, and $x_1 = 0.2$, $x_2 = 0.6$, $x_3 = 1.1$, $x_4 = 1.7$.
Find the maximum likelihood estimate of λ .

$$x_1 = 0.2, \quad x_2 = 0.6, \quad x_3 = 1.1, \quad x_4 = 1.7.$$

$$\sum_{i=1}^n x_i^2 = 4.5. \quad \hat{\lambda} = \frac{4}{9} \approx 0.444.$$

c) Obtain the method of moments estimator of λ , $\tilde{\lambda}$.

$$E(X) = \int_0^{\infty} x \cdot \frac{2\sqrt{\lambda}}{\sqrt{\pi}} e^{-\lambda x^2} dx \quad u = \lambda x^2 \quad du = 2\lambda x dx$$

$$= \int_0^{\infty} \frac{1}{\sqrt{\pi\lambda}} e^{-u} du = \frac{1}{\sqrt{\pi\lambda}}.$$

$$\bar{X} = \frac{1}{\sqrt{\pi\lambda}} \quad \Rightarrow \quad \tilde{\lambda} = \frac{1}{\pi(\bar{X})^2}.$$

d) Suppose $n = 4$, and $x_1 = 0.2$, $x_2 = 0.6$, $x_3 = 1.1$, $x_4 = 1.7$.
Find a method of moments estimate of λ .

$$x_1 = 0.2, \quad x_2 = 0.6, \quad x_3 = 1.1, \quad x_4 = 1.7.$$

$$\bar{x} = 0.9. \quad \tilde{\lambda} \approx 0.393.$$

3. 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

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

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

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

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

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

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

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 .

- c) 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 .

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

$$\bar{x} = 0.6.$$

$$\tilde{\theta} = \mathbf{3}.$$

4. Let $\theta > 0$ and let X_1, X_2, \dots, X_n be a random sample from a Uniform distribution on interval $(0, \theta)$.

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

$$E(X) = \frac{\theta}{2}. \quad \Rightarrow \quad \bar{X} = \frac{\tilde{\theta}}{2}. \quad \Rightarrow \quad \tilde{\theta} = 2\bar{X}.$$

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

$$E(\bar{X}) = E(X) = \frac{\theta}{2}. \quad \Rightarrow \quad E(\tilde{\theta}) = E(2\bar{X}) = \theta. \quad \checkmark$$

$\tilde{\theta}$ is unbiased for θ .

c) Find $\text{Var}(\tilde{\theta})$.

$$\tilde{\theta} = 2\bar{X}. \quad \text{Var}(\tilde{\theta}) = \text{Var}(2\bar{X}) = 4 \text{Var}(\bar{X}) = 4 \cdot \frac{\sigma^2}{n}.$$

$$\text{For Uniform}(0, \theta), \quad \sigma^2 = \frac{\theta^2}{12}. \quad \Rightarrow \quad \text{Var}(\tilde{\theta}) = \frac{\theta^2}{3 \cdot n}.$$

d) Find $\text{MSE}(\tilde{\theta})$.

$$\text{bias}(\tilde{\theta}) = E(\tilde{\theta}) - \theta = 0 \quad \text{and} \quad \text{Var}(\tilde{\theta}) = \frac{\theta^2}{3 \cdot n}.$$

$$\Rightarrow \quad \text{MSE}(\tilde{\theta}) = E[(\tilde{\theta} - \theta)^2] = (\text{bias}(\tilde{\theta}))^2 + \text{Var}(\tilde{\theta}) = 0 + \frac{\theta^2}{3 \cdot n} = \frac{\theta^2}{3 \cdot n}.$$

$$\text{Note that} \quad \text{MSE}(\tilde{\theta}) = \frac{\theta^2}{3 \cdot n} \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty.$$

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

$$f(x; \theta) = \frac{3 + \theta x}{12 + 8\theta}, \quad 0 < x < 4, \quad \theta > -\frac{3}{4}.$$

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

$$\begin{aligned} \mu = E(X) &= \int_0^4 x \cdot \frac{3 + \theta x}{12 + 8\theta} dx = \frac{1}{12 + 8\theta} \cdot \int_0^4 (3x + \theta x^2) dx \\ &= \frac{1}{12 + 8\theta} \cdot \left(\frac{3x^2}{2} + \frac{\theta x^3}{3} \right) \Big|_0^4 = \frac{24 + \frac{64}{3}\theta}{12 + 8\theta} = \frac{72 + 64\theta}{36 + 24\theta} = \frac{18 + 16\theta}{9 + 6\theta}. \end{aligned}$$

$$\bar{X} = \frac{18 + 16\tilde{\theta}}{9 + 6\tilde{\theta}}. \quad 9\bar{X} + 6\bar{X}\tilde{\theta} = 18 + 16\tilde{\theta}.$$

$$\Rightarrow \tilde{\theta} = \frac{12\bar{X} - 24}{64 - 8\bar{X}} = \frac{3\bar{X} - 6}{16 - 2\bar{X}} = \frac{36\bar{X} - 72}{64 - 24\bar{X}} = \frac{9\bar{X} - 18}{16 - 6\bar{X}}.$$

- b) Suppose $n = 5$, and $x_1 = 1.2$, $x_2 = 1.8$, $x_3 = 2.6$, $x_4 = 3.1$, $x_5 = 3.8$. Find the method of moments estimate of θ .

$$x_1 = 1.2, \quad x_2 = 1.8, \quad x_3 = 2.6, \quad x_4 = 3.1, \quad x_5 = 3.8. \quad \bar{x} = 2.5.$$

$$\tilde{\theta} = \mathbf{4.5}.$$

- c) Suppose $n = 4$, and $x_1 = 1.3$, $x_2 = 2.2$, $x_3 = 3.1$, $x_4 = 3.8$. Find the method of moments estimate of θ .

$$x_1 = 1.3, \quad x_2 = 2.2, \quad x_3 = 3.1, \quad x_4 = 3.8. \quad \bar{x} = 2.6.$$

$$\tilde{\theta} = \mathbf{13.5}.$$

6. **6.4-12** **6.1-12** 6.2-12

Let X_1, X_2, \dots, X_n be a random sample from Binomial(1, p) (i.e., n Bernoulli trials).

Thus

$$Y = \sum_{i=1}^n X_i \text{ is Binomial}(n, p).$$

- a) Show that $\bar{X} = \frac{Y}{n}$ is an unbiased estimator of p .

$$E(Y) = np. \quad \Rightarrow \quad E(\bar{X}) = E\left(\frac{Y}{n}\right) = \frac{1}{n} E(Y) = \frac{np}{n} = p.$$

- b) Show that $\text{Var}(\bar{X}) = \frac{p(1-p)}{n}$.

$$\text{Var}(Y) = np(1-p).$$

$$\text{Var}(\bar{X}) = \text{Var}\left(\frac{Y}{n}\right) = \frac{1}{n^2} \text{Var}(Y) = \frac{np(1-p)}{n^2} = \frac{p(1-p)}{n}.$$

- c) Show that $E\left[\frac{\bar{X}(1-\bar{X})}{n}\right] = (n-1)\left[\frac{p(1-p)}{n^2}\right]$.

$$E(Y^2) = \text{Var}(Y) + [E(Y)]^2 = np(1-p) + n^2 p^2.$$

$$E(\bar{X}(1-\bar{X})) = E\left(\frac{Y}{n}\right) - E\left(\frac{Y^2}{n^2}\right) = p - \frac{p(1-p)}{n} - p^2 = (n-1)\left[\frac{p(1-p)}{n}\right].$$

$$\Rightarrow E\left[\frac{\bar{X}(1-\bar{X})}{n}\right] = (n-1)\left[\frac{p(1-p)}{n^2}\right].$$

- d) Find the value of c so that $c \bar{X}(1-\bar{X})$ is an unbiased estimator of $\text{Var}(\bar{X}) = \frac{p(1-p)}{n}$.

$$E(\bar{X}(1-\bar{X})) = (n-1)\left[\frac{p(1-p)}{n}\right]. \quad \Rightarrow \quad E\left(\frac{1}{(n-1)} \bar{X}(1-\bar{X})\right) = \frac{p(1-p)}{n}.$$

$$\Rightarrow c = \frac{1}{(n-1)}.$$

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

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

7. a) Find $E(X^k)$, $k > -4$.

Hint 1: Consider $u = \lambda x^2$ or $u = x^2$.

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

Hint 3: $\Gamma(a) = (a-1)\Gamma(a-1)$.

Hint 4: $\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$.

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

OR

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

since $\frac{1}{\Gamma\left(\frac{k}{2}+2\right)} \lambda^{\frac{k}{2}+2} u^{\frac{k}{2}+1} e^{-\lambda u}$ is the p.d.f. of Gamma $(\alpha = \frac{k}{2}+2, \theta = \frac{1}{\lambda})$.

b) Obtain a method of moments estimator of λ , $\tilde{\lambda}$.

$$\begin{aligned} E(X) &= \lambda^{-1/2} \Gamma\left(\frac{1}{2} + 2\right) = \lambda^{-1/2} \cdot \Gamma\left(\frac{5}{2}\right) = \lambda^{-1/2} \cdot \frac{3}{2} \cdot \Gamma\left(\frac{3}{2}\right) \\ &= \lambda^{-1/2} \cdot \frac{3}{2} \cdot \frac{1}{2} \cdot \Gamma\left(\frac{1}{2}\right) = \lambda^{-1/2} \cdot \frac{3}{2} \cdot \frac{1}{2} \cdot \sqrt{\pi} = \frac{3}{4} \cdot \sqrt{\frac{\pi}{\lambda}}. \end{aligned}$$

$$\frac{3}{4} \cdot \sqrt{\frac{\pi}{\lambda}} = \bar{X} \quad \Rightarrow \quad \tilde{\lambda}_1 = \frac{9\pi}{16(\bar{X})^2}.$$

OR

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

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

c) Suppose $n = 5$, and $x_1 = 0.6$, $x_2 = 1.1$, $x_3 = 2.7$, $x_4 = 3.3$, $x_5 = 4.5$.

Find a method of moments estimate of λ .

$$x_1 = 0.6, \quad x_2 = 1.1, \quad x_3 = 2.7, \quad x_4 = 3.3, \quad x_5 = 4.5. \quad \bar{x} = \frac{12.2}{5} = 2.44.$$

$$\tilde{\lambda}_1 = \frac{9\pi}{16(\bar{x})^2} \approx 0.09448 \pi \approx 0.29682.$$

OR

$$x_1 = 0.6, \quad x_2 = 1.1, \quad x_3 = 2.7, \quad x_4 = 3.3, \quad x_5 = 4.5. \quad \sum_{i=1}^n x_i^2 = 40.$$

$$\tilde{\lambda}_2 = \frac{2n}{\sum_{i=1}^n x_i^2} = 0.25.$$

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

$$L(\lambda) = \prod_{i=1}^n \left(2\lambda^2 x_i^3 e^{-\lambda x_i^2} \right).$$

$$\ln L(\lambda) = n \cdot \ln 2 + 2n \cdot \ln \lambda + \sum_{i=1}^n \ln(x_i^3) - \lambda \cdot \sum_{i=1}^n x_i^2.$$

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

- e) Suppose $n = 5$, and $x_1 = 0.6$, $x_2 = 1.1$, $x_3 = 2.7$, $x_4 = 3.3$, $x_5 = 4.5$.
Find the maximum likelihood estimate of λ .

$$x_1 = 0.6, \quad x_2 = 1.1, \quad x_3 = 2.7, \quad x_4 = 3.3, \quad x_5 = 4.5. \quad \sum_{i=1}^n x_i^2 = 40.$$

$$\hat{\lambda} = \frac{2n}{\sum_{i=1}^n x_i^2} = \mathbf{0.25}.$$

- f) Suppose $n = 5$, and $x_1 = 0.5$, $x_2 = 1.2$, $x_3 = 0.4$, $x_4 = 0.8$, $x_5 = 0.1$.
Find the maximum likelihood estimate of λ .

$$x_1 = 0.5, \quad x_2 = 1.2, \quad x_3 = 0.4, \quad x_4 = 0.8, \quad x_5 = 0.1. \quad \sum_{i=1}^n x_i^2 = 2.5.$$

$$\hat{\lambda} = \frac{2n}{\sum_{i=1}^n x_i^2} = \mathbf{4}.$$

9. A random sample of size $n = 16$ from $N(\mu, \sigma^2 = 64)$ yielded $\bar{x} = 85$.
Construct the following confidence intervals for μ :

$$\bar{x} = 85 \qquad \sigma = 8 \qquad n = 16$$

$$\sigma \text{ is known.} \qquad \text{The confidence interval :} \qquad \bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}.$$

- a) 95%.

$$\alpha = 0.05 \qquad \alpha/2 = 0.025. \qquad z_{\alpha/2} = 1.96.$$

$$85 \pm 1.96 \cdot \frac{8}{\sqrt{16}} \qquad \qquad \qquad \mathbf{85 \pm 3.92} \qquad \qquad \mathbf{(81.08 ; 88.92)}$$

- b) 90%.

$$\alpha = 0.10 \qquad \alpha/2 = 0.05. \qquad z_{\alpha/2} = 1.645.$$

$$85 \pm 1.645 \cdot \frac{8}{\sqrt{16}} \qquad \qquad \qquad \mathbf{85 \pm 3.29} \qquad \qquad \mathbf{(81.71 ; 88.29)}$$

- c) 80%.

$$\alpha = 0.20 \qquad \alpha/2 = 0.10. \qquad z_{\alpha/2} = 1.28.$$

$$85 \pm 1.28 \cdot \frac{8}{\sqrt{16}} \qquad \qquad \qquad \mathbf{85 \pm 2.56} \qquad \qquad \mathbf{(82.44 ; 87.56)}$$

OR

$$\alpha = 0.20 \qquad \alpha/2 = 0.10. \qquad z_{\alpha/2} = 1.282.$$

$$85 \pm 1.282 \cdot \frac{8}{\sqrt{16}} \qquad \qquad \qquad \mathbf{85 \pm 2.564} \qquad \qquad \mathbf{(82.436 ; 87.564)}$$

10. What is the minimum sample size required for estimating μ for $N(\mu, \sigma^2 = 64)$ to within ± 3 with confidence level

$$\varepsilon = 10, \quad \sigma = 8.$$

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{\varepsilon} \right)^2 = \left(\frac{z_{\alpha/2} \cdot 8}{3} \right)^2.$$

a) 95%. $\alpha = 0.05$ $\alpha/2 = 0.025.$ $z_{\alpha/2} = 1.96.$

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{\varepsilon} \right)^2 = \left(\frac{1.96 \cdot 8}{3} \right)^2 \approx 27.318. \quad \text{Round up.} \quad n = \mathbf{28}.$$

b) 90%. $\alpha = 0.10$ $\alpha/2 = 0.05.$ $z_{\alpha/2} = 1.645.$

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{\varepsilon} \right)^2 = \left(\frac{1.645 \cdot 8}{3} \right)^2 \approx 19.243. \quad \text{Round up.} \quad n = \mathbf{20}.$$

c) 80%. $\alpha = 0.20$ $\alpha/2 = 0.10.$ $z_{\alpha/2} = 1.28.$

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{\varepsilon} \right)^2 = \left(\frac{1.28 \cdot 8}{3} \right)^2 \approx 11.651. \quad \text{Round up.} \quad n = \mathbf{12}.$$

OR

$$\alpha = 0.20 \quad \alpha/2 = 0.10. \quad z_{\alpha/2} = 1.282.$$

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{\varepsilon} \right)^2 = \left(\frac{1.282 \cdot 8}{3} \right)^2 \approx 11.687. \quad \text{Round up.} \quad n = \mathbf{12}.$$