

STAT/MATH 415 HW#3

September 22, 2017

EXERCISES

6.4.3 A random sample X_1, X_2, \dots, X_n of size n is taken from a Poisson distribution with a mean of λ , $0 < \lambda < \infty$.

a) Show that the maximum likelihood estimator for λ is $\hat{\lambda} = \bar{X}$.

Answer: Since $X_i \sim \text{Pois}(\lambda)$, the pdf of X is

$$f(x_i; \lambda) = \frac{\lambda^{x_i} e^{-\lambda}}{x_i!}$$

$$L(\lambda) = \prod_{i=1}^n f(x_i; \lambda) = \frac{\lambda^{x_1+x_2+\dots+x_n} e^{-\lambda n}}{x_1! x_2! \dots x_n!} = \frac{\lambda^{n\bar{x}} e^{-\lambda n}}{x_1! x_2! \dots x_n!} \quad \bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$$

Since $L(\lambda)$ is a continuous function of λ , thus $\hat{\lambda}$ can be found at either derivative = 0 or the boundary of parameter space.

$$\frac{\partial L}{\partial \lambda} = \frac{1}{x_1! x_2! \dots x_n!} [(n\bar{x}) \lambda^{n\bar{x}-1} e^{-\lambda n} + \lambda^{n\bar{x}} e^{-\lambda n} (-n)]$$

$$= \frac{1}{x_1! x_2! \dots x_n!} (n\bar{x} - n\lambda) \lambda^{n\bar{x}-1} e^{-\lambda n} = 0 \quad \Rightarrow \quad \bar{x} = \lambda$$

$$\lim_{\lambda \rightarrow 0} L(\lambda) = 0 \quad (\text{check the boundary})$$

Thus, the maximum likelihood estimator is $\hat{\lambda} = \bar{X}$

b) Let X equal the number of flaws per 100 feet of a used computer tape. Assume that X has a Poisson distribution with a mean of λ . If 40 observations of X yielded 5 zeros, 7 ones, 12 twos, 9 threes, 5 fours, 1 five and 1 six. Find the maximum likelihood estimate of λ .

Answer: Since $X \sim \text{Pois}(\lambda)$, $f(x; \lambda) = \frac{\lambda^x e^{-\lambda}}{x!}$, $x = 0, 1, 2, \dots$

$$L(\lambda) = (f(0, \lambda))^5 \times (f(1, \lambda))^7 \times (f(2, \lambda))^{12} \times (f(3, \lambda))^9 \times (f(4, \lambda))^5 \times (f(5, \lambda))^1 \times (f(6, \lambda))^1$$

$$= e^{-5\lambda} \times (\lambda^7 e^{-7\lambda}) \times \left(\frac{\lambda^{24} e^{-12\lambda}}{2^{12}}\right) \times \left(\frac{\lambda^{27} e^{-9\lambda}}{6^9}\right) \times \left(\frac{\lambda^{20} e^{-5\lambda}}{24^5}\right) \times \left(\frac{\lambda^5 e^{-\lambda}}{120}\right) \times \left(\frac{\lambda^6 e^{-\lambda}}{720}\right)$$

$$= \frac{\lambda^{89} e^{-40\lambda}}{2^{12} \times 6^9 \times 24^5 \times 120 \times 720}$$

$$\log L(\lambda) = 89 \log \lambda - 40\lambda - \log(2^{12} \times 6^9 \times 24^5 \times 120 \times 720)$$

$$\frac{\partial L}{\partial \lambda} = \frac{89}{\lambda} - 40 = 0 \quad \Rightarrow \quad \hat{\lambda} = 2.225$$

$$\lim_{\lambda \rightarrow 0} \log L(\lambda) = -\infty \quad (\text{check the boundary})$$

Thus, the maximum likelihood estimator is $\hat{\lambda} = 2.225$

6.4.9 Let X_1, X_2, \dots, X_n be a random sample of size n from the exponential distribution whose pdf is $f(x; \theta) = (1/\theta)e^{-x/\theta}$, $0 < x < \infty$, $0 < \theta < \infty$

a) Show that \bar{X} is an unbiased estimator of θ

Answer: We can assume that X_1, \dots, X_n are independent

$$E(X) = \int_0^{\infty} xf(x)dx = \frac{1}{\theta} \int_0^{\infty} xe^{-x/\theta} dx = \frac{1}{\theta} \theta^2 = \theta$$

$$E(\bar{X}) = E\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) = \frac{1}{n} E(X_1 + X_2 + \dots + X_n) = \frac{E(X_1) + E(X_2) + \dots + E(X_n)}{n} = \theta$$

Thus \bar{X} is an unbiased estimator of θ

b) Show that the variance of \bar{X} is θ^2/n

Answer: Similarly, we can have

$$Var(\bar{X}) = Var\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) = \frac{1}{n^2} Var(X_1 + X_2 + \dots + X_n) = \frac{Var(X_1) + \dots + Var(X_n)}{n^2}$$

$$Var(X) = E(X^2) - E(X)^2 = \int_0^{\infty} x^2 f(x) dx - \theta^2 = 2\theta^2 - \theta^2 = \theta^2$$

$$Var(\bar{X}) = \frac{n\theta^2}{n^2} = \frac{\theta^2}{n}$$

c) What is a good estimator of θ if a random sample of size 5 yielded the sample values 3.5, 8.1, 0.9, 4.4 and 0.5?

Answer: From a and b we can know that \bar{X} is an unbiased estimate of θ and its variance $\rightarrow 0$ as $n \rightarrow \infty$, so \bar{X} is a good estimator. In this case,

$$\bar{X} = \frac{3.5 + 8.1 + 0.9 + 4.4 + 0.5}{5} = 3.48$$

6.4.13 Let X_1, X_2, \dots, X_n be a random sample from a uniform distribution on the interval $(\theta - 1, \theta + 1)$

a) Find the method-of-moments estimator of θ

Answer: Since $X_i \sim U(\theta - 1, \theta + 1)$, the pdf is $f(x) = \frac{1}{2}$, $\theta - 1 < x < \theta + 1$

$$E(X) = \int_{\theta-1}^{\theta+1} xf(x)dx = \frac{1}{4}[(\theta + 1)^2 - (\theta - 1)^2] = \theta$$

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n} = E(X) \quad \Rightarrow \quad \hat{\theta} = \frac{X_1 + X_2 + \dots + X_n}{n}$$

b) Is your estimator in part a an unbiased estimator of θ ?

Answer: Assume that X_i s are independent. Here shows \bar{X} is an unbiased estimator of θ

$$E(\hat{\theta}) = E\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) = \frac{1}{n} [E(X_1) + E(X_2) + \dots + E(X_n)] = \frac{n\theta}{n} = \theta$$

c) Given the following $n = 5$ observations of X , give a point estimate of θ : 6.61 7.70 6.98 8.36 7.26

Answer: Based on a and b, a point estimate of θ is \bar{X}

$$\hat{\theta} = \bar{X} = \frac{6.61 + 7.70 + 6.98 + 8.36 + 7.26}{5} = 7.382$$

d) The method-of-moments estimator actually has greater variance than the maximum likelihood estimator of θ , namely $[\min(X_i) + \max(X_i)]/2$. Compute the value of the latter estimator for the $n = 5$ observations in c.

Answer: The minimum is 6.61 and the maximum is 8.36, thus the maximum likelihood estimator is

$$\hat{\theta}_{MLE} = \frac{6.61 + 8.36}{2} = 7.485$$

7.1.3 To determine the effect of 100% nitrate on the growth of pea plants, several specimens were planted and then watered with 100% nitrate every day. At the end of two weeks, the plants were measured. Here are data on seven of them: 17.5 14.5 15.2 14.0 17.3 18.0 13.8 Assume that these data are a random sample from normal distribution $N(\mu, \sigma^2)$

a) Find the value of a point estimate of μ

Answer: Based on maximum likelihood estimate, we know that \bar{X} is a good estimator for μ

$$\hat{\mu} = \bar{X} = \frac{17.5 + 14.5 + 15.2 + 14.0 + 17.3 + 18.0 + 13.8}{7} = 15.757$$

b) Find the value of a point estimate of σ

Answer: Similarly, the maximum likelihood estimate of $\hat{\sigma} = \sqrt{2.751} = 1.659$

$$\hat{\sigma}^2 = \frac{\sum_{i=1}^7 (x_i - \bar{x})^2}{7} = 2.751$$

c) Give the endpoints for a 90% confidence interval for μ

Answer: Here $\alpha = 0.1$, $t_{0.05(6)} = 1.943$

$$s^2 = \frac{\sum_{i=1}^7 (x_i - \bar{x})^2}{7 - 1} = 3.210 \quad \Rightarrow \quad s = 1.792$$

$$\bar{X} \pm t_{0.05(6)} \frac{s}{\sqrt{n}} = 15.757 \pm 1.943 \frac{1.792}{\sqrt{7}} = 15.757 \pm 1.316$$

Thus the 90% confidence interval for μ is (14.441, 17.073)

7.1.7 Thirteen tons of cheese, including “22-pound” wheels is stored in some gypsum mines. A random sample of $n = 9$ of these wheels (label weight) yielded the following weights in pounds: 21.50 18.95 18.55 19.40 19.15 22.35 22.90 22.20 23.10 Assuming that the distribution of the weights of the wheels of cheese is $N(\mu, \sigma^2)$, find a 95% confidence interval for μ

Answer: Here $\alpha = 0.05$, $t_{0.025(8)} = 2.306$

$$\bar{X} = \frac{\sum x_i}{9} = 20.9$$

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{9 - 1} = 3.454$$

$$\bar{X} \pm t_{0.025(8)} \frac{s}{\sqrt{n}} = 20.9 \pm 2.306 \frac{1.858}{3} = 20.9 \pm 1.429$$

Thus the 95% confidence interval for μ is (19.471, 22.329)

7.1.11 Students took $n = 35$ samples of water from the east basin of Lake Macatawa and measured the amount of sodium in parts per million. For their data, they calculated $\bar{x} = 24.11$ and $s^2 = 24.44$. Find an approximate 90% confidence interval for μ , the mean of the amount of sodium in parts per million.

Answer: Here $\alpha = 0.1$, $t_{0.05(34)} \approx 1.645$

$$\bar{X} \pm t_{0.05(34)} \frac{s}{\sqrt{n}} = 24.11 \pm 1.645 \frac{4.944}{5.916} = 24.11 \pm 1.37 = [22.74, 25.48]$$

Thus the 90% confidence interval for μ is (22.74, 25.48)