STAT/MATH 415 HW#3

September 22, 2017

EXERCISES

6.4.3 A random sample $X_1, X_2, ..., X_n$ 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 $X_i \sim Pois(\lambda)$, the pdf of X is

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

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

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.

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

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

 $\lim_{\lambda \to 0} L(\lambda) = 0 \qquad \text{(check the boundary)}$

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
$$X \sim Pois(\lambda)$$
, $f(x; \lambda) = \frac{\lambda^x e^{-\lambda}}{x!}$, $x = 0, 1, 2...$

$$\begin{split} L(\lambda) &= (f(0,\lambda))^5 \times (f(1,\lambda))^7 \times (f(2,\lambda))^{12} \times (f(3,\lambda))^9 \times (f(4,\lambda))^5 \times (f(5,\lambda))^1 \times (f(6,\lambda))^1 \\ &= e^{-5\lambda} \times (\lambda^7 e^{-7\lambda}) \times (\frac{\lambda^{24} e^{-12\lambda}}{2^{12}}) \times (\frac{\lambda^{27} e^{-9\lambda}}{6^9}) \times (\frac{\lambda^{20} e^{-5\lambda}}{24^5}) \times (\frac{\lambda^5 e^{-\lambda}}{120}) \times (\frac{\lambda^6 e^{-\lambda}}{720}) \\ &= \frac{\lambda^{89} e^{-40\lambda}}{2^{12} \times 6^9 \times 24^5 \times 120 \times 720} \end{split}$$

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

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

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

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

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

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

Answer: We can assume that $X_1, ..., X_n$ are independent

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

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

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

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

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

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

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 $\to 0$ as $n \to \infty$, 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 $X_1, X_2, ..., X_n$ 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 $f(x) = \frac{1}{2}$, $\theta - 1 < x < \theta + 1$

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

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

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

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

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

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)