

Statistics for Data Science-2

Week 8 Solve with us

Table of contents

Week 8 Solve with us

Question 1

Question 2

Question 3

Question 4

Question 5

Question 6

Question 7

1. Let X_1, X_2, \dots, X_n be i.i.d. samples from a distribution X with mean μ and standard deviation σ . Let $\hat{\mu} = 20 \left(\frac{X_1 + X_2 + \dots + X_n}{n} \right)$ be an estimator of μ . Find the risk of $\hat{\mu}$.

(a) $\frac{400\sigma^2}{n} + 400\mu^2$

(b) $\frac{20\sigma^2}{n} + 19\mu^2$

(c) $\frac{400\sigma^2}{n} + 361\mu^2$

(d) $\frac{20\sigma^2}{n} + 20\mu$

Solution:

$$\begin{aligned} E[\hat{\mu}] &= E \left[20 \left(\frac{X_1 + X_2 + \dots + X_n}{n} \right) \right] \\ &= \frac{20}{n} (n\mu) \\ &= 20\mu \end{aligned}$$

And

$$\text{Bias}(\hat{\mu}, \mu) = E[\hat{\mu}] - \mu = 20\mu - \mu = 19\mu$$

$$\begin{aligned}\text{Var}(\hat{\mu}) &= \text{Var} \left[20 \left(\frac{X_1 + X_2 + \dots + X_n}{n} \right) \right] \\ &= \frac{400}{n^2} (n\sigma^2) \\ &= \frac{400\sigma^2}{n}\end{aligned}$$

$$\begin{aligned}\text{Risk}(\hat{\mu}) &= \text{Bias}(\hat{\mu}, \mu)^2 + \text{Var}(\hat{\mu}) \\ &= (19\mu)^2 + \frac{400\sigma^2}{n} \\ &= 361\mu^2 + \frac{400\sigma^2}{n}\end{aligned}$$

2. Let $X_1, X_2, \dots, X_n \sim \text{iid } X$, where X is a random variable with density function

$$f_X(x) = \begin{cases} \frac{\theta}{x^{\theta+1}} & x > 1 \\ 0 & \text{otherwise} \end{cases}$$

The sample from this distribution is taken:

3, 6, 2, 7, 8, 10.

What is the maximum likelihood estimate of θ for the sample.

- a) 1.16
- b) 1.39
- c) 3.85
- d) 0.72

Solution:

$$\begin{aligned} L(x_1, x_2, \dots, x_n) &= \prod_{i=1}^n f_X(x_i) \\ &= \prod_{i=1}^n \frac{\theta}{x_i^{\theta+1}} \\ &= \theta^n \left(\frac{1}{x_1^{\theta+1}} \frac{1}{x_2^{\theta+1}} \cdots \frac{1}{x_n^{\theta+1}} \right) \end{aligned}$$

$$\begin{aligned} \Rightarrow \log(L(x_1, x_2, \dots, x_n)) &= \\ n \log(\theta) - (\theta + 1)[\log(x_1) + \log(x_2) + \dots + \log(x_n)] \end{aligned}$$

Therefore, ML estimator for θ is given by

$$\theta^* = \arg \max_{\theta} [n \log(\theta) - (\theta + 1) [\log(x_1) + \log(x_2) + \dots + \log(x_n)]]$$

$$\text{Let } Y = n \log(\theta) - (\theta + 1) (\log(x_1) + \log(x_2) + \dots + \log(x_n))$$

$$\Rightarrow \frac{dY}{d\theta} = \frac{n}{\theta} - (\log(x_1) + \log(x_2) + \dots + \log(x_n))$$

Now we will equate this value to zero and find the value of θ .

$$\begin{aligned} \frac{n}{\theta} - (\log(x_1) + \log(x_2) + \dots + \log(x_n)) &= 0 \\ \Rightarrow \theta &= \frac{n}{\log(x_1) + \log(x_2) + \dots + \log(x_n)} \end{aligned}$$

$$\text{This implies that } \theta^* = \frac{n}{\log(x_1) + \log(x_2) + \dots + \log(x_n)}$$

Therefore the ML estimator of θ for the given sample 3, 6, 2, 7, 8, 10 will be

$$\begin{aligned}\theta^* &= \frac{6}{\log(3) + \log(6) + \log(2) + \log(7) + \log(8) + \log(10)} \\ &= \frac{6}{\log(20160)} \\ &= 1.39\end{aligned}$$

3. Let $X_1, X_2, X_3, X_4 \sim \text{iid Binomial}(4, p)$. Given a random sample $(2, 0, 4, 3)$, find the maximum likelihood estimate of p .

- a) $\frac{3}{4}$
- b) $\frac{9}{16}$
- c) $\frac{1}{2}$
- d) $\frac{1}{4}$

Solution:

$$X_i \sim \text{Binomial}(4, p)$$

$$\Rightarrow f_{X_i}(x) = {}^4C_x p^x (1-p)^{4-x}$$

Likelihood function is given by

$$L(x_1, x_2, x_3, x_4) = \prod_{i=1}^4 f_{X_i}(x_i)$$

$$\Rightarrow L(x_1, x_2, x_3, x_4) = {}^4C_{x_1} p^{x_1} (1-p)^{4-x_1} \times {}^4C_{x_2} p^{x_2} (1-p)^{4-x_2} \times {}^4C_{x_3} p^{x_3} (1-p)^{4-x_3} \times {}^4C_{x_4} p^{x_4} (1-p)^{4-x_4}$$

$$\begin{aligned} L(2, 0, 4, 3) &= {}^4C_2 {}^4C_0 {}^4C_4 {}^3C_3 p^{2+0+4+3} (1-p)^{16-(2+0+4+3)} \\ &= 24p^9(1-p)^7 \end{aligned}$$

$$\Rightarrow \log(L(2, 0, 4, 3)) = \log(24) + 9\log(p) + 7\log(1-p)$$

Therefore, ML estimator for p is given by

$$\hat{p} = \arg \max_p [\log(24) + 9\log(p) + 7\log(1-p)]$$

$$\text{Let } Y = \log(24) + 9\log(p) + 7\log(1-p)$$

$$\Rightarrow \frac{dY}{dp} = \frac{9}{p} - \frac{7}{1-p}$$

Now we will equate this value to zero and find the value of p

$$\frac{9}{p} - \frac{7}{1-p} = 0 \Rightarrow p = \frac{9}{16}$$

$$\Rightarrow \hat{p}_{ML} = \frac{9}{16}$$

4. Suppose that we want to estimate the true average number of eggs a queen bee lays with 95% confidence. The margin of error we are willing to accept is 0.3. Suppose we also know that standard deviation is 9. What sample size should we use?
- a) 3457
 - b) 3458
 - c) 5144
 - d) 5145

Solution:

Let X denote the number of eggs a queen bee lays.

Given that $\sigma = 9$

To find the value of n such that $P(|\hat{\mu} - \mu| \leq 0.3) = 0.95$

$$\begin{aligned} P(|\hat{\mu} - \mu| \leq 0.3) &= 0.95 \\ \Rightarrow P\left(|\frac{\hat{\mu} - \mu}{\sigma/\sqrt{n}}| \leq \frac{0.3}{\sigma/\sqrt{n}}\right) &= 0.95 \\ \Rightarrow P\left(|Z| \leq \frac{0.3}{\sigma/\sqrt{n}}\right) &= 0.95 \end{aligned}$$

$$\begin{aligned}\frac{0.3}{\sigma/\sqrt{n}} &= 1.96 \\ \Rightarrow \sqrt{n} &= 9 \times \frac{1.96}{0.3} \\ \Rightarrow n &= 3457.44\end{aligned}$$

Therefore the sample size should be 3458.

5. Consider a sample of iid random variables X_1, X_2, \dots, X_n , where $n > 30$, $E[X_i] = \mu$, $\text{Var}(X_i) = \sigma^2$ and the estimator of μ ,

$$\hat{\mu}_n = \frac{1}{n-30} \sum_{i=31}^n X_i. \text{ Find the bias of } \hat{\mu}_n.$$

- a) 30μ
- b) 0
- c) $\frac{30\mu}{n-30}$
- d) 31μ

Solution:

$$\begin{aligned} E[\hat{\mu}_n] &= E \left[\frac{1}{n-30} \sum_{i=31}^n X_i \right] \\ &= \frac{(n-30)\mu}{n-30} \\ &= \mu \end{aligned}$$

And

$$\text{Bias}(\hat{\mu}_n, \mu) = E[\hat{\mu}_n] - \mu = \mu - \mu = 0$$

6. Suppose it is known that a sample consisting of the values 15, 16, 10, 8, 7, 9, 20 and 19 comes from a population with the density function

$$f(x) = \begin{cases} \frac{1}{\theta} e^{\frac{-x}{\theta}}, & x > 0, \\ 0, & \text{otherwise.} \end{cases}$$

Find the maximum likelihood estimate of θ .

- a) 0.07
- b) 13
- c) 104
- d) 8

Solution:

$$\begin{aligned} L(x_1, x_2, \dots, x_n) &= \prod_{i=1}^n f_X(x_i) \\ &= \prod_{i=1}^n \frac{1}{\theta} e^{\frac{-x_i}{\theta}} \\ &= \frac{1}{\theta^n} \left(e^{\frac{-x_1}{\theta}} e^{\frac{-x_2}{\theta}} \dots e^{\frac{-x_n}{\theta}} \right) \\ &= \frac{1}{\theta^n} \left(e^{\frac{-(x_1+x_2+\dots+x_n)}{\theta}} \right) \end{aligned}$$

$$\Rightarrow \log(L(x_1, x_2, \dots, x_n)) = -n \log(\theta) - \frac{(x_1+x_2+\dots+x_n)}{\theta}$$

Therefore, ML estimator for θ is given by

$$\hat{\theta} = \arg \max_{\theta} \left[-n \log(\theta) - \frac{(x_1 + x_2 + \dots + x_n)}{\theta} \right]$$

$$\begin{aligned} \text{Let } Y &= -n \log(\theta) - \frac{(x_1 + x_2 + \dots + x_n)}{\theta} \\ \Rightarrow \frac{dY}{d\theta} &= -\frac{n}{\theta} + \frac{(x_1 + x_2 + \dots + x_n)}{\theta^2} \end{aligned}$$

Now we will equate this value to zero and find the value of θ .

$$\begin{aligned} \Rightarrow -\frac{n}{\theta} + \frac{(x_1 + x_2 + \dots + x_n)}{\theta^2} &= 0 \\ \Rightarrow \theta &= \frac{x_1 + x_2 + \dots + x_n}{n} \\ \Rightarrow \hat{\theta} &= \frac{x_1 + x_2 + \dots + x_n}{n} \end{aligned}$$

Therefore, maximum likelihood estimate of θ for the given sample will be

$$\begin{aligned}\hat{\theta} &= \frac{15 + 16 + 10 + 8 + 7 + 9 + 20 + 19}{8} \\ &= \frac{104}{8} \\ &= 13\end{aligned}$$

7. Let X be a discrete random variable with the following probability mass function

x	0	1	2	3
$f_X(x)$	$\frac{1-p}{2}$	$\frac{p}{2}$	$\frac{1-p}{2}$	$\frac{p}{2}$

Table 8.1: PMF of X

Suppose a sample consisting of the values 0, 2, 1, 3, 0, 2, 1 and 3 is taken from the random variable X . Find the estimate of p using method of moments.

- a) 0.40
- b) 0.50
- c) 0.60
- d) 0.75

Solution:

$$\begin{aligned} E[X] &= 0 \times \frac{1-p}{2} + 1 \times \frac{p}{2} + 2 \times \frac{1-p}{2} + 3 \times \frac{p}{2} \\ &= \frac{p + 2(1-p) + 3p}{2} \\ &= p + 1 \end{aligned}$$

Now

$$M_1 = E[X] = p + 1$$

$$\Rightarrow p = M_1 - 1$$

Therefore, estimate of p will be

$$\frac{X_1 + X_2 + \dots + X_n}{n} - 1.$$

So, the estimate of p for the given sample will be

$$\begin{aligned}\hat{p} &= \frac{0 + 2 + 1 + 3 + 0 + 2 + 1 + 3}{8} - 1 \\ &= \frac{12}{8} - 1 \\ &= 0.5\end{aligned}$$