PROBABILITY AND STATISTICS - P&P 4

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Problem 1. Let $X_1,...,X_n \stackrel{iid}{\sim} Geo\left(\theta\right)$, $\theta \in [0,1]$. Find the MLE for θ

Answer:

$$\mathcal{L}(\theta; \boldsymbol{X}) = P(x_1 \cap x_2 \cap \dots \cap x_n; \theta) \stackrel{=}{\underset{iid}{=}} \prod_{i=1}^{n} (1 - \theta)^{x_i - 1} \theta$$

$$\iff$$

$$\ell(\theta; \boldsymbol{X}) = \sum_{i} [(x_i - 1) \log (1 - \theta) + \log (\theta)] = n \cdot \log (\theta) + \log (1 - \theta) \sum_{i} (x_i - 1)$$

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$$\frac{\partial \ell (\theta; \mathbf{X})}{\partial \theta} = \frac{n}{\theta} + \frac{1}{1 - \theta} (-1) \sum_{i} (x_i - 1) = \frac{n}{\theta} - \frac{\sum_{i} x_i - n}{1 - \theta} \underset{FOC}{=} 0$$

$$\iff n (1 - \theta) = \theta \left(\sum_{i} x_i - n \right) \iff \widehat{\theta}_{MLE} = \frac{n}{\sum_{i} x_i} = \frac{1}{\overline{X}}$$

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Problem 2. Let $X_1,...,X_n$ be an i.i.d with $f_{\theta}(x) = \frac{1}{x}e^{-\pi(\log(x)-\theta)^2}$ Find the MLE for θ

Answer:

$$\mathcal{L}(\theta; \boldsymbol{X}) = f_{X}(x_{1} \cap x_{2} \cap \dots \cap x_{n}; \theta) \stackrel{=}{=} \prod_{i \neq i}^{n} \frac{1}{x_{i}} e^{-\pi(\log(x_{i}) - \theta)^{2}}$$

$$\iff$$

$$\ell(\theta; \boldsymbol{X}) = \sum_{i=1}^{n} \left[\log\left(e^{-\pi(\log(x_{i}) - \theta)^{2}}\right) - \log\left(x_{i}\right) \right]$$

$$= \sum_{i=1}^{n} -\pi\left(\log\left(x_{i}\right) - \theta\right)^{2} - \sum_{i=1}^{n} \log\left(x_{i}\right)$$

$$= -\pi \sum_{i=1}^{n} \left(\log^{2}\left(x_{i}\right) - \log\left(x_{i}\right)\theta + \theta^{2}\right) - \sum_{i=1}^{n} \log\left(x_{i}\right)$$

$$= -\pi \left(\sum_{i=1}^{n} \log^{2}\left(x_{i}\right) - 2\sum_{i=1}^{n} \log\left(x_{i}\right)\theta + n\theta^{2}\right) - \sum_{i=1}^{n} \log\left(x_{i}\right)$$

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$$\frac{\partial \ell(\theta; X)}{\partial \theta} = \pi 2 \sum_{i=1}^{n} \log(x_i) - \pi 2n\theta \underset{FOC}{=} 0$$

$$\iff \widehat{\theta}_{MLE} = \frac{1}{n} \sum_{i=1}^{n} \log(x_i)$$

Problem 3. Let $X_1,...,X_n \stackrel{iid}{\sim} N\left(\mu,\sigma^2\right)$, where μ and σ^2 are unknown. Find the MLE for μ and σ^2 .

Answer:

$$\mathcal{L}(\mu, \sigma; \mathbf{X}) = f_X(\mu, \sigma; x_1 \cap x_2 \cap \dots \cap x_n)$$

$$= \prod_{i \neq i \neq j}^n \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x_i - \mu}{\sigma}\right)^2} = n \cdot \left(\sigma \sqrt{2\pi}\right)^{-1} e^{-\frac{1}{2} \sum_{i=1}^n \left(\frac{x_i - \mu}{\sigma}\right)^2}$$

$$\iff$$

$$\ell(\mu, \sigma; \mathbf{X}) = -n \cdot \log\left(\sigma \sqrt{2\pi}\right) - \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2$$

$$\frac{\partial \ell \left(\mu, \sigma; \boldsymbol{X}\right)}{\partial \mu} = -\frac{1}{2\sigma^2} 2 \sum_{i=1}^n \left(x_i - \mu\right) \cdot \left(-1\right) \underset{FOC}{=} 0 \iff n\mu = \sum_{i=1}^n x_i \implies \widehat{\mu}_{MLE} = \overline{X}$$

$$\frac{\partial \ell \left(\mu, \sigma; \boldsymbol{X}\right)}{\partial \sigma} = -n \cdot \frac{1}{\sigma\sqrt{2\pi}} \cdot \sqrt{2\pi} - \frac{(-2)}{2\sigma^3} \sum_{i=1}^n \left(x_i - \mu\right)^2 \underset{FOC}{=} 0$$

$$\implies \widehat{\sigma}_{MLE}^2 = \frac{1}{n} \sum_{i=1}^n \left(x_i - \mu\right)^2 = \frac{1}{n} \sum_{i=1}^n \left(x_i - \overline{X}\right)^2$$

Problem 4. Let $X_1, ..., X_n \stackrel{iid}{\sim} U(\theta + 2, \theta + 10)$ (continuous).

- (1) Find $\hat{\theta}_{MOM}$ (method of moments estimator for θ).
- (2) Evaluate $\hat{\theta}_{MOM}$ for the sample: $\{12.3, 17.5, 15.1, 14.7\}$

Answer: (1)

$$\mu_1(\theta) = EX = \frac{(\theta+2) + (\theta+10)}{2} = \theta + 6 = m_1 = \overline{X}$$

$$\iff \hat{\theta}_{MOM} = \overline{X} - 6$$

(2)

$$\hat{\theta}_{MOM} = \overline{X} - 6 = \left(\frac{12.3 + 17.5 + 15.1 + 14.7}{4} - 6\right) = 8.9$$

Problem 5. It is assumed that the daily amount of rain (in mm) that falls in London during January is distributed $N(\mu, 25)$. We are interested in estimating P(X > 75). Two approaches were suggested:

A. Estimate μ using the method of moments, and then estimate the probability using $\hat{\mu}_{MOM}$ instead of μ in the normal distribution.

B. Don't assume normality. Estimate the probability by calculating the proportion of observations that are greater than 75.

In a random sample of 10 observations, the following results were received:

$$\{68.49, 63.61, 71.22, 76.38, 75.99, 78.66, 59.08, 68.82, 75.47, 64.56\}$$

- (1) Estimate the required probability using both methods and compare the results.
- (2) Estimate the probability $P\left(X>72\right)$ using both methods and compare the results.

Answer: (1) with MEM:

$$\hat{\mu}_{MEM} = \frac{68.49 + 63.61 + 71.22 + 76.38 + 75.99 + 78.66 + 59.08 + 68.82 + 75.47 + 64.56}{10} = 70.22$$

$$P(X > 75) = 1 - P(X \le 75) = 1 - P\left(\frac{X - \mu}{\sigma} \le \frac{75 - \mu}{\sigma}\right) = 1 - \phi(0.9564) = 1 - 0.8306 = 0.1694$$

$$P(X > 75) = \frac{1}{10} \sum_{i=1}^{10} I\{X > 75\} = 0.4$$
(2)

$$P(X > 72) = 1 - P(X \le 72) = 1 - P\left(\frac{X - \mu}{\sigma^2} \le \frac{72 - 70.22}{5}\right) = 1 - \phi(0.0712) \approx 0.47$$

$$P(X > 72) = 1 - P(X \le 72) = 1 - P\left(\frac{X - \mu}{\sigma} \le \frac{72 - 70.22}{5}\right) = 1 - \phi(0.6391) \approx 0.36$$

$$P(X > 72) = \frac{1}{10} \sum_{i=1}^{10} I\{X > 72\} = 0.4$$

Problem 6. Let $X_1,...,X_n \stackrel{iid}{\sim} Poiss\left(\lambda\right)$

- (1) Compute the MSE of the MLE for λ .
- (2) A researcher believes that λ is approximately 3, so he suggests to use the estimator which is the average between the MLE and 3: $T = \frac{\overline{X_n} + 3}{2}$. Compute the MSE of T.
 - (3) Compare the bias and the variance of the estimators as functions of λ .
- (4) Compare the MSE of the estimators as a function of λ and find for which values of λ each estimator is better than the other. Note that the range of λ might depend on n.

Answer:

$$\mathcal{L}(\lambda; \mathbf{X}) = P(x_1 \cap x_2 \cap \dots \cap x_n; \lambda) = \prod_{i=1}^n \frac{\lambda^{x_i} e^{-\lambda}}{x_i!}$$

$$\iff \ell(\lambda; \mathbf{X}) = \sum_{i=1}^n x_i \cdot \log(\lambda) - n \cdot \lambda - \sum_{i=1}^n \log(x_i!)$$

$$\iff \frac{\partial \ell(\lambda; X)}{\partial \lambda} = \frac{\sum_{i=1}^n x_i}{\lambda} - n = 0 \implies \hat{\lambda}_{MLE} = \frac{1}{n} \sum_{i=1}^n x_i = \overline{X}$$

Notice that

$$E(\hat{\lambda}_{MLE} - \lambda) = E(\overline{X} - \lambda) = E\overline{X} - \lambda = 0$$

$$\Longrightarrow Bias(\hat{\lambda}_{MLE}) = 0$$

$$\begin{split} MSE\left(\hat{\lambda}_{MLE}\right) &= Var\left(\hat{\lambda}_{MLE}\right) + Bias^2\left(\hat{\lambda}_{MLE}\right) = Var\left(\hat{\lambda}_{MLE}\right) + 0 \\ &= Var\left(\frac{1}{n}\sum_{i}X_i\right) = \frac{1}{n^2} \cdot Var\left(\sum_{i}X_i\right) \stackrel{=}{\underset{iid}{=}} \frac{1}{n^2} \cdot n \cdot Var\left(X_i\right) \\ &= \frac{1}{n} \cdot \lambda \\ &\iff \end{split}$$

$$MSE\left(\hat{\lambda}_{MLE}\right) = \frac{\lambda}{n}$$

(2) we are given $T = \frac{3+\lambda}{2}$

$$E(T) = E\left(\frac{\overline{X} + 3}{2}\right) = \frac{1}{2}E\overline{X} + \frac{3}{2} = \frac{1}{2}\lambda + \frac{3}{2} \underset{\lambda \neq 3}{\neq} \lambda \implies T \text{ is Baised}$$

$$MSE(T) = Var(T) + Bias^{2}\left(\hat{\lambda}_{MLE}\right)$$

$$= \frac{1}{4}Var\left(\overline{X}\right) + \left(\frac{1}{2}\lambda + \frac{3}{2} - \lambda\right)^{2}$$

$$\stackrel{=}{\underset{iid}{=}} \frac{1}{4} \cdot \frac{\lambda}{n} + \frac{1}{4}(3 - \lambda)^{2}$$

$$\iff$$

$$MSE(T) = \frac{1}{2}\left[\lambda + (2 - \lambda)^{2}\right]$$

$$MSE(T) = \frac{1}{4} \left[\frac{\lambda}{n} + (3 - \lambda)^2 \right]$$
(3)

$$Var\left(T\right) = \frac{1}{4}Var\left(\overline{X}\right) < Var\left(\overline{X}\right) = Var\left(\hat{\lambda}_{MLE}\right) \ \forall n \in N$$

and

$$Bias^{2}(T) = \left(\frac{3-\lambda}{2}\right)^{2} \underset{\lambda \neq 3}{>} 0 = Bias^{2}(\hat{\lambda}_{MLE})$$

 $Bias^{2}(T) \underset{\lambda=3}{=} 0 = Bias^{2}(\hat{\lambda}_{MLE})$

We can say that if $\lambda \neq 3$ than T has higher squared bias and lower variance. (4) As $n \to \infty$ we get:

$$MSE\left(T\right) = \frac{1}{4} \left[\frac{\lambda}{n} + (3-\lambda)^2\right] \underset{n\to\infty}{=} \frac{(3-\lambda)^2}{4} \underset{\lambda\neq 3}{>} 0 = MSE\left(\hat{\lambda}_{MLE}\right)$$

while only if $\lambda = 3$ we get that

$$MSE\left(T\right) = 0 = MSE\left(\hat{\lambda}_{MLE}\right)$$

Problem 7. The weight of students in some university is normally distributed. A sample of 12 students is drawn with the following results (in Kg):

$$\{53.8, 67.34, 51.7, 52, 58.9, 74, 45.3, 53, 62.5, 48.87, 49, 55.6\}$$

- (1) Assuming that the variance is known and equals 1.5 Kg, calculate the confidence interval for the expected value of the weight with confidence level 95%.
- 2) Repeat part 1, this time for a confidence level of 90%. What can you say about the difference between the results?
- (3) Assuming that the variance is known and equals 2 Kg, calculate the confidence interval for the expected value of the weight with confidence level 95%. What can you conclude from the result?
 - (4) Repeat part 1, assuming that the variance is unknown.

Answer: (1)

$$\overline{X} = \frac{53.8 + 67.34 + 51.7 + 52 + 58.9 + 74 + 45.3 + 53 + 62.5 + 48.87 + 49 + 55.6}{12} \simeq 56$$

$$\frac{\sigma}{\sqrt{12}} = \sqrt{\frac{1.5}{12}} = 0.353$$

we can than write

$$P(C_{1}(X) \leq 56 \leq C_{2}(X)) = 0.95$$

$$\iff$$

$$P\left(-z_{97.5} \leq \frac{56}{\sigma/\sqrt{n}} \leq z_{97.5}\right) = 0.95$$

$$\implies CI_{0.95} = \left[56 - z_{97.5} \cdot \sigma/\sqrt{n}, 56 + z_{97.5} \cdot \sigma/\sqrt{n}\right]$$

$$= \left[56 - 1.96 \cdot 0.353, 56 + 1.96 \cdot 0.353\right]$$

$$= \left[55.30, 56.69\right]$$

(2)

$$CI_{0.9} = [56 - z_{95} \cdot \sigma/\sqrt{n}, 56 + z_{95} \cdot \sigma/\sqrt{n}]$$

= [56 - 1.65 \cdot 0.353, 56 + 1.65 \cdot 0.353]
= [55.41, 56.58]

As we allow for lower confidence that μ_X is in the interval that interval itself get smaller:

$$L(CI_{0.95}) = C_{2,0.95} - C_{1,0.95} = 56.69 - 55.30$$

= 1.39 > 1.17 = 56.58 - 55.41 = $C_{2,0.9} - C_{1,0.9} = L(CI_{0.9})$
 \iff
 $CI_{0.95} > CI_{0.9}$

(3) we now assume that $\sigma^2 = 2$ thus

$$\widetilde{CI}_{0.95} = \left[56 - z_{97.5} \cdot \sigma / \sqrt{n}, 56 + z_{97.5} \cdot \sigma / \sqrt{n} \right]$$

$$= \left[56 - 1.96 \cdot \sqrt{2/12}, 56 + 1.96 \cdot \sqrt{2/12}, \right]$$

$$= \left[55.2, 56.8 \right]$$

$$\Longrightarrow L\left(\widetilde{CI}_{0.95}\right) > L\left(CI_{0.95}\right)$$

A higher variance leeds to a larger CI

(4) Assuming the variance is unknow we have to use T-test confidence interval

$$S^2 = \frac{1}{12 - 1} \sum_{i=1}^{12} (x_i - 56)^2 = 69.64 \implies S = 8.34$$

$$CI_{0.95} = \left[56 - t_{11,97.5} \cdot S / \sqrt{n}, 56 + t_{11,97.5} \cdot S / \sqrt{n} \right]$$

= $\left[56 - 2.201 \cdot 2.4, 56 + 2.201 \cdot 2.4 \right]$
= $\left[50.71, 61.28 \right]$

Problem 8. Let $X \sim N(\mu, \sigma^2)$ (both parameters are unknown). In a random sample of 10 observations we received that:

$$\sum_{i=1}^{10} x_i = 15 \sum_{i=1}^{10} x_i^2 = 27$$

and the CI for μ is [1.09, 1.91]. What is the confidence level of this confidence interval?

Answer:

$$\overline{X} = \frac{1}{10} \sum_{i=1}^{10} x_i = \frac{15}{10} = 1.5$$

$$S^{2} = \frac{1}{9} \sum_{i=1}^{10} (x_{i} - \overline{X})^{2} = \frac{1}{9} \sum_{i=1}^{10} (x_{i}^{2} - 2x_{i}\overline{X} + \overline{X}^{2}) = \frac{1}{9} \left(\sum_{i=1}^{10} x^{2} - 2\overline{X} \sum_{i=1}^{10} x_{i} + \sum_{i=1}^{10} \overline{X}^{2} \right)$$

$$= \frac{1}{9} \left[27 - 2 \cdot 1.5 \cdot 15 + 10 \cdot 1.5^{2} \right] = 0.5$$

$$\implies S = 0.707$$

$$C_1 = \overline{X} - t_{n-1,1-\frac{\alpha}{2}} \cdot S/\sqrt{n} = 1.5 - t_{n-1,\frac{\alpha}{2}} \cdot \sqrt{0.5}/\sqrt{10} = 1.09$$

$$\iff t_{9,\frac{\alpha}{2}} = \frac{1.5 - 1.09}{\sqrt{0.5/10}} = 1.833$$

$$\iff \alpha/2 = 0.05$$

$$1 - 0.1 \approx 0.9$$

the confidence level is 0.9

Problem 9. In a random sample of 100 students, it was found that 30 like Bamba.

- (1) Compute an asymptotic confidence interval for the proportion of Bamba lovers among the students.
- (2) Find the minimal sample size n for which the length of the CI will be at most 0.02.

Answer: (1) We assume all students like or deslike Bamba and thus $X_1, \ldots, X_{100} \sim Bin(\theta)$ given the information we can assume

$$\overline{X} = 0.3 = \hat{\theta}$$

and hence $E\hat{\theta}=0.3~Var\left(\hat{\theta}\right)=0.3\left(1-0.3\right)=0.21$ and we get that The asymptotic confidence interval for $\hat{\theta}$ is:

$$\left[n \cdot \hat{\theta} \pm \frac{\sqrt{n\hat{\theta} \left(1 - \hat{\theta} \right)}}{\sqrt{n}} z_{1 - \frac{\alpha}{2}} \right] = \left[0.3 \pm \frac{\sqrt{0.21}}{10} z_{1 - \frac{\alpha}{2}} \right] = \left[0.3 \pm 0.0458 \cdot z_{1 - \frac{\alpha}{2}} \right]$$
(2)

$$n = 4 \cdot z_{1-\frac{\alpha}{2}}^2 \cdot \frac{\sigma^2}{L_0^2} = 4 \cdot z_{1-\frac{\alpha}{2}}^2 \cdot \frac{0.21}{0.02^2} = 0.21 \cdot 10^4 \cdot z_{1-\frac{\alpha}{2}}^2$$

for $z_{0.975} = 1.96$ we get $n \ge 0.21 \cdot 10^4 \cdot 1.96^2 \approx 8068$