II Università di Roma, Tor Vergata Dipartimento d'Ingegneria Civile e Ingegneria Informatica LM in Ingegneria dell'Informazione e dell'Automazione Complementi di Probabilità e Statistica Homework - 2019-12-06

Problem 1 Let $(\Omega, \mathcal{E}, \mathbf{P}) \equiv \Omega$ be a probability space and let $(X_n)_{n\geq 1}$ be a sequence of independent identically distributed Bernoulli random variables with success probability p. Set

$$Z_n \stackrel{def}{=} \sum_{k=1}^n X_n, \qquad \bar{X}_n = \frac{1}{n} \sum_{n=1}^n X_k.$$

Assume that n is large. What you can say about the distributions, expectation, and variance of Z_n and \bar{X}_n ? Consider the case n=100,000 and p=1/2. Use both the Central Limit Theorem and the Tchebychev inequality to estimate the probability that Z_n lies between 49,500 and 50,500. What you can say about the distributions, expectation, and variance of Z_n and \bar{X}_n if $(X_n)_{n\geq 1}$ is a sequence of independent and Poisson distributed random variables with the same rate parameter λ ?

Solution. Independently of n, the random variable Z_n has the binomial distribution with parameters n and p. In symbols

$$Z_n \sim Bin(n, p)$$
.

As a consequence,

$$\mathbf{E}[Z_n] = np$$
 and $\mathbf{D}^2[Z_n] = np(1-p)$.

By virtue of the Central Limit Theorem, we know that

$$\frac{Z_n - np}{\sqrt{np(1-p)}} \xrightarrow{\mathbf{w}} N(0,1).$$

Otherwise saying: as n is large, the distribution of $(Z_n - np) / \sqrt{np(1-p)}$ is approximately the standard normal distribution. On the other hand, we can write

$$\sqrt{\frac{p(1-p)}{n}}\frac{Z_n - np}{\sqrt{np(1-p)}} = \frac{Z_n}{n} - p = \bar{X}_n - p,$$

that is

$$\bar{X}_n = \sqrt{\frac{p(1-p)}{n}} \frac{Z_n - np}{\sqrt{np(1-p)}} + p$$

This implies that, as n is large, the distribution of \bar{X}_n is approximately normal with

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

In the case n = 100,000 and p = 1/2, thanks to the Central Limit Theorem, we can write

$$\mathbf{P}(49, 500 \le Z_n \le 50, 500) = \mathbf{P}(49, 500 - np \le Z_n - np \le 50, 500 - np)$$

$$= \mathbf{P}\left(\frac{49, 500 - np}{\sqrt{np(1-p)}} \le \frac{Z_n - np}{\sqrt{np(1-p)}} \le \frac{50, 500 - np}{\sqrt{np(1-p)}}\right)$$

$$\simeq \Phi\left(\frac{50, 500 - np}{\sqrt{np(1-p)}}\right) - \Phi\left(\frac{49, 500 - np}{\sqrt{np(1-p)}}\right)$$

$$= \Phi\left(\frac{500}{\sqrt{25,000}}\right) - \Phi\left(\frac{-500}{\sqrt{25,000}}\right)$$

$$= 2\Phi\left(\sqrt{10}\right) - 1$$

$$= 2 \cdot 0.9992 - 1$$

$$= 0.9984.$$

where Φ is the distribution function of the standard normal. Instead, with the goal of applying the Tchebychev inequality, we can write

$$\mathbf{P}(49, 500 \le Z_n \le 50, 500) = \mathbf{P}\left(\frac{49, 500 - np}{\sqrt{np(1-p)}} \le \frac{Z_n - np}{\sqrt{np(1-p)}} \le \frac{50, 500 - np}{\sqrt{np(1-p)}}\right)$$

$$= \mathbf{P}\left(-\sqrt{10} \le \frac{Z_n - 50, 000}{50\sqrt{10}} \le \sqrt{10}\right)$$

$$= \mathbf{P}\left(\left|\frac{Z_n - 50, 000}{50\sqrt{10}}\right| \le \sqrt{10}\right)$$

$$= 1 - \mathbf{P}\left(\left|\frac{Z_n - 50, 000}{50\sqrt{10}}\right| > \sqrt{10}\right)$$

$$= 1 - \mathbf{P}\left(\left|\frac{Z_n - 50, 000}{50\sqrt{10}}\right| \ge \sqrt{10}\right).$$

On the other hand, by the Tchebychev inequality we have

$$\mathbf{P}\left(\left|\frac{Z_n - 50,000}{50\sqrt{10}}\right| \ge \sqrt{10}\right) \le \frac{\mathbf{D}^2\left[\frac{Z_n - 50,000}{50\sqrt{10}}\right]}{10} = \frac{1}{10}.$$

Therefore,

$$\mathbf{P}(49,500 \le Z_n \le 50,500) \ge 1 - \frac{1}{10} = \frac{9}{10} = 0.9.$$

This shows that the central limit approach provides a sharper bound for the desired probability than the Tchebychev inequality approach.

Problem 2 Suppose that a random variable X, which represents the reaction time at some stimulus, has a uniform distribution on an interval $[0,\theta]$, where the parameter $\theta > 0$ is unknown. An investigator wants to estimate θ on the basis of a simple random sample X_1, \ldots, X_n of reaction times. Since θ is the largest possible time in the entire population of reaction times, the investigator consider as a first estimator for the parameter θ the largest sample reaction time. That is to say, the investigator consider as a first estimator the statistic

$$\hat{\theta}_1 \equiv \check{X}_n \equiv \max(X_1, \dots, X_n).$$

- 1. Is \check{X}_n unbiased? In case \check{X}_n is not unbiased, is it possible to derive from \check{X}_n an unbiased estimator of θ ?
- 2. As a second estimator, the investigator consider the statistic

$$\hat{\theta}_2 \equiv \bar{X}_n \equiv \frac{1}{n} \sum_{k=1}^n X_k.$$

Is \bar{X}_n unbiased? In case \bar{X}_n is not unbiased, is it possible to derive from \bar{X}_n an unbiased estimator of θ ?

3. In the investigator's shoes, what estimator would you prefer among those considered?

Solution.

1. Writing $F_{\check{X}_n}: \mathbb{R} \to \mathbb{R}$ for the distribution function of the statistic \check{X}_n , we have

$$F_{\check{X}_n}(x) = \mathbf{P}\left(\check{X}_n \le x\right) = \mathbf{P}\left(X_1 \le x, \dots, X_n \le x\right) = \prod_{k=1}^n \mathbf{P}\left(X_k \le x\right)$$
$$= \prod_{k=1}^n \mathbf{P}\left(X \le x\right) = \mathbf{P}\left(X \le x\right)^n = F_X(x)^n.$$

for every $x \in \mathbb{R}$, where $F_X : \mathbb{R} \to \mathbb{R}$ is the distribution function of the random variable X. On the other hand, since X is uniformly distributed on $[0, \theta]$, it has density $f_X : \mathbb{R} \to \mathbb{R}$ given by

$$f_X(x) \stackrel{\text{def}}{=} \frac{1}{\theta} 1_{[0,\theta]}(x), \quad \forall x \in \mathbb{R}.$$

Hence,

$$F_{X}(x) = \int_{(-\infty,x]} f_{X}(u) \ d\mu_{L}(u) = \int_{(-\infty,x]} \frac{1}{\theta} 1_{[0,\theta]}(u) \ d\mu_{L}(u) = \frac{1}{\theta} \int_{(-\infty,x] \cap [0,\theta]} d\mu_{L}(u)$$

$$= \begin{cases} \frac{1}{\theta} \int_{\varnothing} d\mu_{L}(u) = 0 & \text{if } x < 0 \\ \frac{1}{\theta} \int_{[0,x]} d\mu_{L}(u) = \frac{x}{\theta} & \text{if } 0 \le x \le \theta \\ \frac{1}{\theta} \int_{[0,\theta]} d\mu_{L}(u) = 1 & \text{if } \theta < x \end{cases}$$

$$= \frac{x}{\theta} 1_{[0,\theta]}(x) + 1_{(\theta,+\infty)}(x)$$

It then follows,

$$F_{\check{X}_n}(x) = F_X(x)^n = \frac{x^n}{\theta^n} 1_{[0,\theta]}(x) + 1_{(\theta,+\infty)}(x),$$

for every $x \in \mathbb{R}$. Now, we have

$$F'_{\check{X}_n}(x) = \begin{cases} 0 & \text{if } x < 0\\ \frac{nx^{n-1}}{\theta^n} & \text{if } 0 < x < \theta\\ 0 & \text{if } \theta < x \end{cases}.$$

Note that $F_{\check{X}_n}$ is not everywhere differentiable. Eventually, is not differentiable in the point $x = \theta$. However, considering the function $f_{\check{X}_n} : \mathbb{R} \to \mathbb{R}$ given by

$$f_{\check{X}_{n}}\left(x\right)\stackrel{\mathrm{def}}{=}\frac{nx^{n-1}}{\theta^{n}}1_{\left(0,\theta\right)}\left(x\right),\quad\forall x\in\mathbb{R},$$

a straightforward computation shows that

$$F_{\check{X}_n}(x) = \int_{(-\infty, x]} f_{\check{X}_n}(u) \ d\mu_L(u),$$

for every $x \in \mathbb{R}$. This implies that \check{X}_n is absolutely continuous with density $f_{\check{X}_n}$. As a consequence,

$$\mathbf{E}\left[\check{X}_{n}\right] = \int_{\mathbb{R}} x f_{\check{X}_{n}}(x) \ d\mu_{L}(x) = \int_{\mathbb{R}} x \frac{n x^{n-1}}{\theta^{n}} 1_{(0,\theta)}(x) \ d\mu_{L}(x) = \frac{n}{\theta^{n}} \int_{(0,\theta)}^{\theta} x^{n} \ dx = \frac{n}{\theta^{n}} \int_{0}^{\theta} x^{n} \ dx = \frac{n}{\theta^{n}} \frac{x^{n+1}}{n+1} \Big|_{0}^{\theta} = \frac{n}{n+1} \theta.$$

We can conclude that \check{X}_n is not a unbiased estimator of θ but $\frac{n+1}{n}\check{X}_n$ is an unbiased estimator of θ .

2. We have

$$\mathbf{E}\left[\bar{X}_{n}\right] = \mathbf{E}\left[X\right] = \int_{\mathbb{R}} x f_{X}\left(x\right) d\mu_{L}\left(x\right) = \int_{\mathbb{R}} \frac{x}{\theta} 1_{[0,\theta]}\left(x\right) d\mu_{L}\left(x\right)$$
$$= \frac{1}{\theta} \int_{[0,\theta]} x d\mu_{L}\left(x\right) = \frac{1}{\theta} \int_{0}^{\theta} x dx = \frac{1}{\theta} \left.\frac{x^{2}}{2}\right|_{0}^{\theta} = \frac{\theta}{2}.$$

Hence, \bar{X}_n is not a unbiased estimator of θ but $2\bar{X}_n$ is an unbiased estimator of θ .

3. From 1. and 2. we know that

$$\mathbf{E}\left[\frac{n+1}{n}\check{X}_n\right] = \theta$$
 and $\mathbf{E}\left[2\bar{X}_n\right] = \theta$.

Hence, both $\frac{n+1}{n}\check{X}_n$ and $2\bar{X}_n$ are unbiased estimators of the parameter θ . To choose which is preferable between them, we consider

$$\mathbf{D}^2 \left[\frac{n+1}{n} \check{X}_n \right]$$
 and $\mathbf{D}^2 \left[2\bar{X}_n \right]$.

We have

$$\mathbf{E}\left[\check{X}_{n}^{2}\right] = \int_{\mathbb{R}} x^{2} f_{\check{X}_{n}}\left(x\right) d\mu_{L}\left(x\right) = \int_{\mathbb{R}} x^{2} \frac{nx^{n-1}}{\theta^{n}} 1_{(0,\theta)}\left(x\right) d\mu_{L}\left(x\right) = \frac{n}{\theta^{n}} \int_{(0,\theta)}^{\theta} x^{n+1} dx = \frac{n}{\theta^{n}} \left. \int_{0}^{\theta} x^{n+1} dx = \frac{n}{\theta^{n}} \left. \frac{x^{n+2}}{n+2} \right|_{0}^{\theta} = \frac{n}{n+2} \theta^{2}.$$

Therefore,

$$\mathbf{D}^{2} \left[\check{X}_{n} \right] = \mathbf{E} \left[\check{X}_{n}^{2} \right] - \mathbf{E} \left[\check{X}_{n} \right]^{2} = \frac{n}{n+2} \theta^{2} - \frac{n^{2}}{(n+1)^{2}} \theta^{2} = \frac{n}{(n+1)^{2} (n+2)} \theta^{2}.$$

As a consequence,

$$\mathbf{D}^{2} \left[\frac{n+1}{n} \check{X}_{n} \right] = \left(\frac{n+1}{n} \right)^{2} \mathbf{D}^{2} \left[\check{X}_{n} \right] = \left(\frac{n+1}{n} \right)^{2} \frac{n}{(n+1)^{2} (n+2)} \theta^{2} = \frac{\theta^{2}}{n (n+2)}.$$

On the other hand,

$$\mathbf{D}^{2}\left[2\bar{X}_{n}\right] = 4\mathbf{D}^{2}\left[\bar{X}_{n}\right] = \frac{4}{n}\mathbf{D}^{2}\left[X\right] = \frac{4}{n}\frac{\theta^{2}}{12} = \frac{\theta^{2}}{3n}.$$

Now, for any n > 1 we clearly have

$$\mathbf{D}^2 \left[\frac{n+1}{n} \check{X}_n \right] < \mathbf{D}^2 \left[2\bar{X}_n \right].$$

It follows that the estimator $\frac{n+1}{n}\check{X}_n$ is preferable to $2\bar{X}_n$.

Exercise 3 Let X be a binomially distributed real random variable with number of trials parameter m and unknown success parameter p. An investigator wants to estimate p on the basis of a simple random sample X_1, \ldots, X_n of size n drawn from X.

- 1. Assume the investigator applies the method of moments. What is the estimator $\hat{p}_{M,n}$?
- 2. Assume the investigator applies the likelihood method. What is the estimator $\hat{p}_{ML,n}$?

Solution.

Exercise 4 Let X be a normally distributed random variable with unknown mean μ and variance σ^2 . An investigator wants to estimate μ and σ^2 on the basis of a simple random sample X_1, \ldots, X_n of size n drawn from X.

- 1. Assume the investigator applies the likelihood methods. What are the estimator $\hat{\mu}_{LM}$ and $\hat{\sigma}_{LM}^2$?
- 2. Assume the investigator applies the method of moments. What are the estimators $\hat{\mu}_{MM}$ and $\hat{\sigma}_{MM}^2$? Hint: guess what $\hat{\sigma}_{MM}^2$ could be and get it!

Solution.

Exercise 5 Let X a random variable representing a characteristic of a certain population. Assume that X has a density $f_X : \mathbb{R} \to \mathbb{R}$ given by

$$f_X\left(x\right) \stackrel{def}{=} \frac{1}{\theta} e^{-\frac{x-3}{\theta}} 1_{[3,+\infty)}\left(x\right), \quad \forall x \in \mathbb{R},$$

where θ is a positive parameter.

- 1. Apply the method of moments to find the estimator $\hat{\theta}_M$ of the parameter θ .
- 2. Apply the maximum likelihood method to find the estimator $\hat{\theta}_{ML}$ of the parameter θ .
- 3. Use the estimators $\hat{\theta}_M$ and $\hat{\theta}_{ML}$ to build estimators for $\mathbf{E}[X]$ and $\mathbf{D}^2[X]$.

Solution.

Exercise 6 Assume that the returns of a stock in a financial market are normally distributed with unknown mean μ and variance σ^2 . Let X be the normal random variable representing the realization of the returns and let X_1, \ldots, X_n be a simple random sample of size n drawn from X. Assume that n = 5 and the realizations of the sample are

$$x_1 \equiv -1.5, \quad x_2 \equiv -0.5, \quad x_3 \equiv 1.5, \quad x_4 \equiv 2.0, \quad x_5 \equiv 2.5$$

- 1. Determine a 99% confidence interval for the mean μ .
- 2. Find the confidence for an interval of width 0.1.

3. Determine a 90% confidence interval for the standard deviation σ .

Solution.

Exercise 7 Assume that a library master believes that the mean duration in days of the borrowing period is 20d. However, the library master selects a simple random sample of 100 books in the library and discovers that the sample mean and variance of the borrowing days are 18d and 8d², respectively. Determine a 99% confidence interval for the mean duration of the borrowing days to check whether library master's initial guess is correct.

Solution.

Exercise 8 The mark of a infamous exam of Probability and Statistics are normally distributed with standard deviation $\sigma = 2$. A simple random sample of nine students is selected end the following evaluations are computed

$$\sum_{k=1}^{9} x_k = 237$$
 and $\sum_{k=1}^{9} x_k^2 = 6295$.

- 1. Find a 90% confidence interval for the mean mark.
- 2. Discuss, without computation, whether the length of a 95% confidence interval would be smaller, greater or equal than the length of the interval previously determined.
- 3. How large the minimum sample size should be to obtain a 90% confidence interval for the mean mark with width equal to 3? Besides the confidence interval method is it possible to apply the Tchebychev inequality?

Solution.

Exercise 9 Let $X_1, \ldots, X_n, X_{n+1}$ be a simple random sample of size n+1 drawn from a Gaussian distributed random variable X with unknown mean μ and variance σ^2 . Assume that we have observed X_1, \ldots, X_n and we want use the observed values x_1, \ldots, x_n to determine a confidence interval for the prediction of X_{n+1} . To this goal give detailed answers to the following questions:

- 1. what is the distribution of the statistic \bar{X}_n ?
- 2. what is the distribution of the statistic $(X_{n+1} \bar{X}_n)/\sigma\sqrt{1+1/n}$?
- 3. are the statistics $X_{n+1} \bar{X}_n$ and $S_n^2 \equiv \frac{1}{n-1} \sum_{k=1}^n (X_k \bar{X}_n)^2$ independent?
- 4. what is the distribution of the statistic $(X_{n+1} \bar{X}_n)/S_n\sqrt{1+1/n}$?

After answering the above questions, build an interval in which the random variable X_{n+1} takes its values with probability α and determine the corresponding confidence interval for the prediction of X_{n+1} . In the end, assume that n=7 and we have

$$x_1 = 7005$$
, $x_2 = 7432$, $x_3 = 7420$, $x_4 = 6822$, $x_5 = 6752$, $x_6 = 5333$, $x_7 = 6552$.

compute the 95% confidence interval for the prediction of X_8 .

Exercise 10 Let X be a Gaussian random variable with unknown mean μ_X and variance σ_X^2 representing a certain characteristic of a population. Assume that testing the sample mean \bar{X}_n and the sample standard deviation S_n of a simple random sample X_1, \ldots, X_n of size $n \equiv 9$ drawn from X we obtain the value $\bar{X}_n(\omega) \equiv \bar{x}_n = 251.50$ cm and $S_n(\omega) \equiv s_n = 2.30$ cm.

- 1. Considering both the rejection region method and the p-value method, should the null hypothesis $H_0: \mu_X = 250 cm$ be rejected against the alternative $H_a: \mu_X \neq 250 cm$ at the significance level $\alpha = 0.1$?
- 2. Considering both the rejection region method and the p-value method, should the null hypothesis $H_0: \sigma_X^2 = 4$ be rejected against of the alternative $H_a: \sigma_X^2 > 4$ at the significance level $\alpha = 0.05$? Calculate the probability β (5) of a II type error.

Solution.

Exercise 11 Let X be a Gaussian random variable with unknown mean μ and variance σ^2 representing a certain characteristic of a population and let X_1, \ldots, X_n be a simple random sample of size n drawn from X. Assume that n=25 and that the realizations x_1, \ldots, x_{25} of the sample give an information summarized by

$$\sum_{k=1}^{25} x_k = 100$$
 and $\sum_{k=1}^{25} x_k^2 = 560$

- 1. Considering both the rejection region method and the p-value method, should the null hypothesis $H_0: \sigma^2 = 4$ be rejected against of the alternative $H_1: \sigma^2 > 4$ with a significance level $\alpha = 0.05$? Calculate the probability β (5) of a II type error.
- 2. Considering both the rejection region method and the p-value method, should the null hypothesis $H_0: \sigma^2 = 4$ be rejected against of the alternative $H_1: \sigma^2 \neq 4$ with a significance level $\alpha = 0.05$? Calculate the probability β (5) of a II type error.

Solution.

Exercise 12 In order to measure the dependence between two random variables X and Y a simple sample of size 10 is drawn from the random vector (X,Y) and the following quantities are computed

$$\bar{x}_{10} = 49.6670, \quad \bar{y}_n = -0.4333, \quad s_x^2 = 236.1390, \quad s_y^2 = 0,1750, \quad \gamma_{x,y} = 5.8072.$$

- 1. May you find the equation of the regression live of Y against X?
- 2. May you find the estimated mean square error?
- 3. What value of y would you predict for a corresponding value of x?