Introduction to Probability & Statistics

Assignment 5, 2024/25

Assignment Questions - answers to be uploaded

AQ1. Let X_1, \ldots, X_{16} be an i.i.d. sample from a N(3,1) distribution, and let $S = X_1 + X_2 + \cdots + X_{16}$. Express $\mathbb{P}(S < 52)$ in terms of the distribution function Φ of the standard normal distribution.

Answer

We know that $\mathbb{E}[S] = 16\mathbb{E}[X] = 48$, and Var(S) = 16Var(X) = 16 (because the X_i are independent, and hence uncorrelated).

Furthermore, we know that the sum of independent normal distributions has a normal distribution. Thus

$$S \sim N(48, 16)$$
.

[3]

We can normalize to obtain a standard normal random variable by subtracting the mean and dividing by the standard deviation. Thus

$$\mathbb{P}(S < 52) = \mathbb{P}\left(\frac{S - 48}{\sqrt{16}} < \frac{52 - 48}{\sqrt{16}}\right) = \Phi(1).$$

[2]

AQ2. Let Y_1, Y_2, \ldots be an i.i.d. sequence of random variables, each with a Uniform(0,3) distribution. Define a new sequence of random variables X_1, X_2, \ldots by

$$X_n = \frac{1}{n} \sum_{i=1}^n Y_i^2$$
.

Using the Law of Large Numbers, determine the value of $a \in \mathbb{R}$ for which $\mathbb{P}(\lim_{n \to \infty} X_n = a) = 1$.

Answer

The random variable X_n is just the sample mean of the random variables Y_1^2, \ldots, Y_n^2 . These are i.i.d. and clearly have finite mean and variance (since Y_i^2 can only take values in the finite set [0,9]).

The (strong) law of large numbers says that, with probability one, X_n will converge to $\mathbb{E}[Y^2]$. [3] Finally, we calculate

$$\mathbb{E}\left[Y^{2}\right] = \int_{0}^{3} y^{2} \frac{1}{3} dy = 3,$$

and so the required answer is a=3. [2]

AQ3. Suppose the random variables X_1, X_2 and X_3 all have the same expectation μ . For what values of a and b is

$$M = -4(X_1 - 2) + 9(X_2 - 1) + aX_3 + b$$

an unbiased estimator for μ ?

Answer

M is an unbiased estimator for μ if $\mathbb{E}[M] = \mu$ for any value of μ . [1] We find

$$\mathbb{E}[M] = \mathbb{E}[-4(X_1 - 2) + 9(X_2 - 1) + aX_3 + b]$$

$$= -4(\mathbb{E}[X_1] - 2) + 9(\mathbb{E}[X_2] - 1) + a\mathbb{E}[X_3] + b$$

$$= -4(\mu - 2) + 9(\mu - 1) + a\mu + b$$

$$= (5 + a)\mu + (b - 1).$$

[2]

Thus $\mathbb{E}\left[\mu\right]=\mu$ if and only if a=-4 and b=1. [2]

AQ4. From a dataset x_1, \ldots, x_{10} it has been calculated that

$$\sum_{i=1}^{10} x_i = 491, \quad \sum_{i=1}^{10} (x_i - \bar{x}_{10})^2 = 41.$$

You model the dataset as a random sample from the normal distribution with mean μ and variance σ^2 .

- a) Assume that both μ and σ^2 are unknown. Determine a 95% confidence interval for the mean μ . You can use that $t_{9.0.025}\approx 2.26$.
- b) Now assume that it is known that the variance is $\sigma^2=5$. Give a 95% confidence interval for the mean μ in this case. You can use that $z_{0.025}\approx 1.96$.

Answer

a) Since the variance is unknown, we use the interval

$$\left(\bar{x}_n - t_{n-1,\alpha/2} \frac{s_n}{\sqrt{n}}, \bar{x}_n + t_{n-1,\alpha/2} \frac{s_n}{\sqrt{n}}\right)$$

We first calculate $\bar{x}_n = 491/10 = 49.1$ and

$$s_n^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x}_n)^2 = \frac{1}{10-1} 41 = \frac{41}{9}.$$

We're told that $t_{n-1,\alpha/2}=t_{9,0.025}=2.26$ and

$$\left(\bar{x}_n - t_{n-1,\alpha/2} \frac{s_n}{\sqrt{n}}, \bar{x}_n + t_{n-1,\alpha/2} \frac{s_n}{\sqrt{n}}\right) \approx \left(49.1 - 2.26 \frac{\sqrt{41}}{3\sqrt{10}}, 49.1 + 2.26 \frac{\sqrt{41}}{3\sqrt{10}}\right)$$
$$\approx (47.57, 50.63).$$

[3]

b) When σ^2 is known, we use the confidence interval

$$\left(\bar{x}_n - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{x}_n + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}\right) \approx \left(49.1 - 1.96 \frac{\sqrt{5}}{\sqrt{10}}, 49.1 + 1.96 \frac{\sqrt{5}}{\sqrt{10}}\right)$$
$$\approx (47.71, 50.49).$$

[2]