

IPS: Assignment 5

Due date: 10am, 07/12/23

Stephen Connor

Instructions

Submit your answers to all questions marked  **Hand-in**. You should upload to [Moodle](#) your solutions as a *pdf file*.

Marks will be awarded for clear, logical explanations, as well as for correctness of solutions. You should also look at the other questions in preparation for your Week 11 seminar.

Starter

These questions should help you to gain confidence with the basics.

S1. Hand-in

Let X_1, \dots, X_{16} be an i.i.d. sample from a $N(3, 1)$ distribution, and let $S = X_1 + X_2 + \dots + X_{16}$. Express $\mathbb{P}(S < 40)$ in terms of the distribution function Φ of the standard normal distribution.

S2. You perform 28 independent experiments measuring a random variable X which you know has mean 457 and variance 676. Without using the central limit theorem, give a lower bound on the probability that the average of your measurements is between 433 and 481.

S3. Let X_1, \dots, X_n be an i.i.d. sample from a distribution with mean μ and variance σ^2 . Let \bar{X}_n denote the sample mean, and define

$$Z_n = \frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma}.$$

Check that $\mathbb{E}[Z_n] = 0$ and $\text{Var}(Z_n) = 1$.

S4. Hand-in

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

$$M = -4(X_1 - 1) + 9(X_2 - 1) + a(X_3 - 1) + b$$

an unbiased estimator for μ ?

S5. From a dataset x_1, \dots, 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 $z_{0.025} \approx 1.96$.

S6. Lengths of baguettes are assumed to follow a $N(\mu, \sigma^2)$ distribution. Six baguettes were measured, giving the following lengths in cms: 66, 69, 62, 62, 64, 67.

- a) Calculate unbiased estimates for μ and σ^2 .
b) Calculate a 90% confidence interval for μ .

Main course

These are important, and cover some of the most substantial parts of the course.

M1. 📁 Hand-in

If X has expectation μ and standard deviation σ , the ratio $r = |\mu|/\sigma$ is called the *measurement signal-to-noise-ratio* of X . If we define $D = |(X - \mu)/\mu|$ as the *relative deviation* of X from its mean μ , show that, for $\alpha > 0$,

$$\mathbb{P}(D < \alpha) \geq 1 - \frac{1}{r^2 \alpha^2}.$$

M2. In the lecture I proved Chebychev's inequality for the case of a continuous random variable. Provide a similar proof for the case of a discrete random variable.

M3. 📁 Hand-in

Let Y_1, Y_2, \dots be an i.i.d. sequence of random variables, each with a Uniform $(-2, 2)$ distribution. Define a new sequence of random variables X_1, X_2, \dots by

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

For what value of $a \in \mathbb{R}$ is it true that $\mathbb{P}(\lim_{n \rightarrow \infty} X_n = a) = 1$?

M4. Let X be the number of 1s and Y be the number of 2s that occur in n rolls of a fair die. Use indicator random variables to compute $\text{Cov}(X, Y)$ and $\rho(X, Y)$.
Hint: this is just like the smarties example covered in lectures.

M5. Assume that in Example 17.1 from the lectures (measuring a ball rolling down an inclined plane), we choose to stop the ball always after one time unit, so that

$$X_i = \frac{1}{2}a(1 + U_i)^2 + V_i,$$

where the independent errors are normally distributed with $U_i \sim N(0, \sigma_U^2)$, $V_i \sim N(0, \sigma_V^2)$. Assume the variances of the errors are known. Calculate the bias of the estimator $A = 2\bar{X}_n$ for the acceleration parameter a . Propose an unbiased estimator for a .

M6. Consider the following dataset of lifetimes of ball bearings in hours:

6278 3113 5236 11584 12628 7725 8604 14266 6125 9350
3212 9003 3523 12888 9460 13431 17809 2812 11825 2398.

Suppose that we are interested in estimating the minimum lifetime of this type of ball bearing. The dataset is modelled as a realization of a random sample X_1, \dots, X_n . Each random variable X_i is represented as $X_i = \delta + Y_i$, where Y_i has an $\text{Exp}(\lambda)$ distribution and $\delta > 0$ is an unknown parameter

that is supposed to model the minimum lifetime. The objective is to construct an unbiased estimator for δ . It is known that

$$\mathbb{E}[M_n] = \delta + \frac{1}{n\lambda} \text{ and } \mathbb{E}[\bar{X}_n] = \delta + \frac{1}{\lambda},$$

where $M_n = \min(X_1, \dots, X_n)$ and $\bar{X}_n = (X_1 + \dots + X_n)/n$.

a) Check whether

$$T = \frac{n}{n-1} (\bar{X}_n - M_n)$$

is an unbiased estimator for $1/\lambda$.

b) Construct an unbiased estimator D for δ .

c) Use the dataset to compute an estimate for the minimum lifetime δ .

M7. Let X_1, \dots, X_n be an i.i.d. sample from the normal distribution with mean μ and variance σ^2 .

a) Give a $100(1 - \alpha)\%$ confidence interval estimator for μ in the case where σ^2 is known. Hence or otherwise, find a 95% confidence interval for μ in the case where $\sigma^2 = 9$ and a random sample of size 16 has been taken with values x_1, \dots, x_{16} and it has been found that

$$\sum_{i=1}^{16} x_i = 50, \quad \sum_{i=1}^{16} (x_i - \bar{x}_{16})^2 = 115.$$

b) It is proposed that, from a second independent random sample of size 16 a 99% confidence interval for μ be constructed and that, from a third independent random sample of size 32, a 98% confidence interval for μ be constructed. State the probability that *neither* of these two confidence intervals will contain μ .

M8. Recall that we say that T_m has a t -distribution with $m > 1$ degrees of freedom, and write $T_m \sim t(m)$, if it has density function given by

$$f(x) = k_m \left(1 + \frac{x^2}{m} \right)^{-\frac{m+1}{2}}, \quad x \in \mathbb{R},$$

where k_m is a constant that ensures that the density integrates to 1. If $T_m \sim t(m)$, show that $\mathbb{E}[T_m] = 0$.

Hint: you shouldn't need to explicitly calculate any integrals here!

Dessert

Still hungry for more? Try these if you want to push yourself further.

D1. Let M_n be the maximum of n independent Uniform(0, 1) random variables. Show that for any fixed $\varepsilon > 0$,

$$\lim_{n \rightarrow \infty} \mathbb{P}(|M_n - 1| > \varepsilon) = 0.$$

D2. (A more general law of large numbers, see Exercise 13.12 in the textbook).

Let X_1, X_2, \dots be a sequence of independent random variables with $\mathbb{E}[X_i] = \mu_i$ and $\text{Var}(X_i) = \sigma_i^2$ for $i = 1, 2, \dots$. Let $\bar{X}_n = (X_1 + \dots + X_n)/n$. Suppose that there exists an $M \in \mathbb{R}$ such that $0 < \sigma_i^2 \leq M$ for all i , and let a be an arbitrary positive number.

a) Apply Chebychev's inequality to show that

$$\mathbb{P}\left(\left|\bar{X}_n - \frac{1}{n} \sum_{i=1}^n \mu_i\right| > a\right) \leq \frac{\text{Var}(X_1) + \dots + \text{Var}(X_n)}{n^2 a^2}.$$

b) Conclude from a) that

$$\lim_{n \rightarrow \infty} \mathbb{P} \left(\left| \bar{X}_n - \frac{1}{n} \sum_{i=1}^n \mu_i \right| > a \right) = 0.$$

c) Check that the weak law of large numbers is a special case of this result.

! Challenge question

Suppose that X_1, \dots, X_n are mutually independent random variables, each distributed as $\text{Exp}(\lambda)$. (That is, all events of the kind $\{X_1 \leq x_1\}, \dots, \{X_n \leq x_n\}$ are mutually independent.) Let $Y_n = \max\{X_1, \dots, X_n\}$, and $V_n = \min\{X_1, \dots, X_n\}$.

a) Show that $Y_n \sim \text{Exp}(\lambda n)$.

b) What is the distribution function of V_n ?

c) Show that, for all $s > 0$,

$$\lim_{n \rightarrow \infty} \mathbb{P}(V_n - (\log n)/\lambda \leq s) = \exp(-e^{-\lambda s}).$$