

Problem Set 5

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Introduction

As an application of some of the properties of expected values, this problem set steps through a proof that the expected value of the random variable that defines sample variance is the population variance, given that the population variance is defined.

For each of these questions, let X_1, X_2, \dots, X_n be independent, identically distributed random variables with mean μ and variance σ^2 .

Please complete the following tasks regarding the data in R. Please generate a solution document in R markdown and upload the .Rmd document and a rendered .doc, .docx, or .pdf document. Please turn in your work on Canvas.

These questions were rendered in R markdown through RStudio (<https://www.rstudio.com/wp-content/uploads/2015/02/rmarkdown-cheatsheet.pdf>, <http://rmarkdown.rstudio.com>).

Question 1 (10 points)

Let X_1, X_2, \dots, X_n be independent, identically distributed random variables with mean μ and variance σ^2 , and define the random variable \bar{X} by $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$. Justify the equality

$$\begin{aligned} E \left[\sum_{i=1}^n (X_i - \bar{X})^2 \right] &= E \left[\sum_{i=1}^n X_i^2 \right] - 2E \left[\sum_{i=1}^n \bar{X} X_i \right] + E \left[\sum_{i=1}^n \bar{X}^2 \right] \\ E \left[\sum_{i=1}^n (X_i - \bar{X})^2 \right] &= E \left[\sum_{i=1}^n (X_i^2 - 2\bar{X} X_i + \bar{X}^2) \right] \\ &= E \left[\sum_{i=1}^n X_i^2 - \sum_{i=1}^n 2X_i \bar{X} + \sum_{i=1}^n \bar{X}^2 \right] \\ &= E \left[\sum_{i=1}^n X_i^2 \right] - 2E \left[\sum_{i=1}^n X_i \bar{X} \right] + E \left[\sum_{i=1}^n \bar{X}^2 \right] \checkmark \end{aligned}$$

Question 2 (10 points)

Let X_1, X_2, \dots, X_n be independent, identically distributed random variables with mean μ and variance σ^2 .

In terms of μ and σ^2 , what is the value of $E[X_i^2]$? Note that $Var[X_i] = E[X_i^2] - E[X_i]^2$, while $Var[X_i] = \sigma^2$ and $E[X_i] = \mu$. Please justify your answer.

Confirm numerically that your answer is correct for $X_i \sim \text{exponential}(\text{rate} = \frac{1}{3})$ which has mean equal to 3 and variance equal to 9.

$$\begin{aligned} \text{Var}[X_i] &= E[X_i^2] - E[X_i]^2 \\ E[X_i^2] &= \text{Var}[X_i] + E[X_i]^2 \\ E[X_i^2] &= \sigma^2 + \mu^2 \end{aligned}$$

If we have random variables $X_i \sim \text{exp}(1/3)$, then we can expect $E[X_i^2] = \sigma^2 + \mu^2 = 9 + (3)^2 = 18$. Let's check if our answer is correct.

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f2<-function(x){x^2*dexp(x,1/3)}
integrate(f2,0,Inf)
```

18 with absolute error < 0.00054

Question 3 (5 points)

Assuming that $E[X_i^2] = \sigma^2 + \mu^2$ for all i , what is $E[\sum_{i=1}^n X_i^2]$. Recall that $E[\sum_{i=1}^n Y_i] = \sum_{i=1}^n E[Y_i]$ for any random variables Y_1, Y_2, \dots, Y_n with defined means.

$$\begin{aligned} E\left[\sum_{i=1}^n X_i^2\right] &= \sum_{i=1}^n E[X_i^2] \\ &= \sum_{i=1}^n (\sigma^2 + \mu^2) \\ &= n(\sigma^2 + \mu^2) \end{aligned}$$

Question 4 (5 points)

Define the random variable \bar{X} by $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$. What is the value of $E[\sum_{i=1}^n \bar{X}^2]$? Please justify your answer.

Recall that the mean of \bar{X} equals μ and the variance equals $\frac{\sigma^2}{n}$. The fact that $E[\sum_{i=1}^n Y_i] = \sum_{i=1}^n E[Y_i]$ mentioned above may also be useful. Further, \bar{X} is constant with respect to the index i in the sum.

$$\begin{aligned} E\left[\sum_{i=1}^n \bar{X}^2\right] &= \sum_{i=1}^n E[\bar{X}^2] \\ &= \sum_{i=1}^n \left(\text{Var}[\bar{X}] + E[\bar{X}]^2\right) \\ &= \sum_{i=1}^n \left(\frac{\sigma^2}{n} + \mu^2\right) \\ &= n\left(\frac{\sigma^2}{n} + \mu^2\right) \end{aligned}$$

Question 5 (10 points)

Why is

$$E \left[\sum_{i=1}^n \bar{X} X_i \right] = E \left[\bar{X} \sum_{i=1}^n X_i \right] = E [n\bar{X}^2]$$

Because \bar{X} is a constant with respect to the index i , we can pull it out of the summation. Then recall $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \rightarrow \sum_{i=1}^n X_i = n\bar{X}$. If we can substitute that into our equation, then we get the resulting $E[n\bar{X}^2]$.

Question 6 (5 points)

Assuming that $E [\sum_{i=1}^n X_i^2] = n(\sigma^2 + \mu^2)$, that $E [\sum_{i=1}^n \bar{X}^2] = n(\frac{\sigma^2}{n} + \mu^2)$, and that $E [\sum_{i=1}^n \bar{X} X_i] = E [n\bar{X}^2] = n(\frac{\sigma^2}{n} + \mu^2)$, please simplify $E [\sum_{i=1}^n X_i^2] - 2E [\sum_{i=1}^n \bar{X} X_i] + E [\sum_{i=1}^n \bar{X}^2]$.

$$\begin{aligned} E \left[\sum_{i=1}^n X_i^2 \right] - 2E \left[\sum_{i=1}^n \bar{X} X_i \right] + E \left[\sum_{i=1}^n \bar{X}^2 \right] &= n(\sigma^2 + \mu^2) - 2n(\frac{\sigma^2}{n} + \mu^2) + n(\frac{\sigma^2}{n} + \mu^2) \\ &= n\sigma^2 - 2\sigma^2 + \sigma^2 + n\mu^2 - 2n\mu^2 + n\mu^2 \\ &= (n-1)\sigma^2 \end{aligned}$$

Question 7 (5 points)

If $E [\sum_{i=1}^n X_i^2] - 2E [\sum_{i=1}^n \bar{X} X_i] + E [\sum_{i=1}^n \bar{X}^2] = (n-1)\sigma^2$, what is the value of $E \left[\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \right]$?

$$E \left[\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \right] = \frac{1}{n-1} [(n-1)\sigma^2] = \sigma^2$$