IS 606: Statistics and Probability for Data Analytics Hands-On Laboratory Series The Poisson Distribution: Fundamentals

Overview

This exercise is designed to give you practice in working with the Poisson distribution.

Prerequisites

You should know the basic concepts of the Poisson distribution, including the essential characteristics, the core formulas that go with the distribution, and how to apply the distribution to answer basic questions.

Materials

This lab exercise is entirely self-contained.

Instructions

This lab exercise is to be completed step by step according to the instructions given. If you are struggling with a particular step, then our recommendation is that you look to the solution *for only that step* for help. Once you have sorted out the details of the step in question, proceed to the next task.

- 1. For most of this lab exercise, we will be interested in a Poisson distribution with parameter λ =2.2. For review purposes, describe the basics of a Poisson distribution.
- 2. The Poisson distribution has outcomes that range from 0 to infinity (though the probabilities approach 0 very quickly as X increases far beyond the parameter value). It is therefore impossible to calculate the entire probability distribution in table form. Let's get the first several values, though. Calculate the probability distribution for X ranging from 0 to 8 using the formula:

$$P(X=k) = \frac{e^{-\lambda}\lambda^k}{k!}$$

3. Using your results from the previous step (and ignoring the possibility of X being greater than 8 for now), calculate the expected value of the distribution using the theoretical formula:

$$E(X) = \sum_{i=1}^{n} x_i p(x_i)$$

1

Lab: Poisson Distribution Fundamentals

4. Calculate the expected value of the distribution using the Poisson-specific formula:

$$E(X) = \lambda$$

(Yes, this is so easy I cannot believe we are asking you to do it. How does your answer here compare with your answer from the previous part? What accounts for any differences? Is it just rounding error? Or does excluding outcomes greater than 8 have a significant effect?)

5. Using your distribution from step 2 (and again ignoring outcomes greater than 8), calculate the variance and standard deviation of the distribution using the theoretical formulas:

$$Var(X) = \sum_{i=1}^{n} (x_i - E(X))^2 p(x_i)$$
 and $SD(X) = \sqrt{Var(X)}$

6. Calculate the standard deviation of the distribution using the Poisson-specific formula:

$$SD(X) = \sqrt{\lambda}$$

That's almost as easy as the expected value, right? How does your answer here compare with your answer from the previous part? What accounts for any differences?

- 7. Generate a random sample of 2,000 observations from a Poisson distribution with λ =2.2 using the rpois() function. Be sure to set a random seed with set.seed() and assign the resulting observations to a vector named **poissonsample**.
- 8. Create a histogram of the simulated data (using breaks=c(-0.5, 0.5, 1.5, 2.5, 3.5, 4.5, 5.5, 6.5, ...) so that the breaks extend past your largest observation) and compare it with the probability distribution in step 2. How closely does it match?
- 9. Using the sample you drew in step 7 (poissonsample), calculate the simulated mean using the mean() function. How close is it to the theoretical value?
- 10. Using the sample you drew in step 7 (poissonsample), calculate the simulated standard deviation using the sd() function. How close is it to the theoretical value?
- 11. Construct the cumulative distribution function from the probability distribution.
- 12. Obtain the five-number summary from the cumulative distribution you just constructed. (For max, just put ∞.)
- 13. Obtain the same five-number summary from your simulated data using the quantile() function. How closely does it match the theoretical results from the previous step?

Summary

The exercise above walks you through the basics of the Poisson distribution. In an applications lab, we'll apply these and similar concepts to answer real questions and model real scenarios.

Summary of Useful R Techniques hist() rpois() set.seed() mean() var() sd()

quantile()