Statistical Inference Course Project (Simulation_runs)

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Overview

In this project we will investigate the exponential distribution in R and compare it with the Central Limit Theorem. The exponential distribution can be simulated in R with rexp(n, lambda) where lambda is the rate parameter. The mean of exponential distribution is 1/lambda and the standard deviation is also 1/lambda. The lambda = 0.2 for all of the simulations as mentioned in Coursera course porject. I will investigate the distribution of averages of 40 exponentials. Note: However I will be performing 1000 simulation runs.

Simulations

load neccesary libraries for plotting

library(ggplot2)

set constants as mentioned

lambda <- 0.2 # lambda for rexp

n <- 40 # number of exponetials

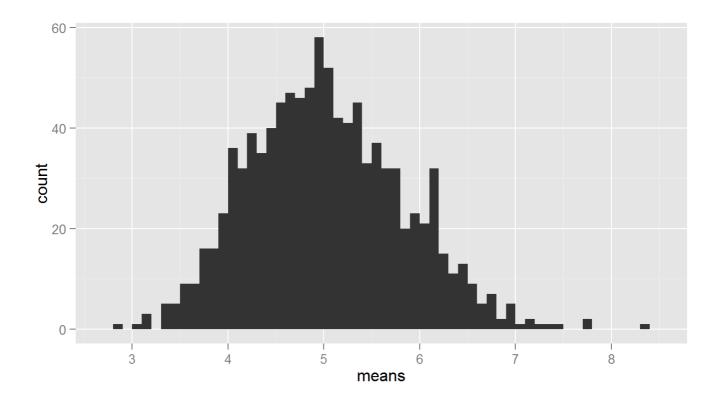
numberOfSimulations <- 1000 # number of tests

set the seed to create reproducability

set.seed(11081979)

run the test resulting in n x numberOfSimulations matrix

 $exponential Distributions <- \ matrix (data=rexp(n * number Of Simulations, lambda), \ nrow=number Of Simulations) \\ exponential Distribution Means <- \ data.frame(means=apply(exponential Distributions, 1, mean)) \\$



Sample Mean versus Theoretical Mean

The expected mean μ of a exponential distribution of rate λ is

$$\mu = \frac{1}{\lambda}$$

```
mu <- 1/lambda
mu
```

[1] 5

Let \bar{X} be the average sample mean of 1000 simulations of 40 randomly sampled exponential distributions.

```
meanOfMeans <- mean(exponentialDistributionMeans$means)
meanOfMeans
```

[1] 5.027126

As you can see the expected mean and the avarage sample mean are very close

Sample Variance versus Theoretical Variance

The expected standard deviation σ of a exponential distribution of rate λ is

$$\sigma = \sqrt[\frac{1/\lambda}{\sqrt{n}}]{}$$

The e

```
sd <- 1/lambda/sqrt(n)
sd
```

[1] 0.7905694

The variance Var of standard deviation σ is

 $Var = \sigma^2$

Var <- sd^2 Var

[1] 0.625

Let Var_X be the variance of the average sample mean of 1000 simulations of 40 randomly sampled exponential distribution, and σ_X the corresponding standard deviation.

$$\label{eq:sd_x loss} \begin{split} & sd_x < \text{-} sd(exponentialDistributionMeans} \\ & sd_x \end{split}$$

[1] 0.8020334

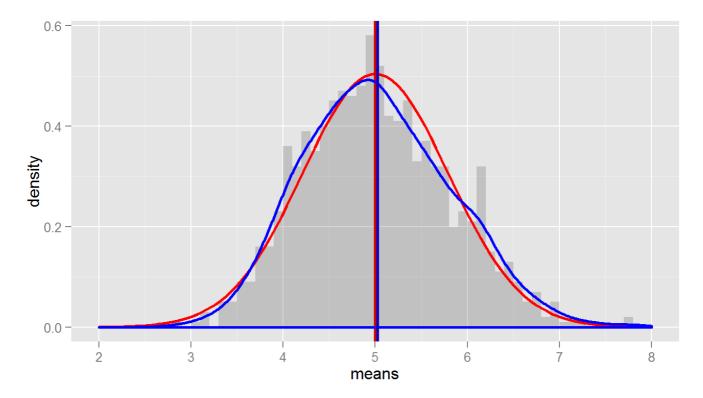
 $Var_x <- var(exponential Distribution Means \$ means) \\ Var_x$

[1] 0.6432577

As you can see the standard deviations are very close Since variance is the square of the standard deviations, minor differences will we enhanced, but are still pretty close.

Distribution

Comparing the population means & standard deviation with a normal distribution of the expected values. Added lines for the calculated and expected means



As you can see from the graph, the calculated distribution of means of random sampled exponantial distributions, overlaps quite nice with the normal distribution with the expected values based on the given

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