# Investigating Exponential Distribution

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## Overview

In this project we will investigate the exponential distribution in R and compare it with the Central Limit Theorem. We will see how much theoritical **mean** and **standard deviation** of exponential distribution differ from the value obtain from **simulation**.

In this project we will assume Lambda = 0.2

### Compairing theoritical statistics with simulated statistics

The theoritical mean is 1/lambda which is 1/0.2=5. The simulation mean is found by **replicate()** command in which the second argument is repeated by the amount given in first argument.

## [1] 5.035787

As the simulated mean is much closer to the theoritical mean of 5. it shows that Central limit theorum has been correctly applied. From the below figure it can also be seen that the mean is the centre of distribution.

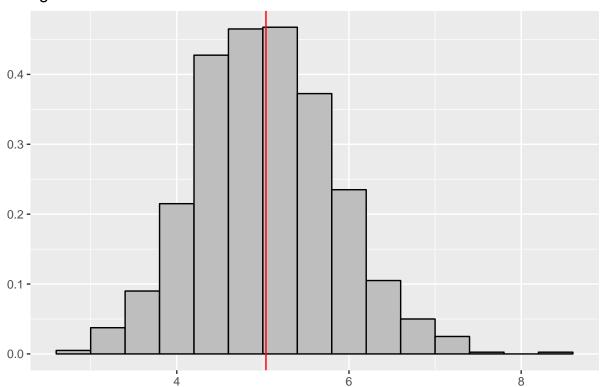


fig 1:Distribution of mean

The theoritical variance of sampling distribution is given by variance/n. Its value is

means

## ((1/0.2)/sqrt(40))<sup>2</sup>

#### ## [1] 0.625

the variance from the sampling distribution is

#### ## [1] 0.6455312

From the table we can easily see how close the simulated variance is to the theoritical variance

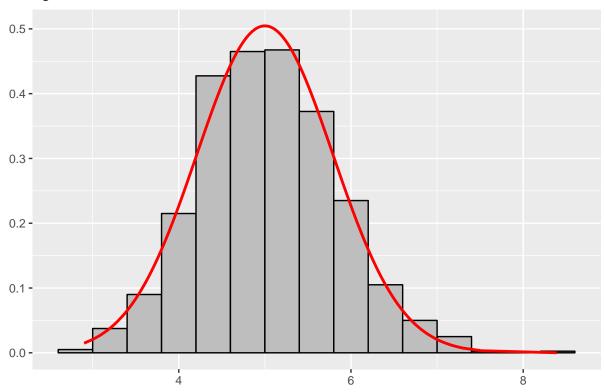
Table 1: Comparison of theoritical and simulated variance

theoritical_variance	simulated_variance
0.625	0.6455312

## Simulated Exponential Distribution is Normal

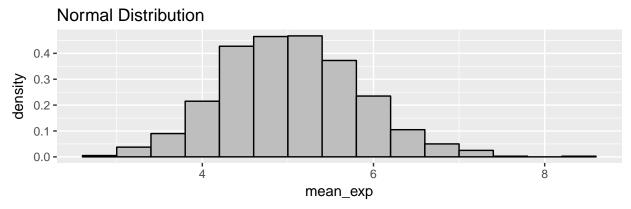
As one can see from the graph below, since the distribution can be approximated by the shape of a bell curve, it is safe to say that the distribution from simulation is normal.

fig 2:Resemblance with bell curve

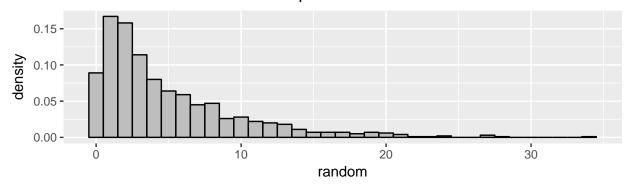


Also if we compare the above distribution with the distribution of 1000 exponential variable, we will observe that the distribution of 1000 exponential values is right skewed.

fig 3:Comparison



## Distribution of 1000 random exponentials



## **APPENDIX**

## 1-Library used are ggplot2,gridExtra,knitr

#### 2-Code for the Table-1

 $\label{lem:dfvar} $$ dfvar<-data.frame(theoritical\_variance=((1/0.2)/sqrt(40))^2,simulated\_variance=var(mean\_exp))$ kable(dfvar,caption = "Comparison of theoritical and simulated variance")$ 

## 3-Code for fig 1,2,3

## fig 1

 $p < -ggplot(data = data.frame(mean = mean\_exp), aes(mean\_exp, ..density..)) + geom\_histogram(fill = "grey", color = "black", binwidth = 0.4) + ggtitle("Normal Distribution") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + xlab("means") + ylab("") \\ p + geom\_vline(xintercept = mean(mean\_exp), col = "red") + ggtitle("fig 1:Distribution of mean") + ylab("") + ggtitle("fig 1:Distribution of mean") + ylab("") +$ 

### fig 2

 $\label{eq:continuous_continuous$ 

#### fig 3

y < -rexp(1000, 0.2)

 $z < -ggplot(data = data.frame(random = y), aes(random, ...density..)) + geom\_histogram(fill = "grey", color = "black", binwidth = 1) + ggtitle("Distribution of 1000 random exponentials") \\ grid.arrange(p,z,nrow = 2, top = "fig 3: Comparison")$