# Statistical Inference Project Part 1 - A Simulation Exercise

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## Overview:

This project investigates the Exponential distribution in R and compares it with the Central Limit Theorem. The mean of the Exponential distribution is  $\frac{1}{\lambda}$  and the standard deviation is also  $\frac{1}{\lambda}$ . A thousand simulations of the distribution of 40 exponentials would be investigated.

## Simulations:

The exponential distribution can be simulated in R with rexp(n, lambda), where lambda is the rate parameter and n is the number of observations. For the purpose of all the simulations in this project, value of lambda is set to 0.2.

First we load the ggplot2 plotting library.

```
library(ggplot2)
```

We then initialize the simulation controlling variables.

```
noSim <- 1000
sampSize <- 40
lambda <- 0.2
```

Set the seed of the Random Number Generator, so that the analysis is reproducible.

```
set.seed(3)
```

Create a matrix with thousand rows corresponding to 1000 simulations and forty columns corresponding to each of 40 random simulations.

```
simulationMatrix <- matrix(rexp(n = noSim * sampSize, rate = lambda), noSim, sampSize)</pre>
```

Create a vector of thousand rows containing the mean of each row of the simulationMatrix.

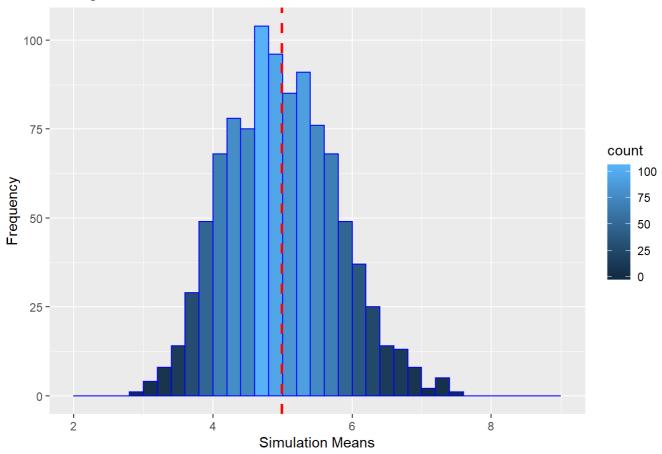
```
simulationMean <- rowMeans(simulationMatrix)
```

Create a data frame containing the whole data.

```
simulationData <- data.frame(cbind(simulationMatrix, simulationMean))</pre>
```

We plot the simulation data to visualize it.





# Sample Mean Versus Theoretical Mean:

The actual mean of the simulated mean sample data is 4.9866197, calculated by:

```
actualMean <- mean(simulationMean)
```

And the theoretical mean is 5, calculated by:

```
theoreticalMean <- (1 / lambda)
```

Thus, we can see that the actual mean of the simulated mean sample data is very close to the theoretical mean of original data distribution.

# Sample Variance Versus Theoretical Variance:

The actual variance of the simulated mean sample data is 0.6257575, calculated by:

```
actualVariance <- var(simulationMean)
```

And the theoretical variance is 0.625, calculated by:

```
theoreticalVariance <- ((1 / lambda) ^ 2) / sampSize
```

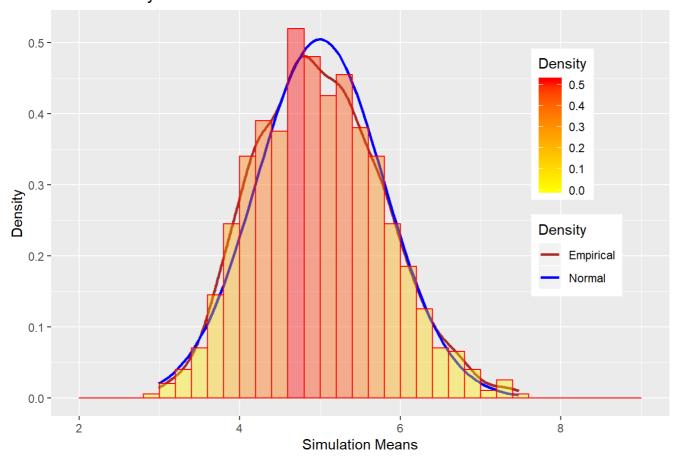
Thus, we can see that the actual variance of the simulated mean sample data is very close to the theoretical variance of original data distribution.

## Distribution:

To prove that the simulated mean sample data approximately follows the Normal distribution, we perform the following three steps:

Step 1: Create an approximate normal distribution and see how the sample data alligns with it.

#### Mean Density Distribution



From above histogram, the simulated mean sample data can be adequately approximated with the normal distribution.

Step 2: Compare the 95% confidence intervals of the simulated mean sample data and the theoretical normally distributed data.

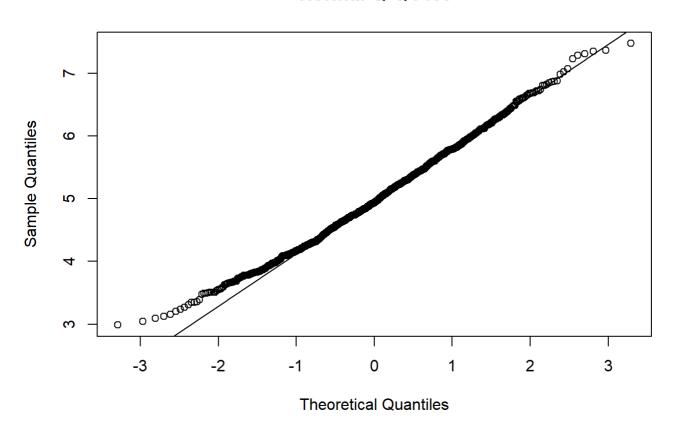
```
actualConfInterval <- actualMean+c(-1,1)*1.96*sqrt(actualVariance)/sqrt(sampSize)
theoreticalConfInterval <- theoreticalMean+c(-1,1)*1.96*
sqrt(theoreticalVariance)/sqrt(sampSize)</pre>
```

Actual 95% confidence interval is [4.7414712, 5.2317681] and Theoretical 95% confidence interval is [4.755, 5.245] and we see that both of them are approximately same.

#### Step 3: q-q Plot for Qunatiles.

```
qqnorm(simulationMean)
qqline(simulationMean)
```

### **Normal Q-Q Plot**



The actual quantiles also closely match the theoretical quantiles, hence the above three steps prove that the distribution is approximately normal.