

# Statistical Interference Course Project. Part 1

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## Overview

This project involves investigating the properties of the exponential distribution in R and comparing it with the Central Limit Theorem. The exponential distribution can be simulated in R using the `rexp(n, lambda)` function, where `lambda` is the rate parameter. For this project, `lambda` will be set at 0.2 for all simulations.

The aim is to investigate the distribution of averages of 40 exponentials by conducting a thousand simulations. Specifically, the project demonstrates a difference between sample characteristics, such as mean and variance, and theoretical values of it.

Finally, the normality of the distribution will be tested using appropriate methods such as a histogram or QQ plot to demonstrate that the distribution is approximately normal.

## Sample simulation

First, for having a stable random sample we will set the seed:

```
set.seed(5327)
```

Then we need to set some values needed for our project

```
lambda <- 0.2  
n <- 40  
size <- 1000
```

And finally we can set our sample by producing random values:

```
simulated_sample = NULL  
for (i in 1 : size) simulated_sample = c(simulated_sample, mean(rexp(n, lambda)))
```

## Sample mean and Theoretical mean comparisson

Calculating mean from the simulations

```
sample_mean <- mean(simulated_sample)  
sample_mean
```

```
## [1] 4.97759
```

Calculating theoretical mean using formula

```
theor_mean <- lambda^(-1)  
theor_mean
```

```
## [1] 5
```

Comparing of this two values by calculating their difference

```
mean_dif <- abs(sample_mean - theor_mean)
mean_dif
```

```
## [1] 0.02241024
```

From this calculations, it is clear that our result has only slight difference from a theoretical value.

## Sample variance and Theoretical variance comparisson

Calculating variance from the simulations

```
sample_var <- var(simulated_sample)
sample_var
```

```
## [1] 0.6452081
```

Calculating theoretical variance using formula

```
theor_var <- 1 / (lambda^2 * n)
theor_var
```

```
## [1] 0.625
```

Comparing of this two values by calculating their difference

```
var_dif <- abs(sample_var - theor_var)
var_dif
```

```
## [1] 0.02020813
```

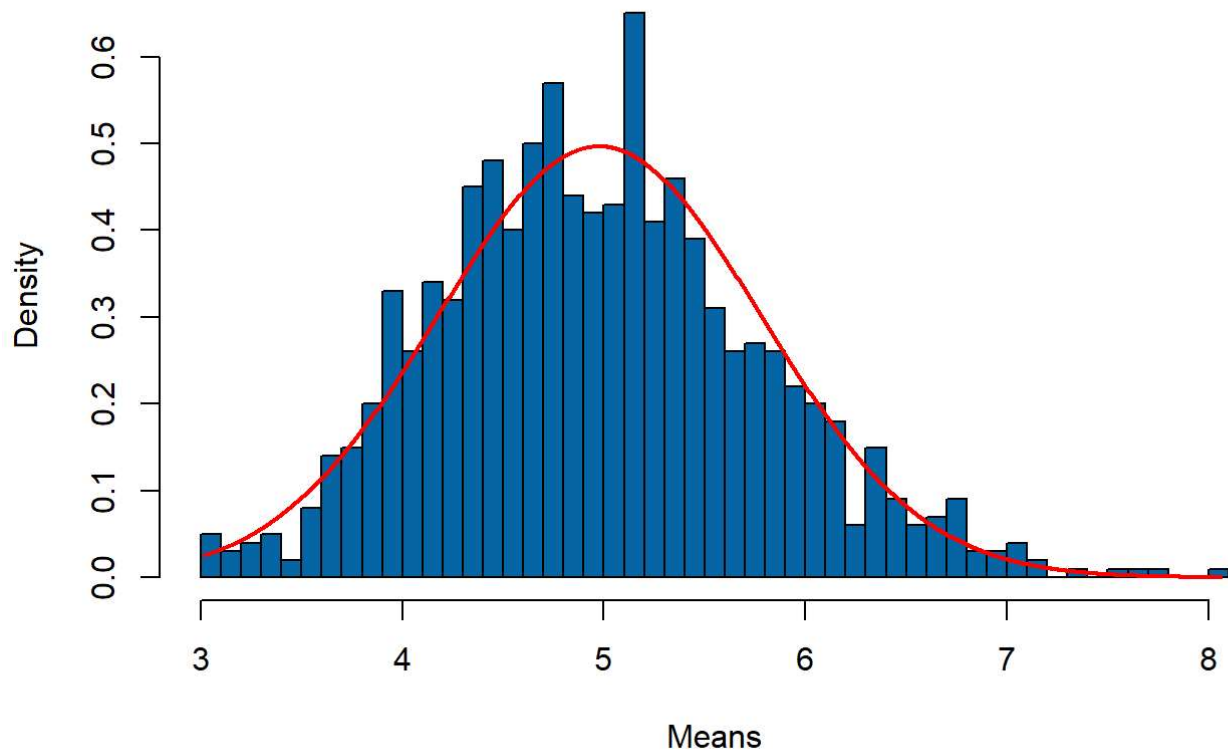
From this calculations, it is clear again that our result has only slight difference from a theoretical value.

## Distribution

By plotting a histogram and a normality Q-Q plot, we can conclude whether the distribution is normal.

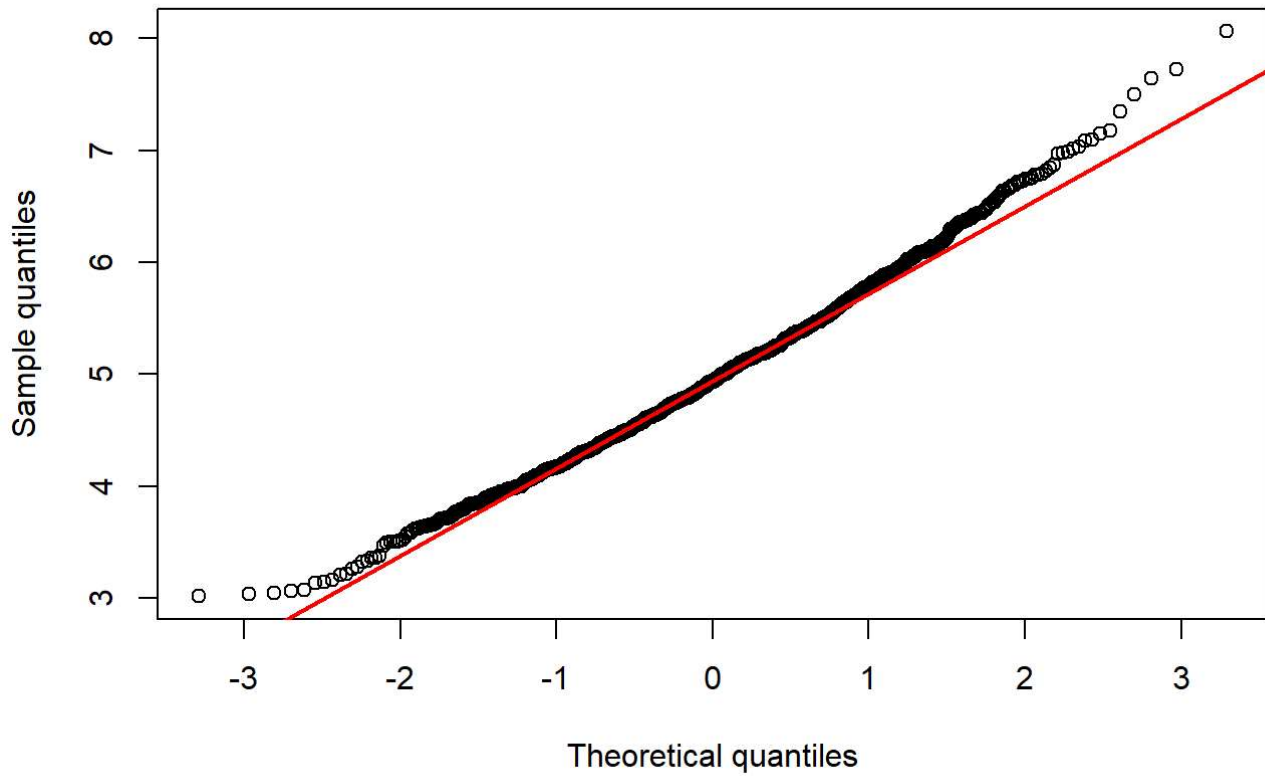
```
xx <- seq(min(simulated_sample), max(simulated_sample), length=1000)
hist(simulated_sample, breaks = n, freq=F, col='#0665a5', xlab='Means', ylab='Density', main='Histogram of means')
lines(xx, dnorm(xx, mean=sample_mean, sd=sd(simulated_sample)), col='red', lwd=2)
```

## Histogram of means



```
qqnorm(simulated_sample, xlab='Theoretical quantiles', ylab='Sample quantiles', main = 'Q-Q plot')  
qqline(simulated_sample, col='red', lwd=2)
```

Q-Q plot



Both plots are showing that the distribution is approximately normal.