

# Statistical Inference Course Project Part 1

TK

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## Synopsis

In this project I 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  $1/\lambda$ . For this project:

- Set `lambda = 0.2` for all of the simulations.
- Investigate the distribution of averages of 40 exponentials.
- We need to do a thousand simulations.
- Illustrate via simulation and associated explanatory text the properties of the distribution of the mean of 40 exponentials.

I should:

- Show the sample mean and compare it to the theoretical mean of the distribution.
- Show how variable the sample is (via variance) and compare it to the theoretical variance of the distribution.
- Show that the distribution is approximately normal.

## Simulation of an Exponential Distribution

I calculated the average of 40 samples drawn from the exponential distribution a thousand times

```
set.seed(1000)

# set lambda
lambda <- 0.2

# samples
n <- 40

# simulations
NSim <- 1000

# simulate
simExp <- replicate(NSim, rexp(n, lambda))
```

## Calculate mean of exponentials

```
meanExp <- apply(simExp, 2, mean)
```

### 1. Show the sample mean and compare it to the theoretical mean of the distribution

```
# Simulated mean  
smean <- mean(meanExp)  
print(smean)
```

```
## [1] 4.986963
```

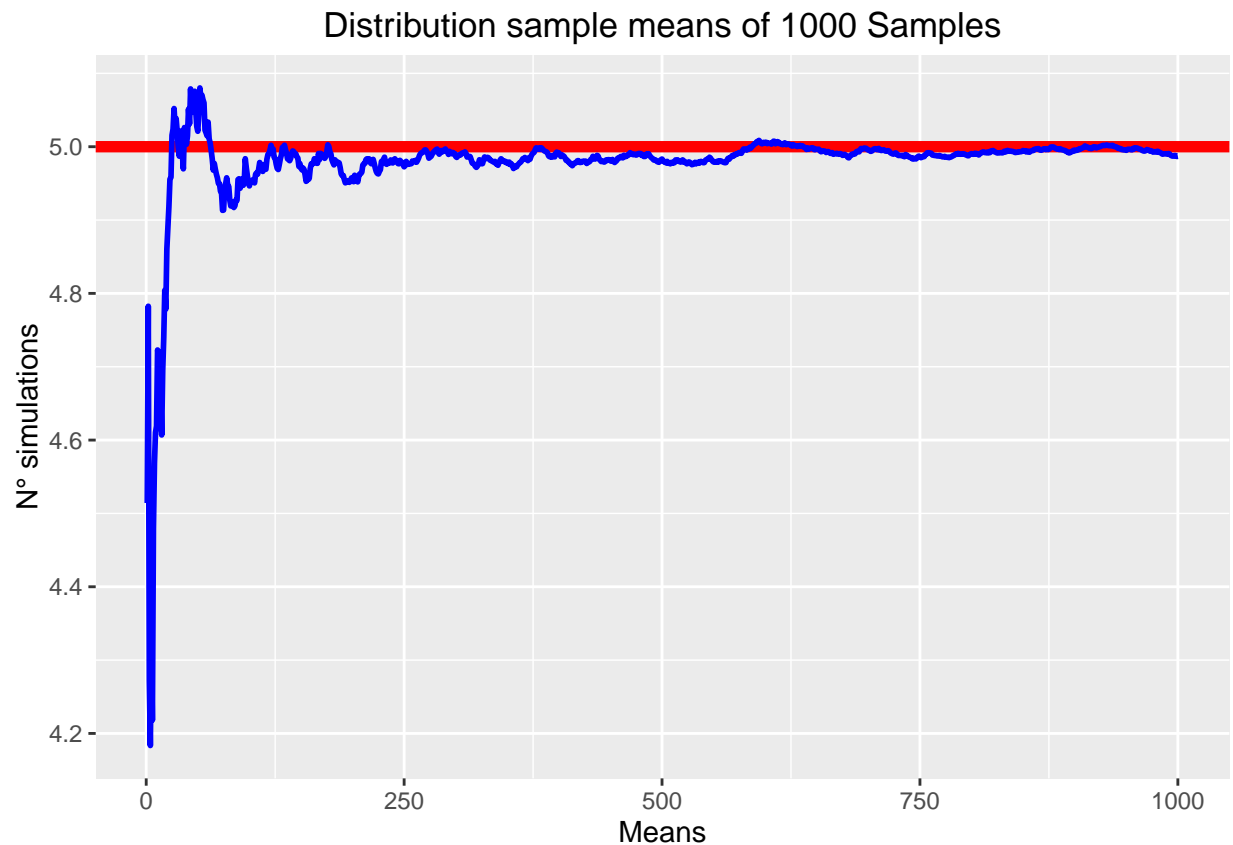
```
# Theoretical mean  
tmean <- 1/lambda  
print(tmean)
```

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

```
means <- cumsum(meanExp)/1:NSim
```

```
# Construction plot
```

```
library(ggplot2)  
g <- ggplot(data.frame(y=means, x=1:NSim), aes(x=x, y=y ))+  
  geom_hline(yintercept = tmean, color = "red", size = 2)+  
  geom_line(size=1, color = "blue")+  
  labs(title="Distribution sample means of 1000 Samples", x = "Means", y="N° simulations")  
print(g)
```



We can see that simulated mean is very close to theoretical mean.

**2. Show how variable the sample is (via variance) and compare it to the theoretical variance of the distribution.**

```
# standard deviation and variance of distribution of averages of 40 exponentials
```

```
sdSim <-sd(meanExp)
print(sdSim)
```

```
## [1] 0.8089147
```

```
varSim <- sdSim^2
print(varSim)
```

```
## [1] 0.654343
```

```
# theoretical standard deviation and variance
```

```
Thsd <-(1/lambda)/sqrt(n)
print(Thsd)
```

```
## [1] 0.7905694
```

```
varTh <- Thsd^2  
print(varTh)
```

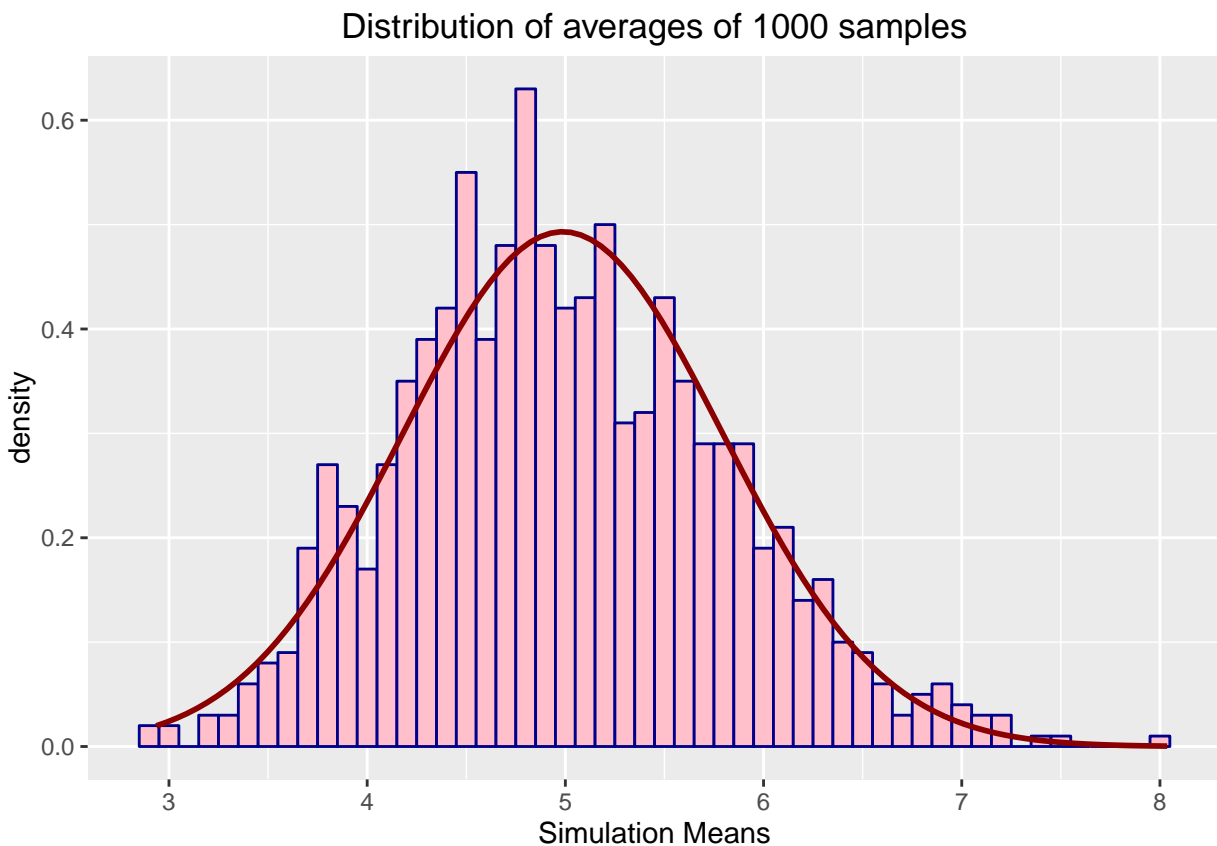
```
## [1] 0.625
```

The simulated standard deviation and variance are closed to theoretical standard deviation and variance.

### 3. Show that the distribution is approximately normal.

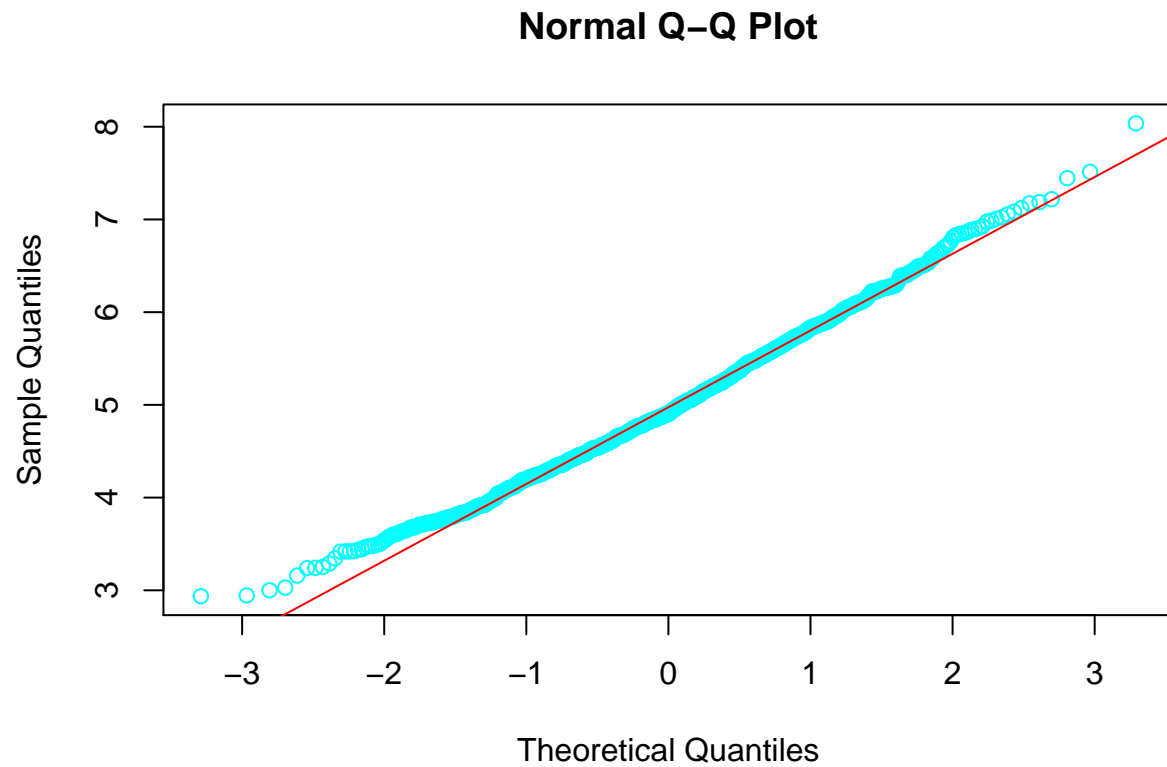
#### 1. Plot of distrubution simulated means.

```
data <- data.frame(meanExp, 1:NSim)  
ggplot(data.frame(y=meanExp), aes(x=y)) +  
  geom_histogram(aes(y=..density..), binwidth=0.1, fill="pink",  
                 color="darkblue") +  
  stat_function(fun=dnorm, args = list(mean=smean,  
                                       sd=sdSim), color = "darkred",  
               size=1) +  
  labs(title="Distribution of averages of 1000 samples", x="Simulation Means")
```



2. Compare our distribution with normal distribution.

```
qqnorm(meanExp, col="5"); qqline(meanExp, col="2")
```



This plots show us that distribution of simulated mean is approximately normal.