

Statistical Inference Course Project - Simulation

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Overview

In this project we'll investigate the distribution of averages of 40 exponentials in R and compare it with the Central Limit Theorem by doing a thousand simulations.

We'll break it into 5 broad sections -

- 1) **Simulations**
- 2) **Sample Mean versus Theoretical Mean**
- 3) **Sample Variance versus Theoretical Variance**
- 4) **Understanding the Distribution**
- 5) **Conclusion**

Let's dive in!!

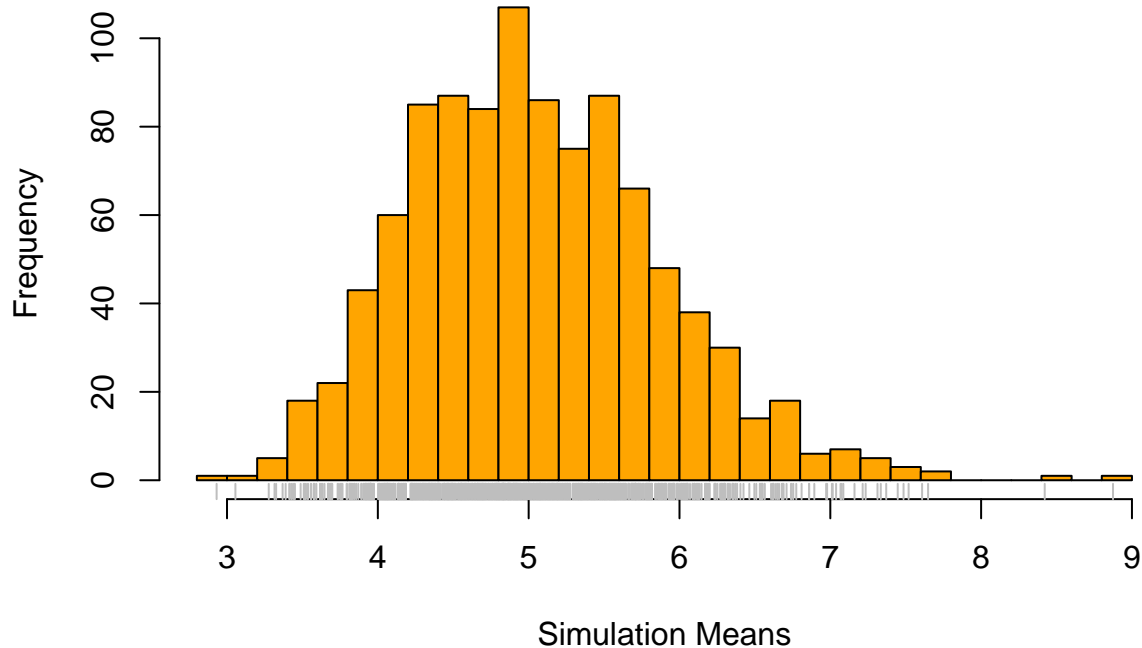
Section 1: Simulations

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 $1/\lambda$. Set `lambda = 0.2` for all of the simulations.

```
# Setting seed for reproducibility
set.seed(7815)
# Setting simulation parameters
n = 40
lambda = 0.2
simulations = 1000
# Generating 1,000 simulations
simulated_exp <- replicate(simulations, rexp(n, lambda))
# calculate columnar mean of exponentials
means_exp <- apply(simulated_exp, 2, mean)

# Visualizing the means using histogram and adding rugs to see finer density scatter
hist(means_exp, col = "orange", xlab = "Simulation Means",
     main = "Frequency plot of Simulated means for 1,000 observations", breaks = 40)
rug(means_exp, col = "grey", lwd = 1)
```

Frequency plot of Simulated means for 1,000 observations



Section 2: Sample Mean versus Theoretical Mean

We know that the theoretical mean of an exponential distribution is $1/\lambda$, which implies for our distribution its 5

```
(sample_mean <- mean(means_exp))      # Sample Mean
```

```
## [1] 5.046637
```

```
(theoretical_mean <- 1/lambda)        # Theoretical Mean
```

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

```
(delta_mean <- abs(sample_mean - theoretical_mean))  # Delta
```

```
## [1] 0.0466371
```

With the delta of 0.046, we can say that The sample mean is very close to the theoretical mean

Section 3: Sample Variance versus Theoretical Variance

```
(sample_var <- var(means_exp))        # Sample variance
```

```
## [1] 0.6888332
```

```
(theoretical_var <- (1/lambda)^2/n)   # Theoretical variance
```

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

```
(delta_var <- abs(sample_var - theoretical_var))    # Delta
```

```
## [1] 0.06383324
```

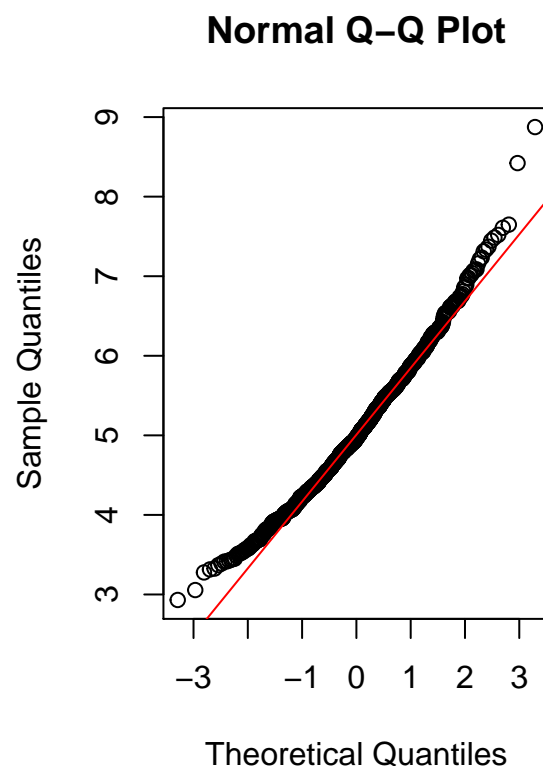
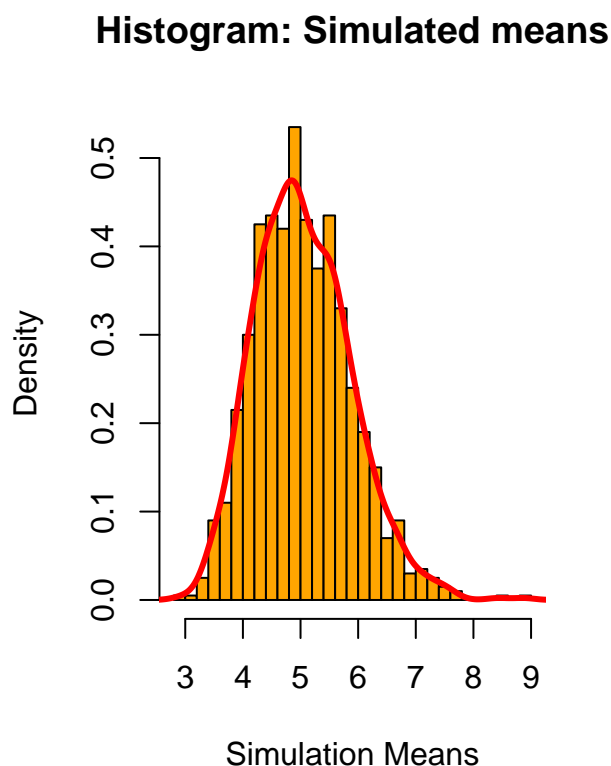
With the delta of 0.064, we can say that the sample variance is very close to the theoretical variance

Section 4: Understanding the Distribution

The exponential distribution is approximately normal. Due to the Central Limit Theorem, the means of samples should also follow a normal distribution. Let's see if this holds true.

```
par(mfrow = c(1,2))
hist(means_exp, probability = TRUE, col = "orange", xlab = "Simulation Means",
     main = "Histogram: Simulated means", breaks = 40)
lines(density(means_exp), lwd=3, col = "red")

qqnorm(means_exp, main="Normal Q-Q Plot", xlab="Theoretical Quantiles", ylab="Sample Quantiles")
qqline(means_exp, col="red")
```



Section 5: Conclusion

Yes, indeed the graphs show the distribution as a close approximation of Gaussian and as we simulate for higher n sizes, the graph will tend to become more closer to the Gaussian Normal!