

# Exponential Distribution vs. Central Limit Theorem

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*November 26, 2018*

## Overview

This project is comparing the exponential distribution to the Central Limit Theorem. We are first comparing the sample mean to the theoretical mean of the distribution. Second, comparing how variable the sample is (via variance) to the theoretical variance of the distribution. Lastly, showing that the distribution is approximately normal.

## Simulations

The below is the given data for comparison purposes:

```
#Number of exponentials
```

```
exps <- 40
```

```
#Lambda
```

```
lbda <- 0.2
```

```
#Simulating the exponential distribution
```

```
experiment <- rexp(exps, lbda)
```

```
#Finding the mean for the experiment
```

```
mean(experiment)
```

```
## [1] 6.05919
```

```
#Set the seed to make sure research is reproducible
```

```
set.seed(12345)
```

```
#Run experiment 1000 times and run histogram
```

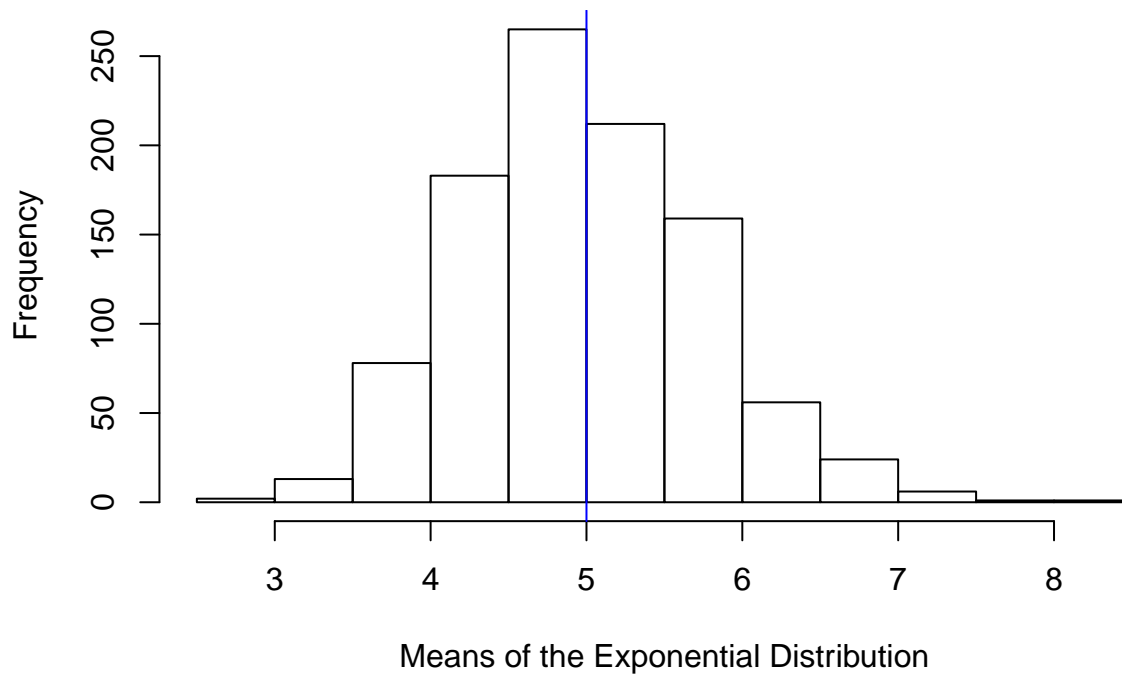
```
experiment1k <- replicate(1000, mean(rexp(exps, lbda)))
```

```
hist(experiment1k, xlab = "Means of the Exponential Distribution", ylab = "Frequency", main = "1000 Mean")
```

```
#Theoretical mean and include in histogram
```

```
abline(v = 1/lbda, col = "blue")
```

## 1000 Means of the Exponential Distribution



```
#Variance vs Theoretical Variance
variance <- sd(experiment1k)^2
variance
```

```
## [1] 0.5954369
```

```
theoVar <- (1/(lbda^2))/exps
theoVar
```

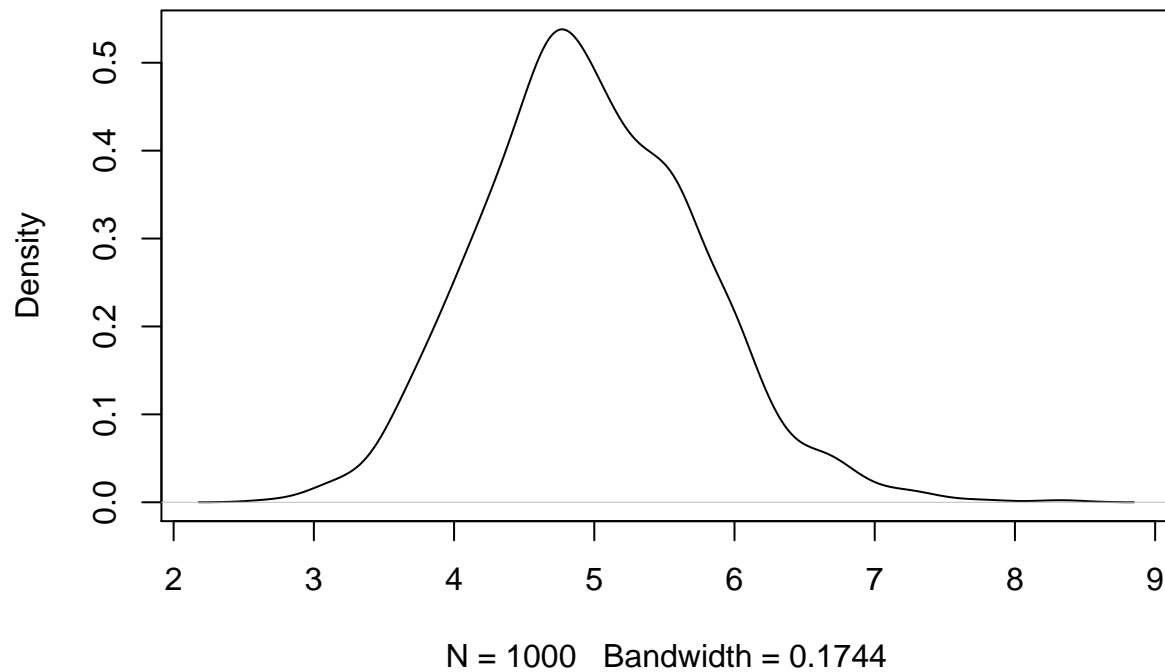
```
## [1] 0.625
```

Both the actual and theoretical variance are close.

## Plot of Experiments

```
plot(density(experiment1k), main = "Density plot of 1000 experiments ran")
```

### Density plot of 1000 experiments ran



The above density plot is showing where the most values are concentrated over the interval, and showing the shape of the distribution is mostly normal. This density plot does allow for smoothing out some of the noise, but with more experiments ran, the smoother and more normal the plot will become.