### Statistical Inference Project 1

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#### Overview

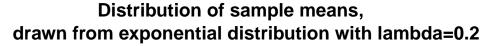
The idea behind this project assignment is to investigate the exponential distribution in R and compare it with the Central Limit Theorem. The exponential distribution can be simulated in R with rexp(n, lambda) where lambda is the rate parameter. The mean of exponential distribution is 1/lambda and the standard deviation is also 1/lambda. We will set lambda = 0.2 for all of the simulations. The distribution of averages of 40 exponentials will be investigated. We will perform 1000 simulations.

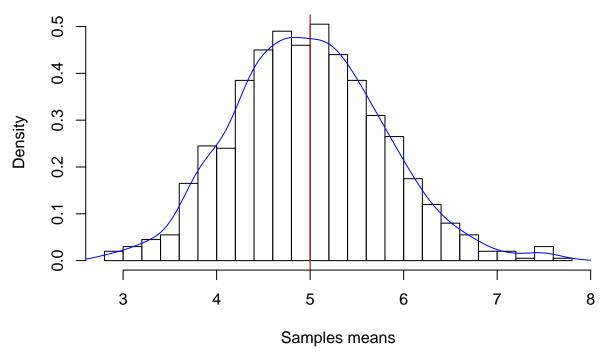
## 1. Show the sample mean and compare it to the theoretical mean of the distribution.

```
# setting constants
lambda <- 0.2
n <- 40 # number of exponetials
numSim<- 1000 # number of simulations

# provide reproducability by setting the seed
set.seed(1808)

# Simulation vs theory
sim <- matrix(data=rexp(n * numSim, lambda),numSim,n)
simMeans <- rowMeans(sim)
meanSim <- mean(simMeans)
sdSim <- sd(simMeans)
varSim<-var(simMeans)
meanTheory <- 1/lambda</pre>
```





So, as can be seen, simulated mean 4.9850712 is rather close to theoretical mean  $\frac{1}{\lambda} = 5$ .

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

Standard deviation for simulated sample means is 0.801991 while the theoretical expected standard deviation  $\sigma$  of a exponential distribution of rate  $\lambda$  is  $\sigma = \frac{1/\lambda}{\sqrt{n}}$ , i.e.:

```
sdTheory <- 1/lambda/ sqrt(n)
sdTheory</pre>
```

#### ## [1] 0.7905694

Variance for simulated sample means is 0.6431895 while the theoretical variance Var of standard deviation  $\sigma$  is  $Var = \sigma^2$ , i.e.:

```
varTheory <- sdTheory^2
varTheory</pre>
```

#### ## [1] 0.625

As can be observed standard deviations are very close and the same goes for the variance.

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

Due to the central limit theorem, the sample means follow the normal distribution, which can be observed already with the first plot. Furthermore the normal q-q plot below inicates that the distribution is quite close to normal.

```
qqnorm(simMeans)
qqline(simMeans, col = 2)
```

### Normal Q-Q Plot

