

StatInference_Project - Part A

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Overview

In this project I will investigate the **exponential distribution** in R and compare it with the **Central Limit Theorem**. The **Central Limit Theorem** in brief describes that the distribution of the means of iid variables tends to be a standard normal distribution. $\gg (\text{Estimate} - \text{Mean of Estimates}) / \text{Std Error of means} \rightarrow \text{Normal}$

Simulations

I will create 1000 random exponential distributions of iid variables, all with rate $\lambda = 0.2$ and # of samples $n=40$. For each one of these I will be calculating and saving their mean.

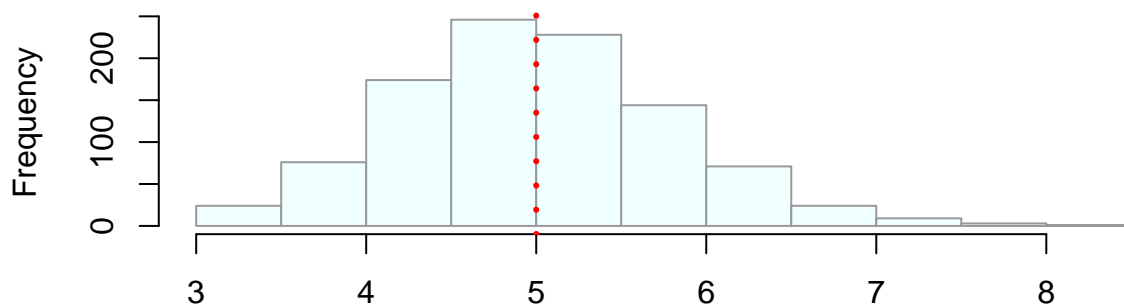
```
#Setting lambda = 0.2 for all of the simulations
lambda <- 0.2
#calculate the mean, the standard deviation and the variance of each exponential distribution
mean<- 1/lambda ; std<- 1/lambda ; var= std^2 ; n<-40
#running 1000 simulations of exponential distributions, all with lambda=0.2, n=40
#for each one, I am calculating each mean (mean(rexp(n, lambda))) and saving it to a table "mns"
ens<- NULL; sens<- NULL; mns<-NULL ; vns<- NULL
for (i in 1 : 1000)
{
  #run simulation of exponential & add to the estimates table
  dist<- rexp(n, lambda) ; ens<-c(ens, dist)
  #mean & variance of n samples simulation calculated and stored in the relevant table, aka mns and vns
  mns <- c(mns, mean(dist)); vns <- c(vns, sd(dist)^2 ) }
# normalized distribution of means
d <- (ens - mns)*sqrt(n)/std
```

Sample Mean versus Theoretical Mean

In the following diagram, I will be demonstrating the distribution of the means of the simulated exponentials vs the theoretical mean, which only depends on the lambda. Since lambda does not change the theoretical mean is also fix $= 1/\lambda \rightarrow \text{Theoretical mean} = 5$

```
#plot the distribution of sample mean
hist(mns, main="Samples Mean Distribution & Theoretical Mean", xlab="samples mean", col= "azure", border= "black")
#plot the theoretical mean, in red
abline(v=mean, col= "red", lty=3, lwd= 3)
```

Samples Mean Distribution & Theoretical Mean



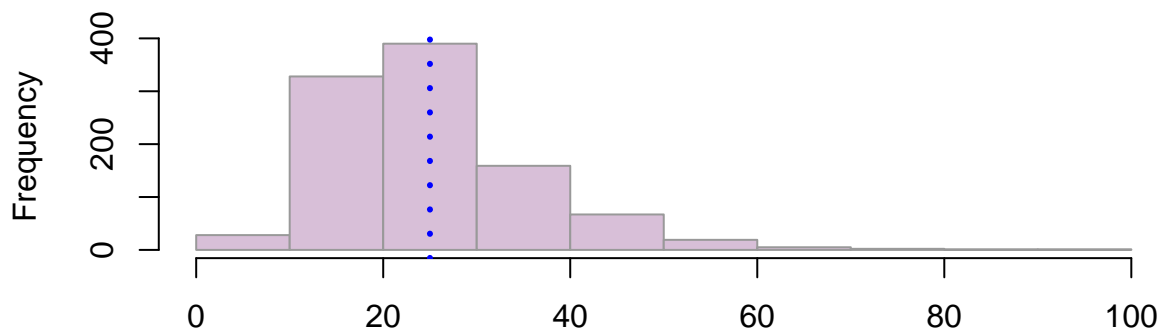
samples mean

##Sam-

ple Variance versus Theoretical Variance In the following diagram, I will be demonstrating the distribution of the variances of the simulated exponentials vs the the theoretical variance, which only depends on the lambda. Since lambda does not change, the theoretical variance is also fixed = $(1/\lambda)^2 \rightarrow$ **Theoretical Variance = 25**

```
#plot the distribution of sample variance
hist(vns, main="Samples Variance Distribution & Theoretical Variance", xlab="samples variance", col= "t")
#plot the theoretical variance in blue
abline(v=var, col= "blue", lty=3, lwd= 3)
```

Samples Variance Distribution & Theoretical Variance



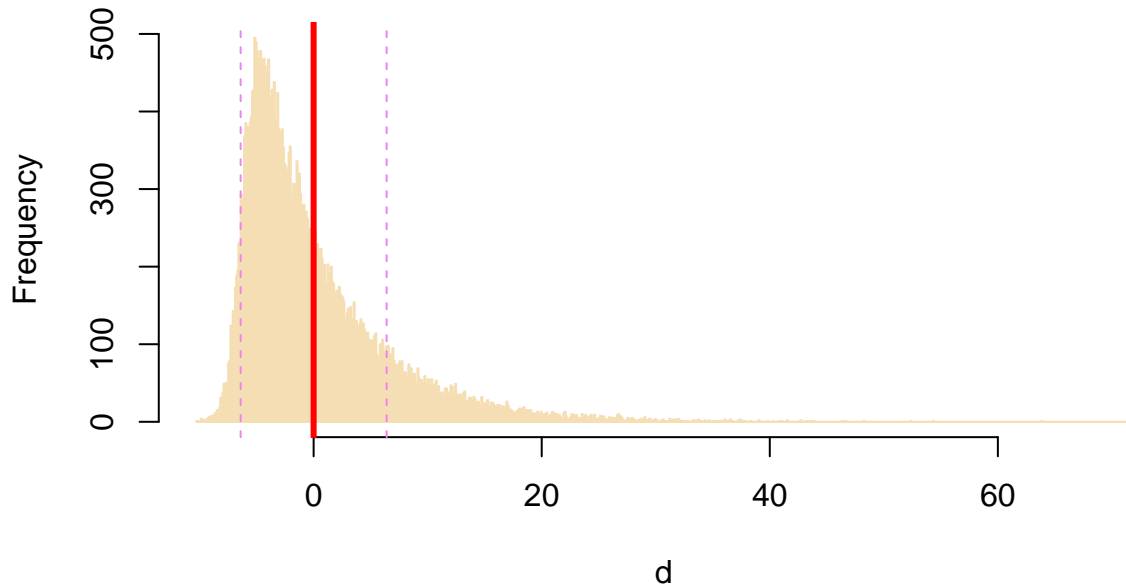
samples variance

#Distri-

bution Via figures and text, explain how one can tell the distribution is approximately normal.

```
hist(d, main="Distribution of means, normalized", col="wheat", border="wheat", breaks=1000)
abline(v=mean(d), col="red", lwd= 3) ; abline(v=sd(d), col="violet", lty=2, lwd= 1)
abline(v=-sd(d), col="violet", lty=2, lwd= 1)
```

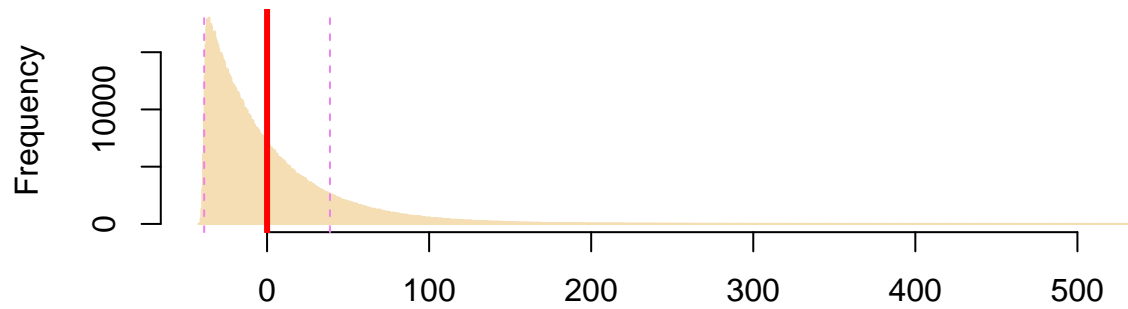
Distribution of means, normalized



Distribution **mean**= 0 tends to equal the mean of standard normal distribution, which is 0. Distribution **st. deviation**= 6.4 tends to equal the st. deviation of standard normal distribution, which is 1. As $n=40$ is relatively small, I will rerun my simulations with a larger n , to see whether this round will result in a better approximation of a standard normal distribution for the simulation means.

```
#set a larger n
n2 <- 1500
#re-running 1000 simulations of exponential distributions, all with lambda=0.2, n=1500
ens2<- NULL; sens2<- NULL; mns2<-NULL ; vns2<- NULL
for (i in 1 : 1000)
{
  #run simulation of exponential & add to the estimates table
  dist2<- rexp(n2, lambda) ; ens2<-c(ens2, dist2)
  #mean & variance of n samples simulation calculated and stored in the relevant table, aka mns2 and vns2
  mns2 <- c(mns2, mean(dist2)) ; vns2 <- c(vns2, sd(dist2)^2 ) }
# normalized distribution of means
d2 <- (ens2 - mns2)*sqrt(n2)/std
#plot new distribution with larger n
hist(d2, main="Distribution of means, n=1500, normalized", col="wheat3", border="wheat", breaks=1000)
abline(v=mean(d2), col="red", lwd= 3); abline(v=sd(d2), col="violet", lty=2, lwd= 1) ; abline(v=-sd(d2), col="violet", lty=2, lwd= 1)
```

Distribution of means, n=1500, normalized



The distribution mean 0 tends to equal the mean of standard normal distribution, which is 0. The distribution standard deviation 38.81 tends to equal the standard deviation of standard normal distribution, which is 1.