

Statistical Inference Project - Part 1

```
lambda=.2

nosim<-1000

mean<-1/lambda

sd<-1/lambda

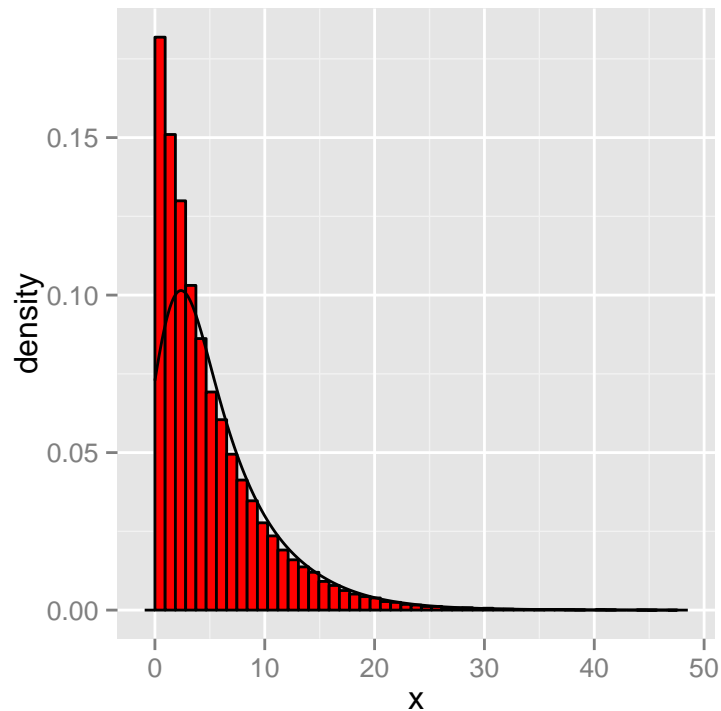
set.seed(1234)

data<-rexp(nosim * 40,lambda)

n1<-data.frame(x=data)
n10<-data.frame(x=apply(matrix(data,nrow = nosim,ncol = 10),1,mean))
n20<-data.frame(x=apply(matrix(data,nrow = nosim,ncol = 20),1,mean))
n25<-data.frame(x=apply(matrix(data,nrow = nosim,ncol = 25),1,mean))
n40<-data.frame(x=apply(matrix(data,nrow = nosim,ncol = 40),1,mean))
```

The exponential density is strongly skewed, we need a large value of n before we see the bell curve emerge.

```
ggplot(n1,aes(x = x)) +
  geom_histogram(binwidth=diff(range(data))/50,fill="red",colour = "black",
    aes(y = ..density..)) +
  stat_density(geom = "line",adjust=5)
```



- 1. Show the sample mean and compare it to the theoretical mean of the distribution.
Both almost the same as we knew for the LLN theorem.

```
#Sample mean  
mean(n40$x)
```

```
## [1] 4.974239
```

```
#Theoretical mean of the distribution  
mean
```

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

- 2. Show how variable the sample is (via variance) and compare it to the theoretical variance of the distribution. For the CLT theorem we knew that Standard deviation of the sample $Sx \sim sd/\sqrt{n}$.
As we will see in the graph the distribution is approximately normal but not completely.

```
#Standar deviation of the sample  
sd(n40$x)
```

```
## [1] 0.7713431
```

```
#The theoretical standard deviation of the distribution  
sd/sqrt(40)
```

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

- 3. Show that the distribution is approximately normal.

The larger is n the more the closer the exponential distribution approximate to the normal. I have use the *stat_density* function that estimates the density of the sample so visually it is easier to see how the shape is moving to a normal distribution.

```
g1 <- ggplot(n10, aes(x = x)) +
  geom_histogram(binwidth=diff(range(n10$x))/50,
    fill="red",colour = "black", aes(y = ..density..)) +
  stat_function(fun=dnorm, arg = list(mean=mean(n10$x)),aes(colour = "Normal")) +
  stat_density(geom = "line",adjust=5, aes(colour = "Sample")) +
  scale_colour_manual("",values = c("green","black")) +
  ggtitle("Averages of 10 samples")

g2 <- ggplot(n20, aes(x = x)) +
  geom_histogram(binwidth=diff(range(n20$x))/50,
    fill="red",colour = "black", aes(y = ..density..)) +
  stat_function(fun=dnorm, arg = list(mean=mean(n20$x)),aes(colour = "Normal")) +
  stat_density(geom = "line",adjust=5, aes(colour = "Sample")) +
  scale_colour_manual("",values = c("green","black")) +
  ggtitle("Averages of 20 samples")

g3 <- ggplot(n25, aes(x = x)) +
  geom_histogram(binwidth=diff(range(n25$x))/50,
    fill="red",colour = "black", aes(y = ..density..)) +
  stat_function(fun=dnorm, arg = list(mean=mean(n25$x)),aes(colour = "Normal")) +
  stat_density(geom = "line",adjust=5, aes(colour = "Sample")) +
  scale_colour_manual("",values = c("green","black")) +
  ggtitle("Averages of 25 samples")

g4 <- ggplot(n40, aes(x = x)) +
  geom_histogram(binwidth=diff(range(n40$x))/50,
    fill="red",colour = "black", aes(y = ..density..)) +
  stat_function(fun=dnorm, arg = list(mean=mean(n40$x)),aes(colour = "Normal")) +
  stat_density(geom = "line",adjust=5, aes(colour = "Sample")) +
  scale_colour_manual("",values = c("green","black")) +
  ggtitle("Averages of 40 samples")

grid.arrange(g1, g2, g3,g4, ncol = 2, nrow=2)
```

