

Simulation of Central Limit Theorem

Dorota

21 June 2015

Overview

By performing simulations in R we will show that average of n exponentially distributed variables is approximately normal with mean $1/\lambda$ and the standard deviation $1/(\lambda\sqrt{n})$. It will visually present the central limit theorem, that states that the distribution of average of a large number of independent, identically distributed variables will be approximately normal, regardless of the underlying distribution.

Simulations:

We will perform 1000 simulation. The variable we simulate is average of 40 variables, exponentially distributed with $\lambda = 0.2$.

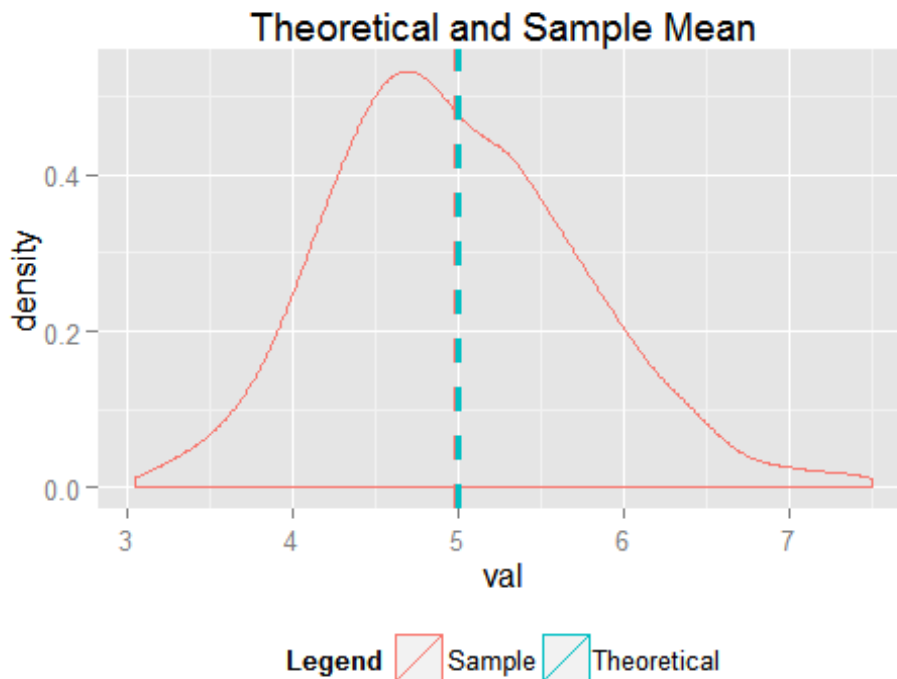
```
set.seed(7)
mns = NULL
lambda=0.2
n=40
for (i in 1 : 1000) mns = c(mns, mean(rexp(n,lambda)))
mns<-as.data.frame(merge('Sample',mns))
colnames(mns)<-c('Legend', 'val')
```

Sample Mean versus Theoretical Mean

Below we can compare sample mean (4.98) vs theoretical mean of exponentially distributed variables ($\text{round}(\text{as.numeric}(1/\lambda), 2)$)

```
# calculate mean
mean=as.data.frame(merge('Sample', as.numeric(mean(mns$val, na.rm=T))))
mean=rbind(mean, merge('Theoretical', as.numeric(1/lambda)))
colnames(mean)<-c('Legend', 'mean_val')

#plot
library(ggplot2)
ggplot(data=mns, aes(x=val, colour=Legend)) +
  geom_density() +
  geom_vline(data=mean, aes(xintercept=mean_val, colour=Legend), linetype="dashed",
    size=1.1) +
  theme(legend.position="bottom") +
  labs(title="Theoretical and Sample Mean")
```

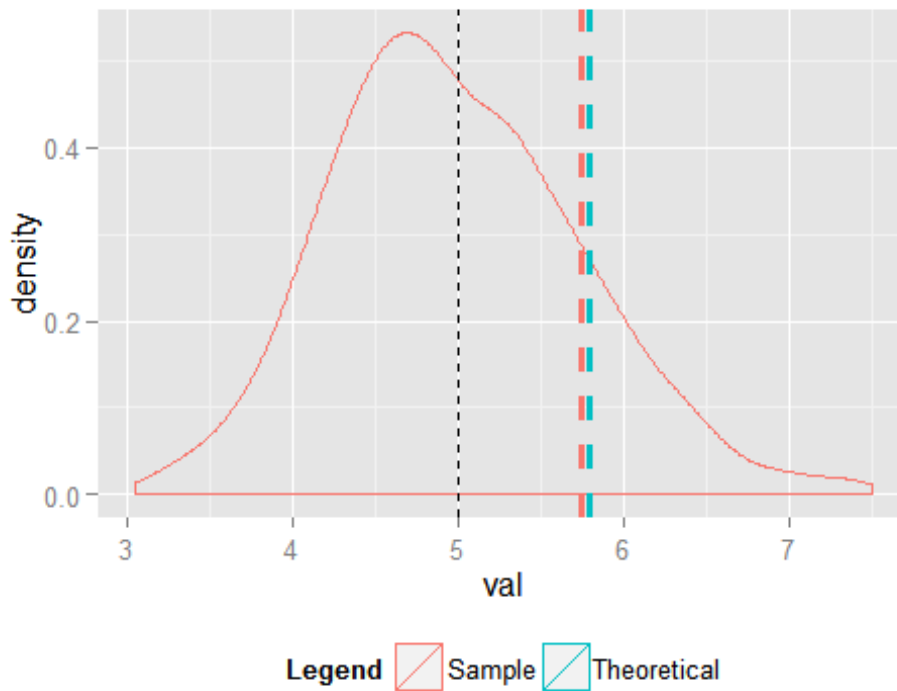


Sample Variance versus Theoretical Variance

Below we can compare sample standard deviation (0.76) vs theoretical standard deviation of exponentially distributed variables ($r = \text{round}(\text{as.numeric}(1/(n \cdot \lambda)), 2)$). Chart below shows mean + one standard deviation.

```
# calculate mean
sdev=as.data.frame(merge('Sample', as.numeric(sd(mns$val, na.rm=T))))
sdev=rbind(sdev, merge('Theoretical', as.numeric( 1/(lambda*sqrt(n)) ))
sdev<-cbind(sdev, as.data.frame(sdev[,2]+mean[,2]))
colnames(sdev)<-c('Legend', 'sdev_val', 'sm_val')

#plot
ggplot(data=mns, aes(x=val, colour=Legend)) +
  geom_density() +
  geom_vline(data=mean[2, ], aes(xintercept=mean_val), colour='black', linetype="dashed",
size=0.25) +
  geom_vline(data=sdev, aes(xintercept=sm_val, colour=Legend), linetype="dashed",
size=1.1) +
  theme(legend.position="bottom")
```



Distribution

Comparing the shapes we can see that the distribution is approximately normal with mean $1/\lambda$ and the standard deviation $1/(\lambda \cdot \sqrt{n})$.

```
#normal distribution
set.seed(7)
rnormal = rnorm(1000, 1/lambda, 1/(sqrt(n)*lambda))
rnormal <- as.data.frame(merge('Normal', rnormal))
colnames(rnormal) <- c('Legend', 'val')
dist <- rbind(mns, rnormal)

#plot
ggplot(data=dist, aes(x=val, colour=Legend)) +
  geom_density() +
  theme(legend.position="bottom") +
  labs(title="Sample and Normal Distribution")
```

