

# Sample Means of an Exponential Distribution

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The exponential distribution represents the distribution of waiting times in a Poisson process with rate of change  $\lambda$ . The probability density function for the exponential distribution is

$$f(x; \lambda) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

The distribution is supported on the interval  $[0, \infty)$ . The mean of an exponential distribution is  $\frac{1}{\lambda}$  and the standard deviation is also  $\frac{1}{\lambda}$ .

In R the exponential distribution is simulated with `rexp(n, lambda)` where `lambda` is the rate parameter.

```
n <- 40
lambda <- 0.2

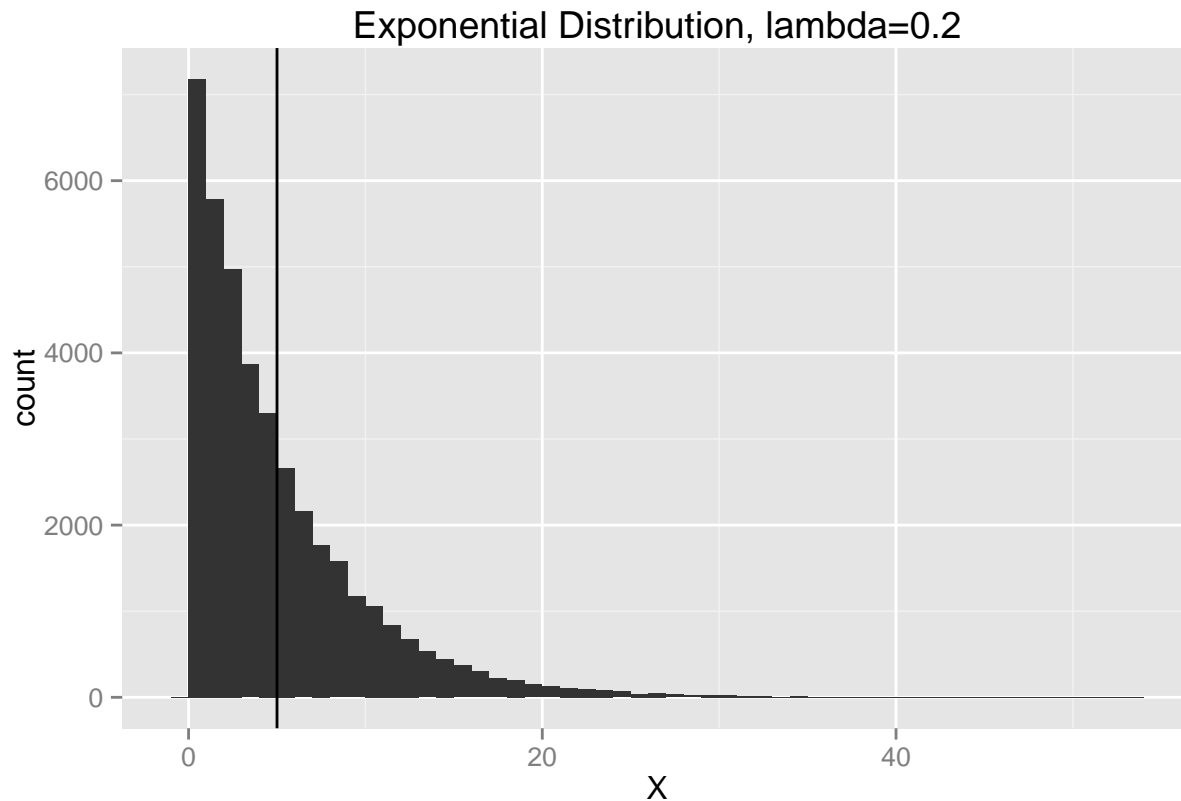
# our expected mean and standard deviation
mu <- 1/lambda
sigma <- 1/lambda
```

We expect to observe a mean  $\mu = 5$  and standard deviation of  $\sigma = 5$  from our simulation.

## Simulation

To get a more intuitive understanding of the exponential distribution, let us first see what it looks like. We ask R to give us 40,000 random values from the distribution, and plot them on a histogram. The distribution mean of 5 is indicated by the vertical line.

```
library("ggplot2")
# 1,000 simulations
num.sims <- 1000
X <- rexp(n*num.sims, lambda)
p <- qplot(X, geom=c("histogram"), binwidth=1, main="Exponential Distribution, lambda=0.2") + geom_vline(aes(x=5))
print(p)
```



### Mean and Standard Deviation

We compute the mean and standard deviation of the random values and compare them to our expected value and standard deviation of  $\frac{1}{\lambda}$ .

```
print(Xbar<-mean(X))
```

```
## [1] 5.052
```

The mean for the 40,000 generated random values is 5.0523.

```
sd(X)
```

```
## [1] 5.037
```

The standard deviation for the random values is 5.0374.

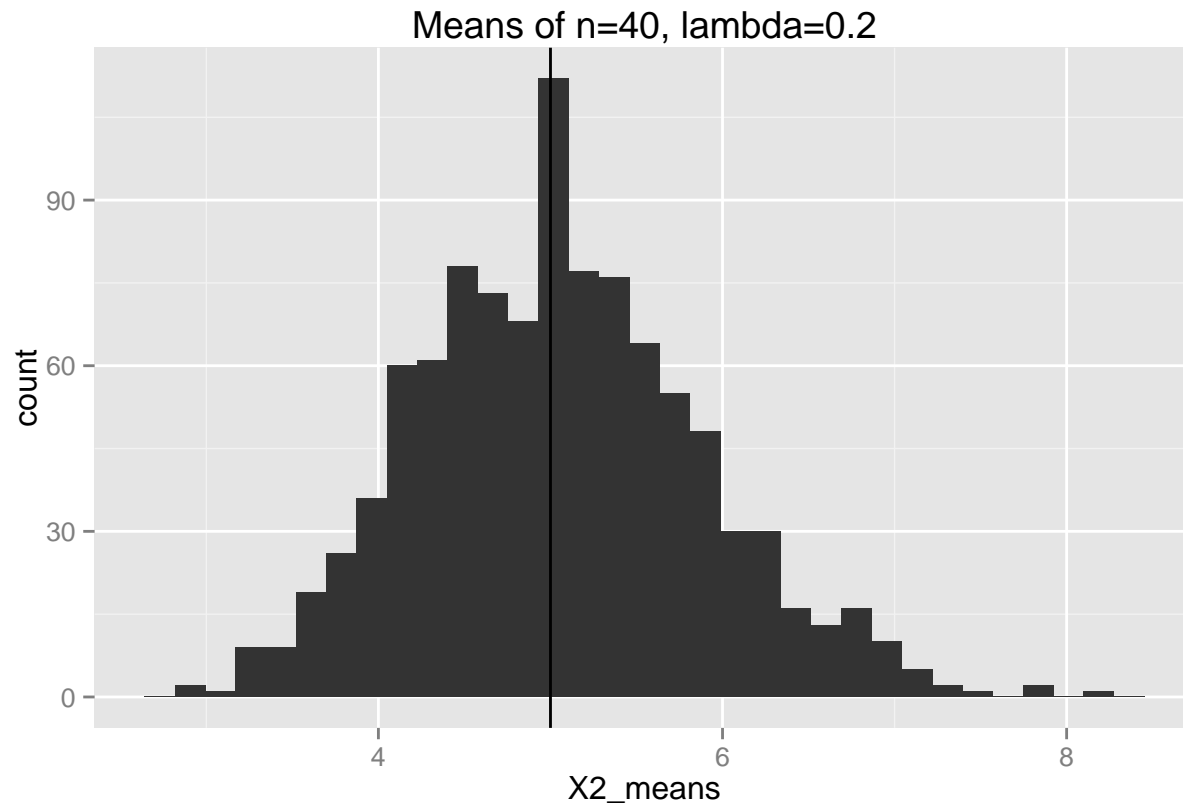
Both the observed mean and standard deviation are close to  $\frac{1}{\lambda}$  which we know to be 5.

### Sample means for n=40.

We divide the 40,000 random values we got earlier into 1000 samples of  $n=40$ , and get the mean of each sample. The expected value is indicated by a vertical line.

```
X2 <- matrix(X,nrow=num.sims)
X2_means <- apply(X2,1,mean)
#hist(X_means,breaks=50, main ="Means of sample size 40")
p<-qplot(X2_means,geom=c("histogram"),
  main="Means of n=40, lambda=0.2") +
  geom_vline(aes(xintercept=1/lambda))
print(p)
```

## stat\_bin: binwidth defaulted to range/30. Use 'binwidth = x' to adjust this.



The distribution of the sample means *looks* normal and is centered around the distribution mean of 5. The mean of the sample means is in fact 5.0523.

### Standard error for n=40.

The standard error of means of size n is  $\frac{\sigma}{\sqrt{n}}$ .

```
std.err = sigma/sqrt(n)
```

We find that for samples of size n=40, standard error is 0.7906. We can compare this to the standard deviation of the sample means.

```
sd(X2_means)
```

```
## [1] 0.823
```

So this confirms our understanding that the standard deviation of the sample means is estimated by the standard error  $\frac{\sigma}{\sqrt{n}}$ .