

# ECON 403A HMWK1

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## Question 1: *The Birthday Problem*

suppose there are  $n=25$  people in a room Assume the following:

1. Ignore leap years and assume there are only 365 days in a year.
  2. Births are uniformly distributed throughout the year.
  3. The people in the room are randomly distributed throughout the year.
- i. What is the probability that two or more of them have the same birthday? Solve this analytically.

$\text{Prob}(\text{at least 2 people have the same birthday}) = 1 - \text{Prob}(\text{no one has same birthday})$

$$\text{Prob}(\text{no one has same birthday}) = \frac{365}{365} \times \frac{364}{365} \times \frac{363}{365} \times \dots \times \frac{341}{365}$$

$$\text{Prob}(\text{no one has same birthday}) = \frac{365!}{340!} \times \frac{1}{365^{25}} \approx 0.4313$$

$\text{Prob}(\text{at least 2 people have the same birthday}) = 1 - \text{Prob}(\text{no one has same birthday})$

$$\text{Prob}(\text{at least 2 people have the same birthday}) \approx 1 - 0.4313 \approx 0.5687 \approx 57\%$$

- ii. Solve [i] computationally in R using the **prod** function. Next, to better understand the relationship between  $n$  and the probability  $p$ , plot  $p$  against  $n$  by looping through rooms of size 1 to 60.

```
n=25                                     #set size of room
y=365^n
x=prod((365-(n-1)):365)/y               #probability that no one has same birthday
p=1-x                                   #probability that at least two people have same birthday
print(p)
```

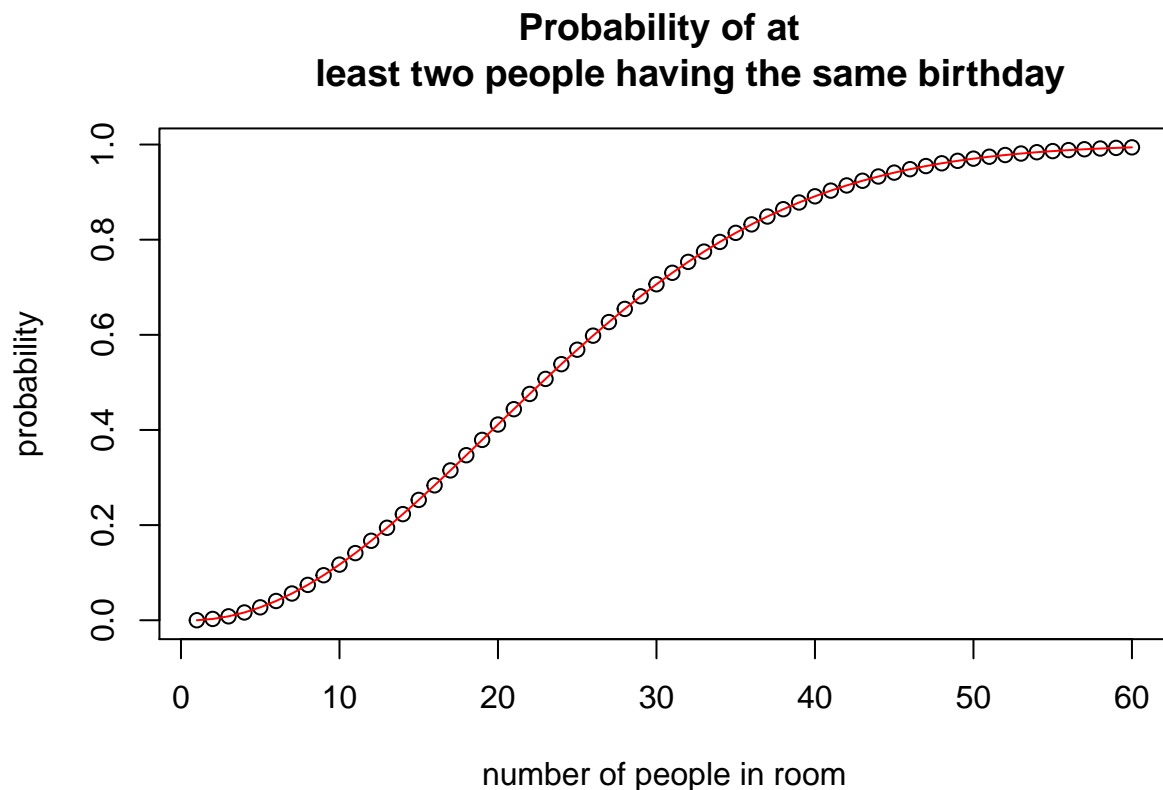
```
## [1] 0.5686997
```

Plot probability against the room size

```
vec<-0

for(i in 1:60){
  n=i
  y=365^n
  x=prod((365-(n-1)):365)/y
  p=1-x
  vec[i]=p                               #store probabilities in a vector
}

plot(1:60,vec,xlab="number of people in room",ylab="probability", main="Probability of at
least two people having the same birthday")
lines(1:60,vec,col="red")
```



iii. Based on your results from [ii], what is the minimum number of people in a room such that the probability of a match is greater than or equal to 50%?

```
which(vec>=0.5)
```

```
## [1] 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45
## [24] 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60
```

From this result, we can see that there needs to be at least 23 people in the room for there to be a probability of greater than or equal to 50%.

## Question 2: *The Birthday Problem -Again*

Based on the same assumptions of Problem 1, write a simulation in R that generates 100,000 simulated rooms of 25 people, and plot a histogram of the density vs. the number of birthday matches. The histogram represents your approximation of the distribution of birthday matches. Do your results agree with Problem 1? *Hint:* Consider using the R functions **sample** and **unique** to make the calculations easier.

```
match=0 #set if it is a match
num_match=0 #set the number of matches that occur
people=25
for(j in 1:100000){
  room<-sample(1:365,people,replace=TRUE)
  if(length(unique(room))<people){
    match=match+1 #add 1 if there is a match
    num_match[j]=length(room[room %in% room[duplicated(room)]])
  }else{
    num_match[j]=0 #indicate that there were no matches
  }
}
```

```

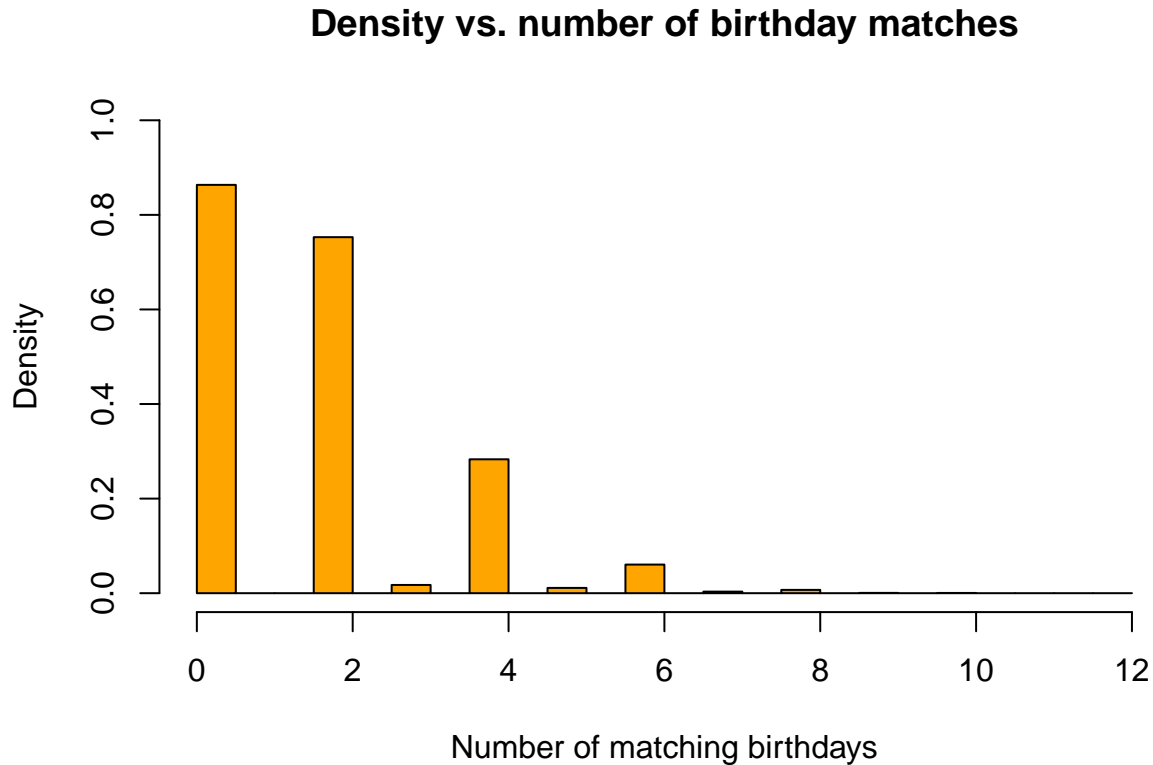
}

result = (match/100000)      #find the probability over all simulations
print(result)

## [1] 0.56827

hist(num_match,freq=FALSE, ylab="Density", xlab="Number of matching birthdays",
     main="Density vs. number of birthday matches", ylim=c(0:1), col="orange")

```



Since my answer came to be approximately 57% again, my results do agree with my results from Question 1.

### Question 3

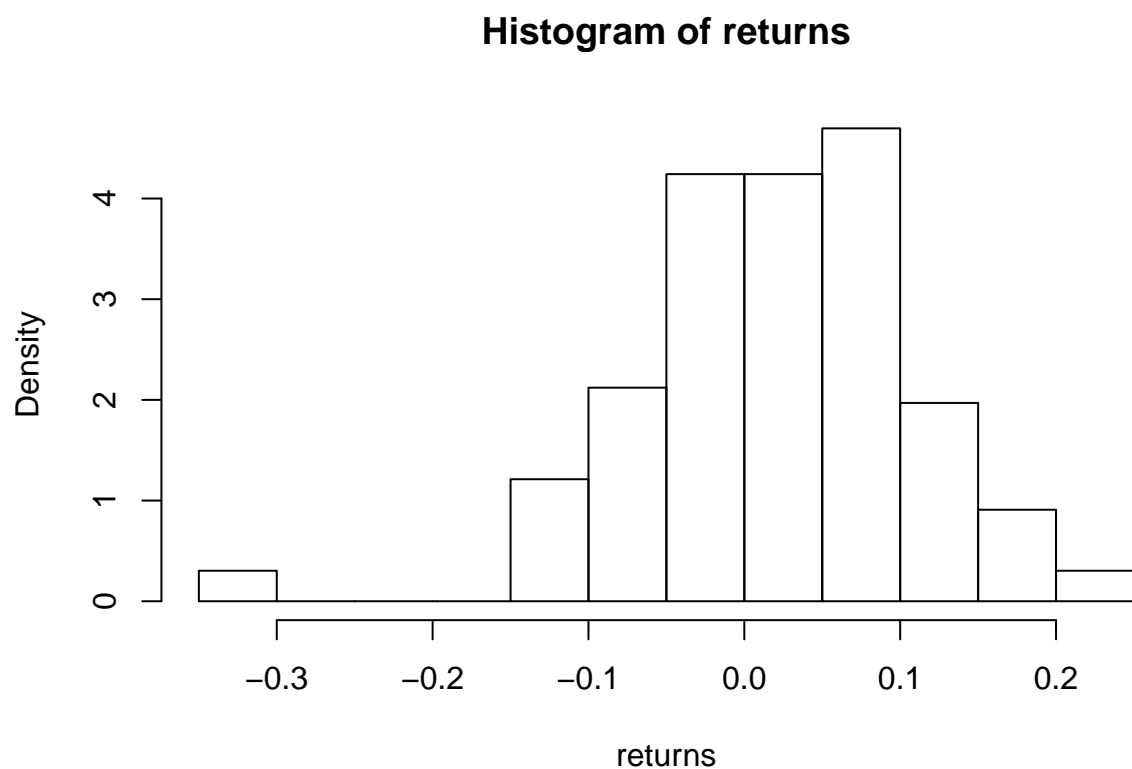
R Exercise: Plot a histogram of monthly returns of your favourite stock (use at least 10 years of data), and fit an appropriate distribution. Comment on your fit. Use this distribution to compute  $P(r > 10\%)$ , where  $r$ =return.

Taking stock data from APPLE between September 2007 to September 2018, I get the monthly returns and plot the following histogram.

```

returns<-monthlyReturn(AAPL)      #get monthly returns
hist(returns, prob=TRUE)           #make histogram

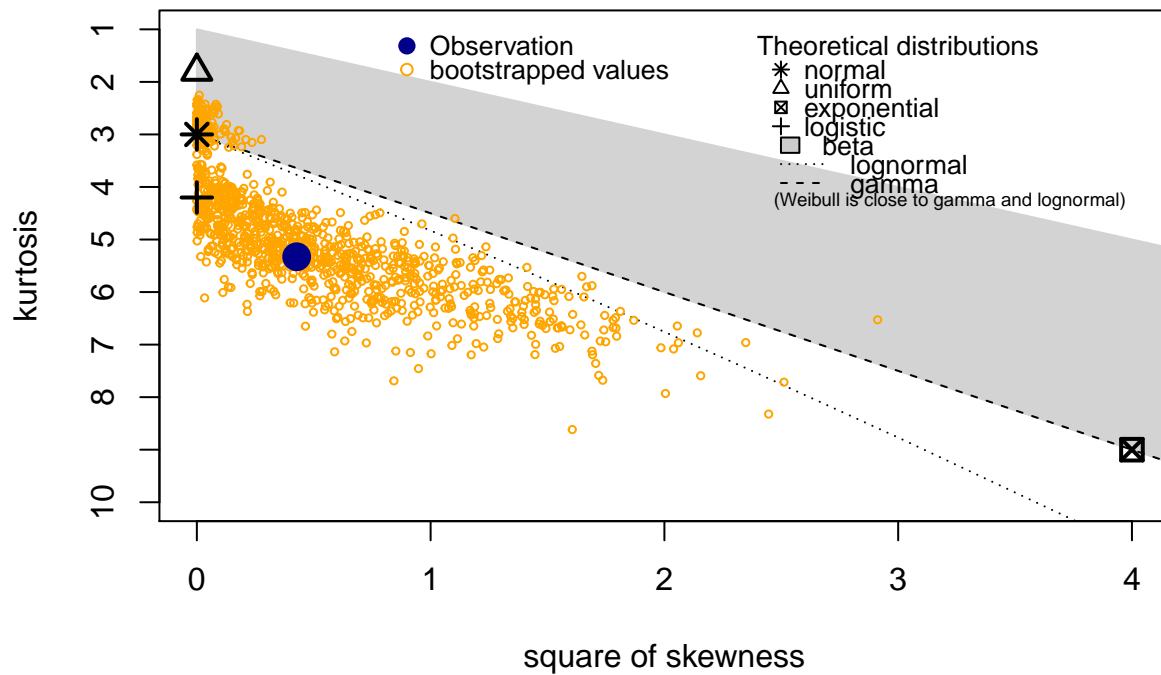
```



In order to determine which distribution method will have the best fit, I use the bootstrapping method and achieve the following Cullen Frey graph.

```
returns<-as.vector(returns)
descdist(returns, boot=1000)
```

## Cullen and Frey graph



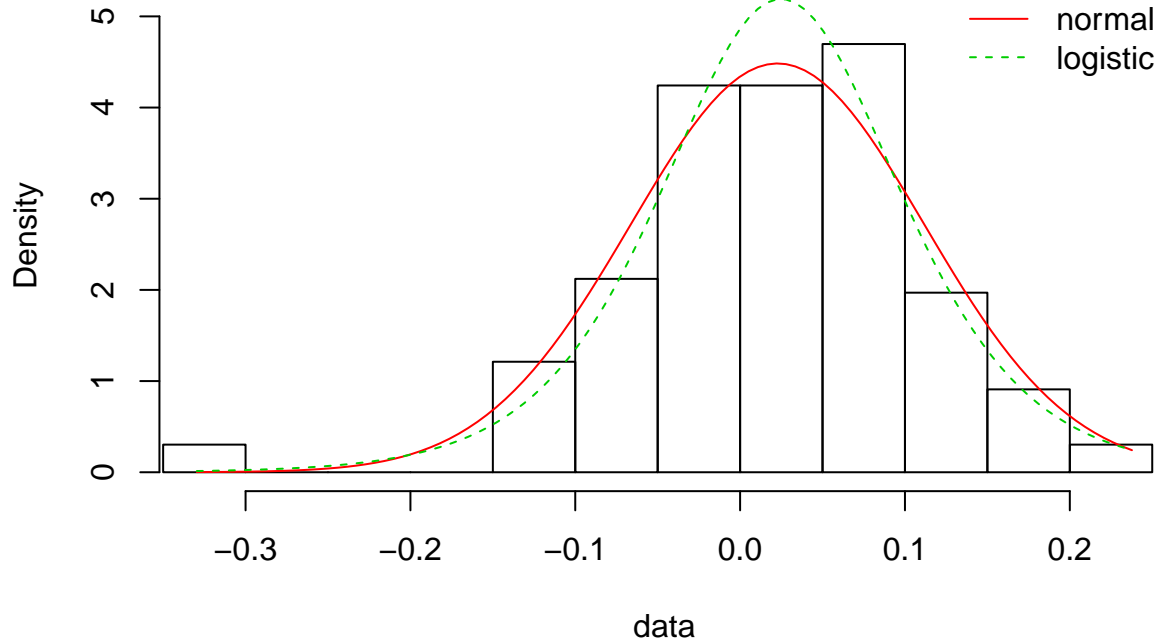
```
## summary statistics
## -----
## min: -0.3295582 max: 0.2377012
## median: 0.02520135
## mean: 0.02269983
## estimated sd: 0.08931283
## estimated skewness: -0.6534544
## estimated kurtosis: 5.32779
```

Given this information I believe the best fit in the logistic distribution. As we see below, the fit is much better in comparison to the normal distribution. Its higher peak accommodates more data points.

```
fn <- fitdist(returns, "norm")
fl <- fitdist(returns, "logis")

plot.legend <- c("normal", "logistic")
denscomp(list(fn, fl), legendtext = plot.legend)
```

## Histogram and theoretical densities



To compute  $P(r > 10\%)$ , I get the following result:

```
plogis(0.1, location=mean(returns), scale=sd(returns), lower.tail = FALSE)
```

```
## [1] 0.2961917
```

Therefore, the probability that  $r > 10\%$  is about 30%.

### Question 4:

Let  $X$  and  $Y$  have the joint density  $f(x, y) = cx(y - x)e^{-y}, 0 \leq x \leq y < \infty$ .

(a) Find  $c$

$$\int_0^{\infty} \int_0^y cx(y - x)e^{-y} dx dy = 1$$

$$\int_0^{\infty} \int_0^y x(y - x)e^{-y} dx dy = \frac{1}{c}$$

$$\int_0^{\infty} \int_0^y xye^{-y} - x^2e^{-y} dx dy = \frac{1}{c}$$

$$\int_0^{\infty} \int_0^y xye^{-y} - x^2e^{-y} dx dy = \frac{1}{c}$$

$$\int_0^{\infty} \frac{y^3 e^{-y}}{2} - \frac{y^3 e^{-y}}{3} dy = \frac{1}{c}$$

$$\int_0^{\infty} \frac{y^3 e^{-y}}{6} dy = \frac{1}{c}$$

$$\int_0^{\infty} y^3 e^{-y} dy = \frac{6}{c}$$

$$-y^3 e^{-y} + \int 3y^2 e^{-y} dy = \frac{6}{c}$$

$$-y^3 e^{-y} - 3y^2 e^{-y} + \int 6y e^{-y} dy = \frac{6}{c}$$

$$-y^3 e^{-y} - 3y^2 e^{-y} - 6y e^{-y} + 6 \int e^{-y} dy = \frac{6}{c}$$

$\vdots$

Solving using the bounds of integration going from 0 to  $\infty$  we get:

$$6 = \frac{6}{c}$$

$$c = 1$$

(b) find  $f_{X|Y}(x|y)$  and  $f_{Y|X}(y|x)$

First solve for marginals

$$f_x(x) = \int_x^{\infty} x(y-x)e^{-y} dy$$

$$= \int_x^{\infty} xy e^{-y} - x^2 e^{-y} dy$$

$$= x \int y e^{-y} dy - x^2 \int e^{-y} dy$$

$\vdots$

$$= x[xe^{-x} + e^{-x}] - x^2 e^{-x}$$

$$= xe^{-x}$$

$$\begin{aligned}
f_y(y) &= \int_0^y x(y-x)e^{-y}dx \\
&= \int_0^y xy e^{-y} - x^2 e^{-y} dx \\
&= y \int_0^y x e^{-y} dx - \int_0^y x^2 e^{-y} dx \\
&\quad \vdots \\
&= \frac{y^3 e^{-y}}{2} - \frac{y^3 e^{-y}}{3} \\
&= \frac{y^3 e^{-y}}{6}
\end{aligned}$$

Therefore,

$$f_{X|Y}(x|y) = \frac{6x(y-x)}{y^3}$$

and

$$f_{Y|X}(y|x) = \frac{(y-x)e^{-y}}{e^{-x}}$$

(c) Find  $E(X|Y)$  and  $E(Y|X)$

$$\begin{aligned}
E[X|Y] &= \int_0^y \frac{x6x(y-x)}{y^3} dx \\
&= \int_0^y \frac{6x^2y - 6x^3}{y^3} dx \\
&= \frac{1}{y^3} \int_0^y 6x^2y - 6x^3 dx \\
&\quad \vdots \\
&= \frac{1}{y^3} [2y^4 - \frac{3}{2}y^4] \\
&= \frac{y}{2}
\end{aligned}$$



$$\begin{aligned}
E[Y|X] &= \int_x^\infty \frac{y^2 e^{-y} - xy e^{-y}}{e^{-x}} dy \\
&= \frac{1}{e^{-x}} \int_x^\infty y^2 e^{-y} - xy e^{-y} dy \\
&= \frac{1}{e^{-x}} \int_x^\infty y^2 e^{-y} dy - \int_x^\infty xy e^{-y} dy \\
&\quad \vdots \\
&= \frac{1}{e^{-x}} [x^2 e^{-x} + 2xe^{-x} + 2e^{-x} - x^2 e^{-x} - xe^{-x}] \\
&= x + 2
\end{aligned}$$

## Question 5

Suppose that  $X_1, X_2, \dots, X_{10}$  is an i.i.d. sequence from an  $N(0, 1)$  distribution. Generate a sample of  $N = 10^3$  values from the distribution of  $\max(X_1, X_2, \dots, X_{10})$ . Calculate the mean and standard deviation of this sample.

```

x<-0 #make x empty

for (i in 1:1000){ #loop 10^3 times
  samp<-rnorm(10,mean=0, sd=1) #take a sample of 10 normal RVs
  x[i] <- max(samp) #store the maximums
}

#calculate the mean and standard deviation
mean(x)

## [1] 1.52893

sd(x)

## [1] 0.5576135

```

## Question 6

Generate i.i.d.  $X_1, \dots, X_n$  distributed  $\text{Exponential}(5)$  and compute  $M_n$  when  $n = 20$ . Repeat this  $N$  times, where  $N$  is large (if possible, take  $N = 10^5$ , otherwise as large as is feasible), and compute the proportion of values of  $M_n$  that lie between 0.19 and 0.21. Repeat this with  $n = 50$ . What property of convergence in probability do your results illustrate?

```

x<-0

for( i in 1:10^5){
  samp<-rexp(20, rate = 5)
  x[i]=mean(samp)
}

sub <- subset(x, x>=0.19 & x<=0.21)
length(sub)/length(x)

```

```
## [1] 0.17545
x<-0

for( i in 1:10^5){
  samp<-rexp(50, rate = 5)
  x[i]=mean(samp)
}

sub <- subset(x, x>=0.19 & x<=0.21)
length(sub)/length(x)
```

```
## [1] 0.27627
```

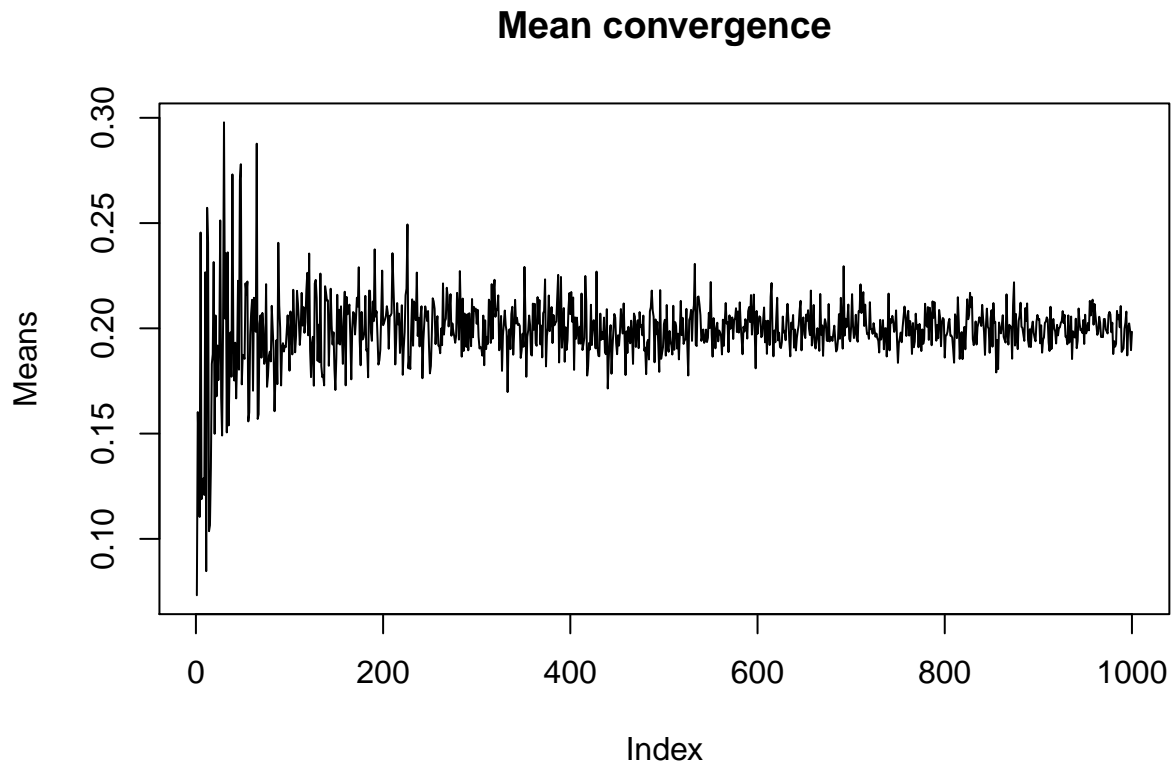
For  $n=20$  the proportion is approximately 18% and approximately 27% for  $n=50$ . This illustrates the Weak Law of Large Numbers.

## Question 7

Generate i.i.d.  $X_1, \dots, X_n$  distributed  $\text{Exponential}(5)$  with  $n$  large (take  $n = 10^5$  if possible). Plot the values  $M_1, M_2, \dots, M_n$ . To what value are they converging? How quickly?

```
sz<- 1
x<-0
for(i in 1:10^3){
  samp<-rexp(sz,rate=5)
  x[i]=mean(samp)
  sz<-sz+1
}

plot(x, type = "l", ylab="Means", main="Mean convergence")
```



From the plot we can see that the sample means converge to 0.2 and begins converging very quickly (almost right away).

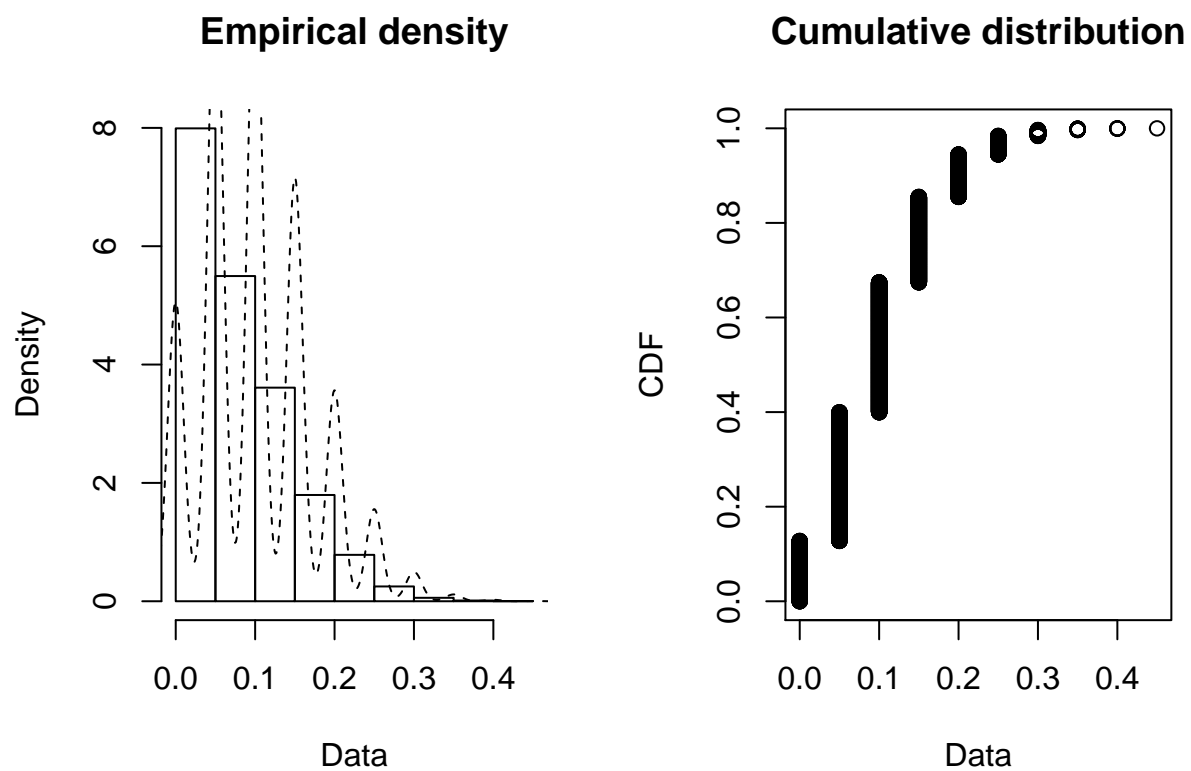
### Question 8

Generate  $N$  samples  $X_1, X_2, \dots, X_N$  from the  $\text{Binomial}(10, 0.01)$  distribution for  $N$  large ( $N = 10^4$ , if possible). Use these samples to construct a density histogram of the values of  $M_{20}$ . Comment on the shape of this graph.

```
x<-0

for( i in 1:10^4){
  samp<-rbinom(20, size=10, prob=0.01)
  x[i]=mean(samp)
}

plotdist(x, histo=TRUE, denv=TRUE)
```



We can see that the graph is skewed and the density is converging to zero.