```
#it has the property that for any set B of real numbers
```

 $\#P\{X \text{ in } B\}=\text{integrate}(f(x)dx) \text{ lim tending to } B$

#This function is called probability density function

#####R nomenclature for distributions

#For every distribution there are four commands. The commands for each distribution are prepende d with

#a letter to indicate the functionality:

```
#"d" returns the height of the probability density function
```

#"p" returns the cumulative density function

#"q" returns the inverse cumulative density function (quantiles)

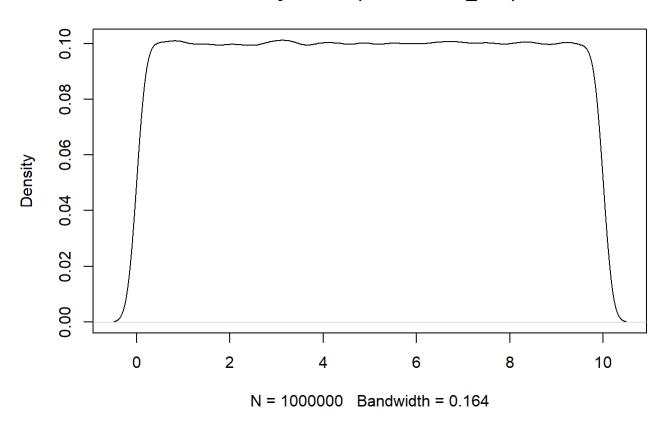
"r" returns randomly generated numbers

######THE UNIFORM DISTRIBUTION########

#The uniform distribution is a probability distribution where each value within a certain range #is equally liely to occur and values outside the range never occur. If we make a density plot #of a uniform distribution, it appears flat because no value is more likely(hence same density) t han any other value

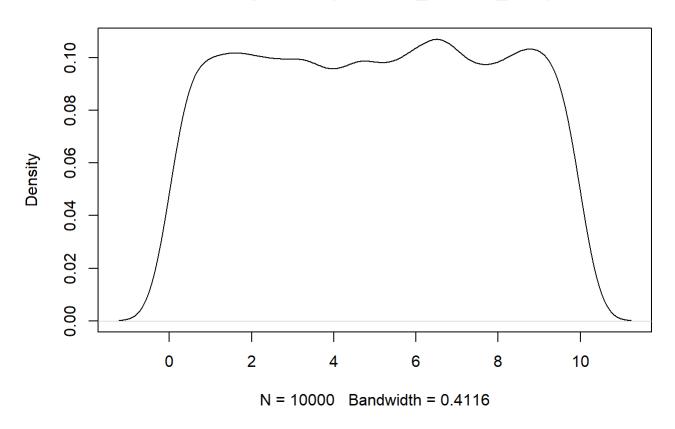
##creating data for a uniform distribution
uniform_dist<-runif(1000000,0,10)
#plotting the distribution
plot(density(uniform_dist))</pre>

density.default(x = uniform_dist)



more_random_data<-runif(n=10000,min=0,max=10)
plot(density(more_random_data))</pre>

density.default(x = more_random_data)



#The prefix "p" is used to determine the

#probability that an observation drawn from a distribution falls below a specified value (known as the

#cumulative distribution function.). In essence, "p" gives you the area under the distribution's

#density curve to the left of a certain value on the x axis. For example, in the uniform distribution

#above, there is a 25% chance that an observation will be in the range 0 to 2.5 and a 75% chance it will

#fall in the range 2.5 to 10.

#The p function gives the area covered in the pdf

punif(q=2.5,min=0,max=10)

[1] 0.25

#The prefix "q" is the inverse of the prefix "p": it
#returns the cutoff value (quantile) associated with a given probability.

qunif(p=0.4,min=0,max=10)

[1] 4

#Finally the "d" prefix gives you the density (height of the density curve)
#at a given point. Our plot of randomly generated uniform data was flat at
#0.1 so the density of the distribution should be constant at 0.1 throughout the range 0 to 10:

#checking the height of the curve at diff values
dunif(x=0,min=0,max=10)

[1] 0.1

dunif(5,0,10)

[1] 0.1

dunif(7.5,0,10)

[1] 0.1

dunif(10,0,10)

[1] 0.1

#The normal or Gaussian distribution is a continuous probability distribution
#characterized by a symmetric bell-shaped curve. A normal distribution is defined by its
#center (mean) and spread (standard deviation.). The bulk of the observations generated from
#a normal distribution lie near the mean, which lies at the exact center of the distribution:
as a rule of thumb, about 68% of the data lies within 1 standard deviation of the mean,
#95% lies within 2 standard deviations and 99.7% lies within 3 standard deviations.

#a standard normal curvel has mena=0 and sd =1

normal_dist<-rnorm(1000000,mean=0,sd=1)
plot(density(normal_dist))</pre>

##geting prob value at a particular point

pnorm(q=0.5, mean=0, sd=1)

[1] 0.6914625

```
pnorm(q=1,mean=0,sd=1)
```

[1] 0.8413447

```
##this value should be equal to 0.5+0.34
##84.13% area under the curve at point q=1
##this indiscates that there is 84.13% prob of getting a value <=1

#prob of values lesser than -1

prob_less_minusone<-pnorm(q=-1,mean=0,sd=1)
prob_less_minusone</pre>
```

[1] 0.1586553

```
##getting prob of observing a value greater than 1
1-pnorm(q=1,mean=0,sd=1)
```

[1] 0.1586553

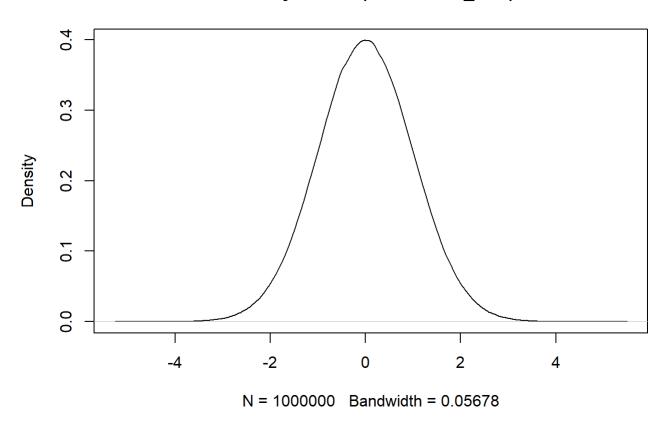
```
#this indicates that there is 15.86% prob of getting a value greater than 1
##getting a prob value between -1 and +1
pnorm(q=1,mean=0,sd=1)-pnorm(q=-1,mean=0,sd=1)
```

[1] 0.6826895

library(ggplot2)

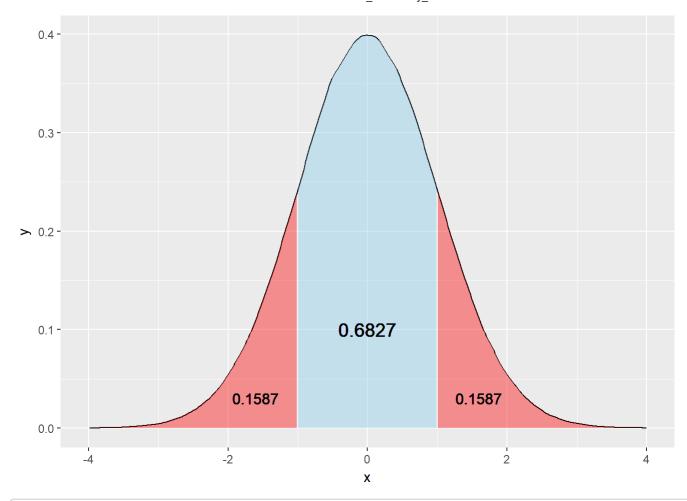
Warning: package 'ggplot2' was built under R version 3.4.4

density.default(x = normal_dist)



```
# Plot the density curve with the cutoff areas
norm frame = with(density(normal dist), # Create data frame density values
                  data.frame(x,y))
myplot <- ggplot(data = norm_frame, aes(x = x, y = y)) + # Create the plot
  geom_line() +
  geom_ribbon(data=subset(norm_frame,x < -1),</pre>
              aes(ymax=y),
              ymin=0,
              fill="red",
              alpha=0.4) +
  geom_ribbon(data=subset(norm_frame,x > 1),
              aes(ymax=y),
              ymin=0,
              fill="red",
              alpha=0.4) +
  geom ribbon(data=subset(norm frame, x > -1 \& x < 1),
              aes(ymax=y),
              ymin=0,
              fill="skyblue",
              alpha=0.4) +
  geom text(x=-1.6, y=0.03, label=round(prob less minusone, 4), size=4) +
  geom_text(x=1.6,y=0.03,label=round(prob_less_minusone,4),size=4) +
  geom_text(x=0,y=0.1,label=round(1-(prob_less_minusone*2),4),size=5) +
  xlim(-4,4)
myplot
```

Warning: Removed 132 rows containing missing values (geom path).



###finding quantile values

qnorm(p=0.025,mean=0,sd=1)

[1] -1.959964

qnorm(p=0.975,mean=0,sd=1)

[1] 1.959964

###checking the height of the curve
##density at that particular point in the pdf

dnorm(x=0, mean=0, sd=1)

[1] 0.3989423

dnorm(x=1.959964, mean=0, sd=1)

[1] 0.05844507

##thus we see when set cut of value of 0.05 for p value the cut-off value ##for z is 1.96 $\,$

#A continuous random variable whose probability function is given as $\#f(x) = Lambda*e^{-(-lambda*x)} \quad \text{if } x>=0 \\ \# = 0 \qquad \text{if } x<0$

#is called an exponentail random variable with parameter lambda.

#Finally the cumulative value of this function is given as follows $\#F(a)=1-e^{(lambda*a)}$

#Mean or expected value=1/lambda
#variance=1/lambda^2

#Use of exponential distribution is given by

#In practice, the exponential distribution often arises as the distribution of the #amount of time until some specific event occurs. For instance, the amount of time #(starting from now) until an earthquake occurs, or until a new war breaks out, or #until a telephone call you receive turns out to be a wrong number are all random #variables that tend in practice to have exponential distributions.

#example problem

#Suppose that the length of a phone call in minutes is an exponential random variable #with parameter lambda = 1
#10 . If someone arrives immediately ahead of you at a public #telephone booth, find the probability that you will have to wait #more than 10 minutes
#between 10 and 20 minutes

1-pexp(q=10, rate=1/10)

[1] 0.3678794

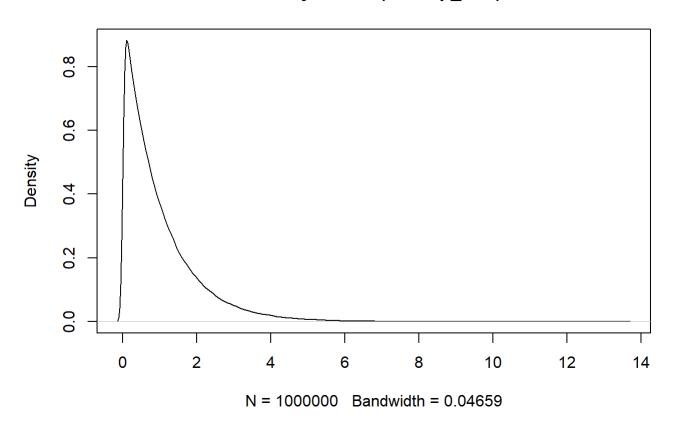
pexp(q=20, rate=1/10) - pexp(q=10, rate=1/10)

[1] 0.2325442

##creating exponential probability distributions

exp_dist<-rexp(n=1000000,rate=1)
plot(density(exp_dist))</pre>

density.default(x = exp_dist)



pexp(q=1,rate=1,lower.tail = FALSE)

[1] 0.3678794

1-(pexp(q=1,rate=1))

[1] 0.3678794

p_longer_than_1<-(1-(pexp(q=1,rate=1)))
p_longer_than_1</pre>

[1] 0.3678794

##the average arrival time in an exp dist is 1/lambda
##checking particular quantile value
qexp(p=0.6321206,rate=1)

[1] 1

```
#checking the height of the curve dexp(x=0,rate=1)
```

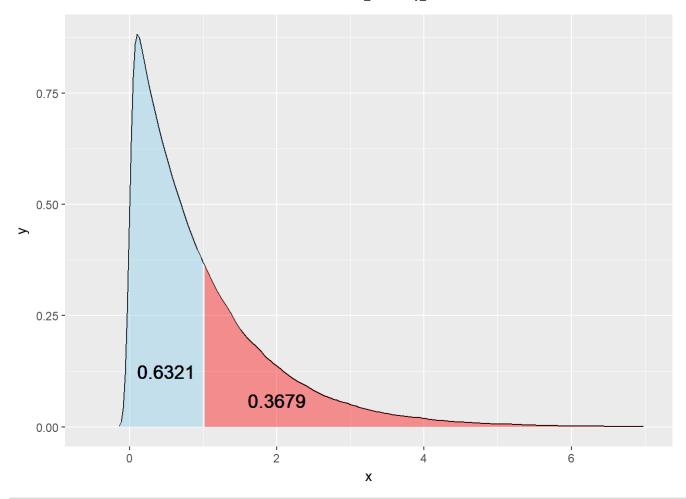
```
## [1] 1
```

```
dexp(x=2,rate=1)
```

```
## [1] 0.1353353
```

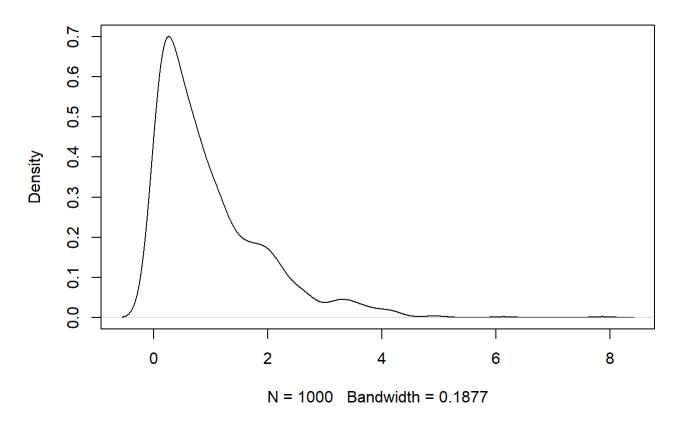
```
#Plot the density curve of the exp dist
exp_frame = with(density(exp_dist), # Create data frame of x,y density values
                 data.frame(x,y))
myplot2 \leftarrow ggplot(data = exp frame, aes(x = x, y = y)) + # Create the plot
  geom_line() +
  geom_ribbon(data=subset(exp_frame,x > 1),
              aes(ymax=y),
              ymin=0,
              fill="red",
              alpha=0.4)+
  geom_ribbon(data=subset(exp_frame,x < 1),</pre>
              aes(ymax=y),
              ymin=0,
              fill="skyblue",
              alpha=0.4) +
  geom_text(x=2,y=0.06,label=round(p_longer_than_1,4), size=5) +
  geom\_text(x=0.5,y=0.125,label=round(1-p\_longer\_than_1,4), size=5) +
  xlim(-0.5,7)
myplot2
```

Warning: Removed 248 rows containing missing values (geom path).



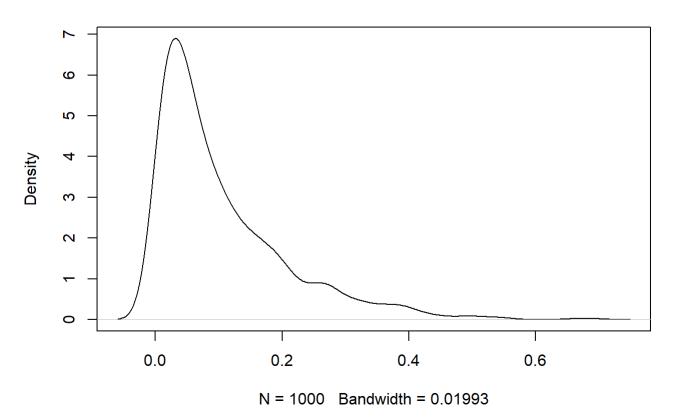
plot(density(rexp(n=1000,rate=1)))

density.default(x = rexp(n = 1000, rate = 1))



plot(density(rexp(n=1000,rate=10)))

density.default(x = rexp(n = 1000, rate = 10))



#as the value of rate increases the value of x decreases in the exp dist but #the value of density increases

#In probability and statistics, Student's t-distribution (or simply the t-distribution)
#is any member of a family of continuous probability distributions that arises when estimating
#the mean of a normally distributed population in situations where the sample size is small and
#population standard deviation is unknown.

##T_dist considers mean =0 and sd=1

#Generating a random t dist with diff degree of freedom

par(mfrow=c(2,2))

t_dist_5<-rt(n=1000000,df=5)
plot(density(t_dist_5))

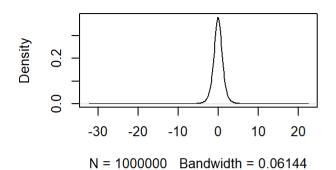
t_dist_10<-rt(n=1000000,df=10)
plot(density(t_dist_10))

t_dist_20<-rt(n=1000000,df=20)
plot(density(t_dist_20))</pre>

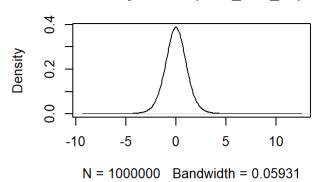
t dist 30<-rt(n=1000000,df=30)

plot(density(t_dist_30))

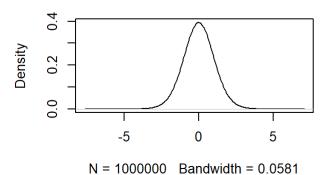
density.default(x = t_dist_5)



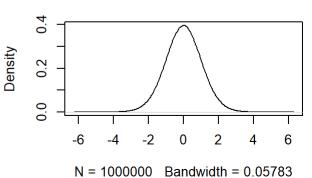
density.default(x = t_dist_10)



density.default(x = t_dist_20)



density.default(x = t_dist_30)

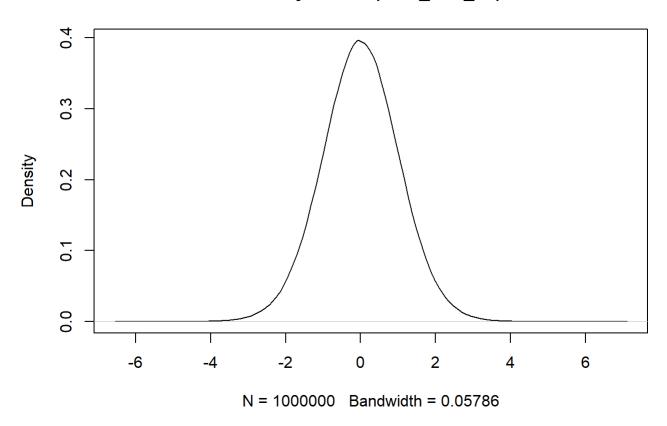


###checking prob values

par(mfrow=c(1,1))

t_dist_30<-rt(n=1000000,df=30)
plot(density(t_dist_30))</pre>

density.default(x = t_dist_30)



pt(q=0,df=30)

[1] 0.5

pt(q=0,df=10)

[1] 0.5

##we see the prob value is 0.5 since this is the mean value qt(p=0.5,df=30)

[1] 0

qt(p=0.99,df=30)

[1] 2.457262

##checking the height of the curve dt(x=0,df=30)

[1] 0.3956322