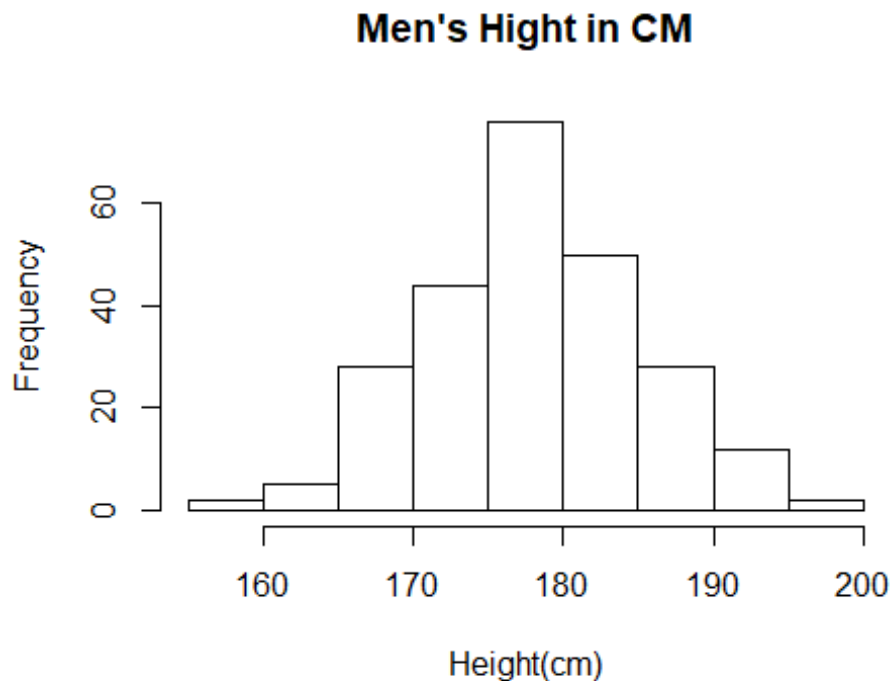


Lab4

```
download.file("http://www.openintro.org/stat/data/bdims.RData", destfile =
"bdims.RData")
load("bdims.RData")
mdims <- subset(bdims, bdims$sex == 1)
fdims <- subset(bdims, bdims$sex == 0)
mhgtmean <- mean(mdims$hgt)
mhgtstd <- sd(mdims$hgt)
```

Exercise 1: Generate separate histograms of the men's and women's heights. Then, compare and contrast the center, spread, and shape of these two height distributions.

```
#Male Height
hist(mdims$hgt, main = "Men's Hight in CM", xlab = "Height(cm)")
```

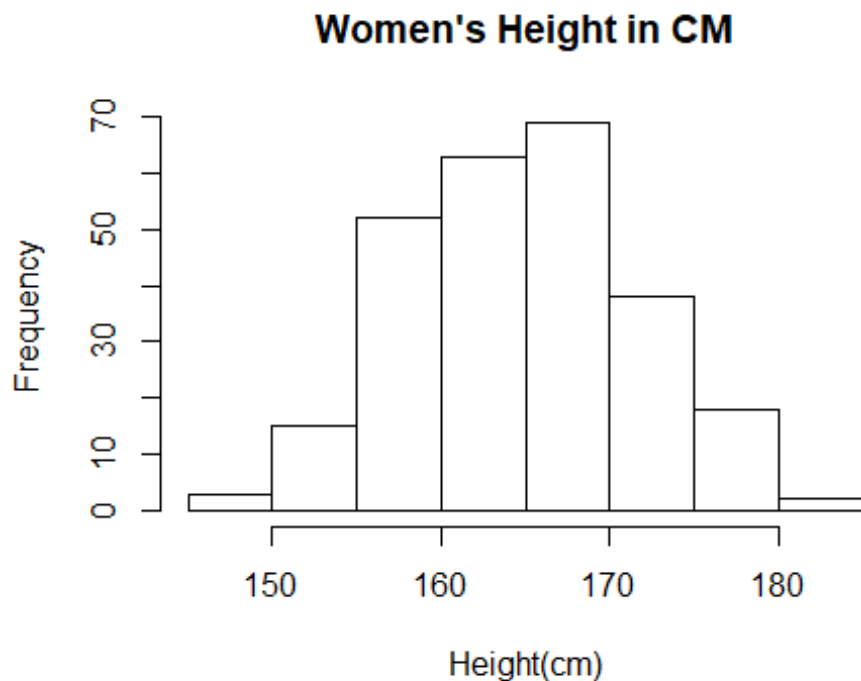


```
summary(mdims$hgt) # 5 Number Summary of Men's Height in cM.
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 157.2   172.9   177.8   177.7   182.7   198.1
```

```
#Female Height
```

```
hist(fdims$hgt, main = "Women's Height in CM", xlab = "Height(cm)")
```



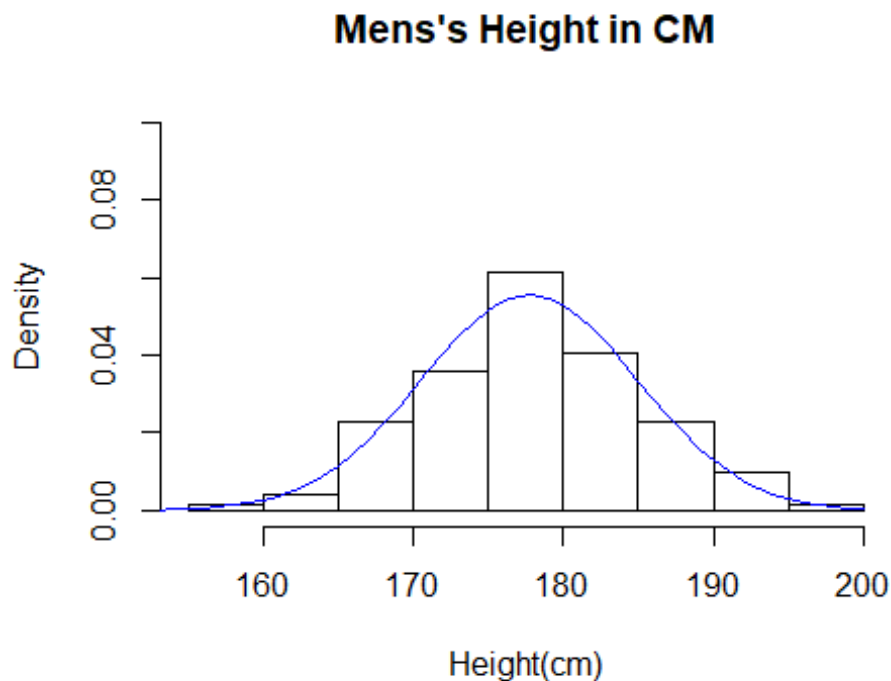
```
summary(fdims$hgt) # 5 Number Summary of Women's Height in cm.
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 147.2   160.0   164.5   164.9   169.5   182.9
```

The men's distribution appears to be symmetrical while the women's distribution is skewed left. The Men's median is 177.8 while the Women's is 164.5.

Exercise 2: Based on this plot, does it appear that the men's height data follow a nearly normal distribution? Explain.

```
hist(mdims$hgt, main = "Men's Height in CM", probability = TRUE,
     xlab="Height(cm)", ylim = c(0,0.1))
x <- 150:200
y <- dnorm(x = x, mean = mhgtmean, sd = mhgtstd)
lines(x = x, y = y, col = "blue")
```

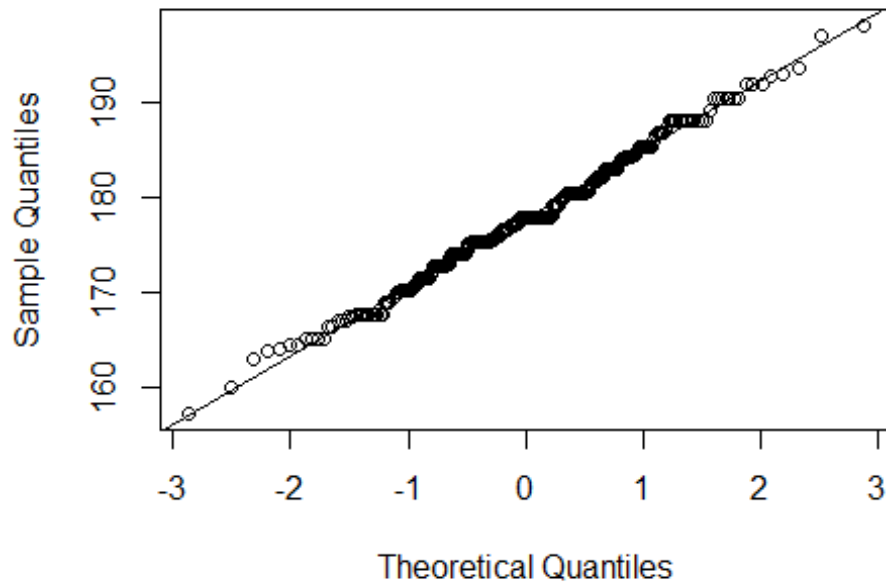


Based on this plot, the data follows a normal distribution. It follows the curve that has been imprinted from the `dnorm` command

Exercise 3: Make a normal probability plot of `sim_norm`. Do all of the points fall on the line? How does this plot compare to the probability plot for the actual data?

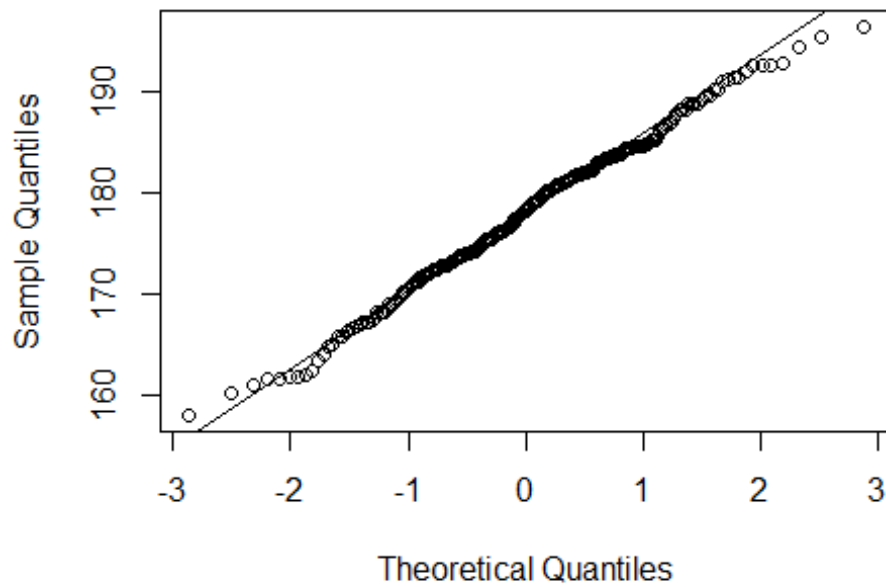
```
qqnorm(mdims$hgt)
qqline(mdims$hgt)
```

Normal Q-Q Plot



```
sim_norm <- rnorm(n = length(mdims$hgt), mean = mhgtmean, sd = mhgtsd)
qqnorm(sim_norm)
qqline(sim_norm)
```

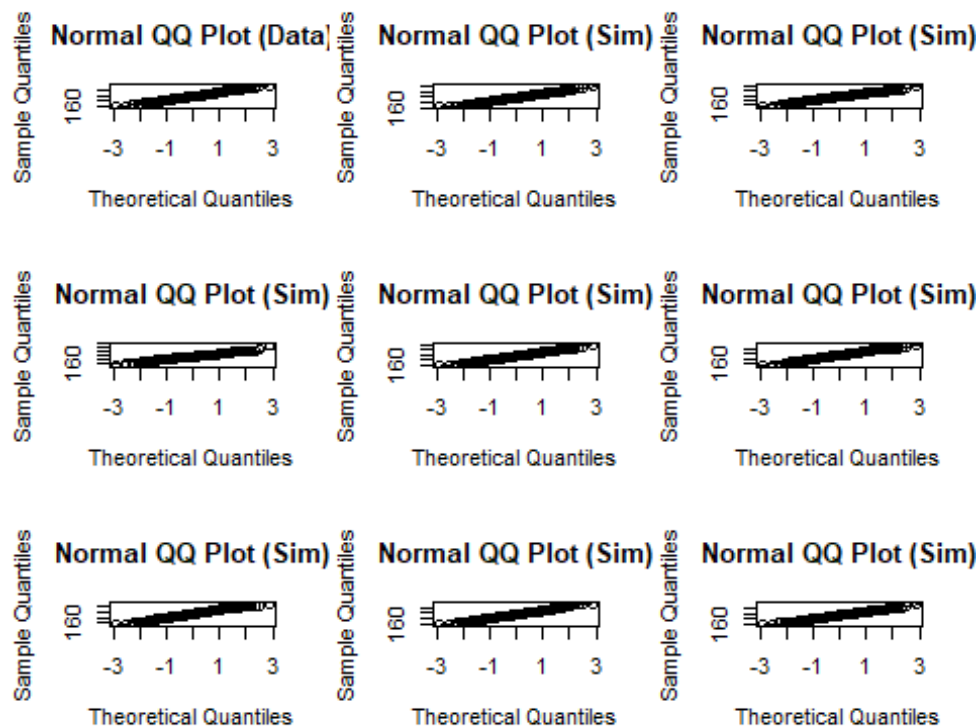
Normal Q-Q Plot



When comparing the simulated data most of the points fall either near or close to the line. The simulated data is very close to the actual data.

Exercise 4: Does the normal probability plot for `mdims$hgt` look similar to the plots created for the simulated data? That is, do the plots provide evidence that the male heights are nearly normal? Explain.

```
qqnormsim(mdims$hgt)
```

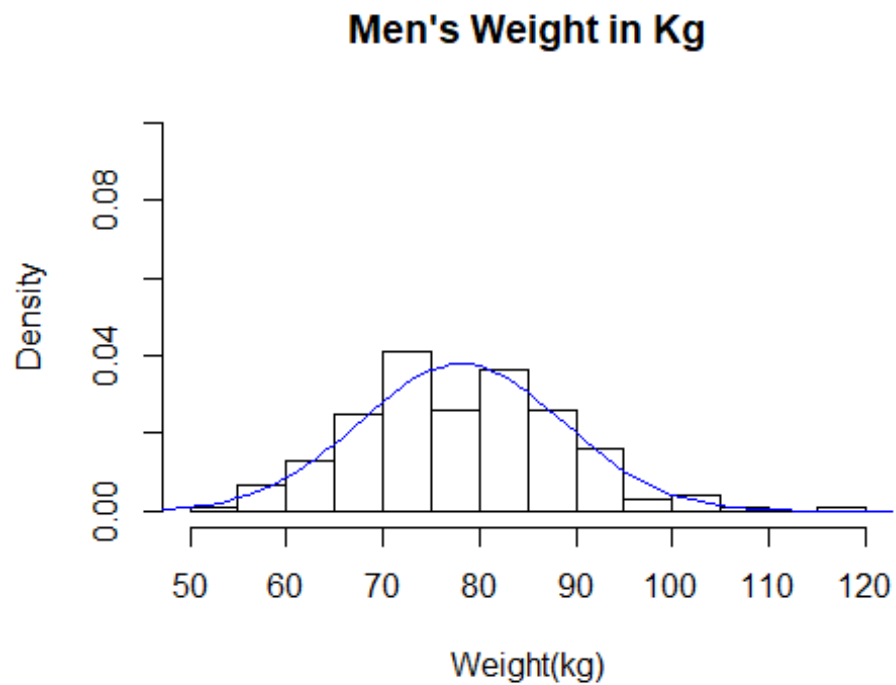


###The normal probability plot of `mdims$hgt` does look similar to the plots that were created for the simulated data. The plots do provide evidence that the males heights are normal. Most of the points fall either on the line or very close to it.

Exercise 5: Using the same procedure you used to judge the normality of the male height data in Exercises 2 through 4, explain your judgment as to whether or not the male weights appear to come from a normal distribution.

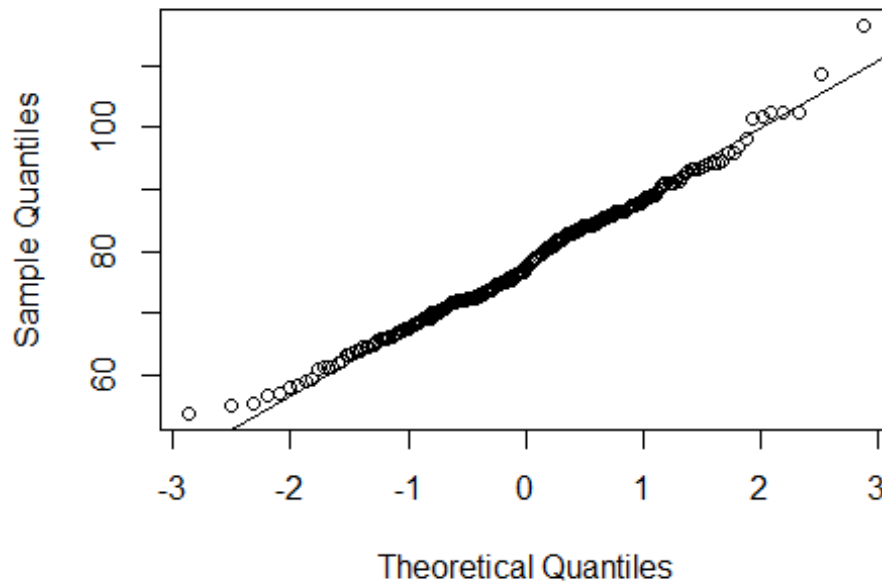
```
mwgtmean <- mean(mdims$wgt)
mwgtstd <- sd(mdims$wgt)
hist(mdims$wgt, main="Men's Weight in Kg", probability = TRUE, xlab =
"Weight(kg)", ylim = c(0,0.1))
x <- 0:1000
```

```
y <- dnorm(x = x, mean = mwgtmean, sd = mwgtstd)
lines(x = x, y = y, col = "blue")
```



```
qqnorm(mdims$wgt)
qqline(mdims$wgt)
```

Normal Q-Q Plot



```
summary(mdims$wgt)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  53.90   70.95   77.30   78.14   85.50   116.40
```

The male's weights seem to have a skewed right distribution.

Write out two probability questions that you would like to answer - one regarding male heights and one regarding male weights. Calculate those probabilities using both the theoretical normal distribution as well as the empirical distribution (four probabilities in all). Which variable, height or weight, had a closer agreement between the theoretical and empirical probabilities? Explain your reasoning.

#What is the probability of a male's height being greater than 5 foot 9 inches (175.26 cm)

```
1 - pnorm(q = 175.26, mean = mhgtmean, sd = mhgtstd)
```

```
## [1] 0.6353186
```

```
sum(mdims$hgt > 175.26)/length(mdims$hgt)
```

```
## [1] 0.6801619
```

#What is the probability of a female's height being greater than 5 foot 9 inches (175.26 cm)

```
1 - pnorm(q = 175.26, mean = mean(fdims$hgt), sd = sd(fdims$hgt))
```

```
## [1] 0.05623195
```

```
sum(fdims$hgt > 175.26)/length(fdims$hgt)
```

```
## [1] 0.06538462
```

The two questions I choose was what was the probability of either a male's or female's height being greater than 5 foot 9 inches. Both the theoretical and empirical probabilities were reasonably close. For males, the empirical distribution number was .635 while the empirical distribution number was .68. However with the female's empirical distribution number of .056 and the empirical distribution number was .065. Males are more likely to be taller than 5 foot 9 inches but females are not as likely.

Homework Assignment

1.1. D

1.2. A

1.3. B

1.4. C

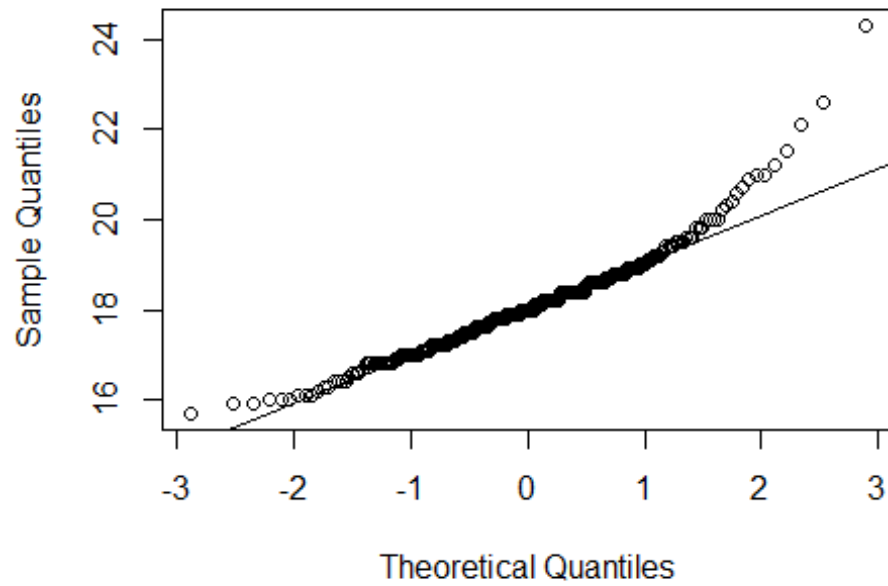
2. Age is a discrete variable and will produce a stepwise pattern.

As you can see, normal probability plots can be used both to assess normality and visualize skewness. Produce a normal probability plot for female knee diameter (kne.di). Based on this normal probability plot, is this variable left skewed, symmetric, or right skewed? Explain your reasoning. Use a histogram to confirm your findings.

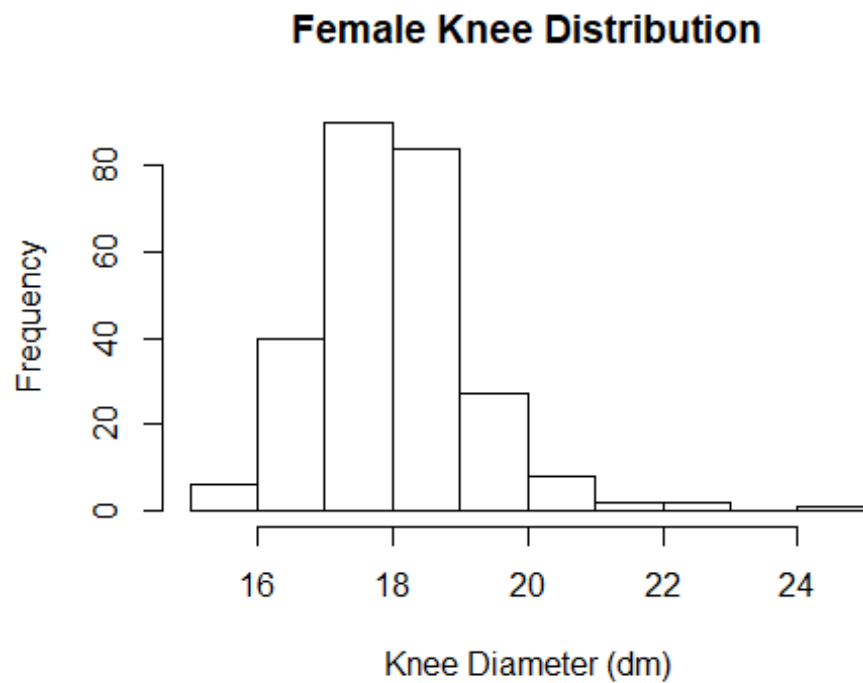
```
qqnorm(fdims$kne.di)
```

```
qqline(fdims$kne.di)
```


Normal Q-Q Plot



```
hist(fdims$kne.di, main = "Female Knee Distribution", xlab = "Knee Diameter  
(dm)")
```



###With the normal distribution and using the histogram, we can conclude that the female knee diameter is skewed right.