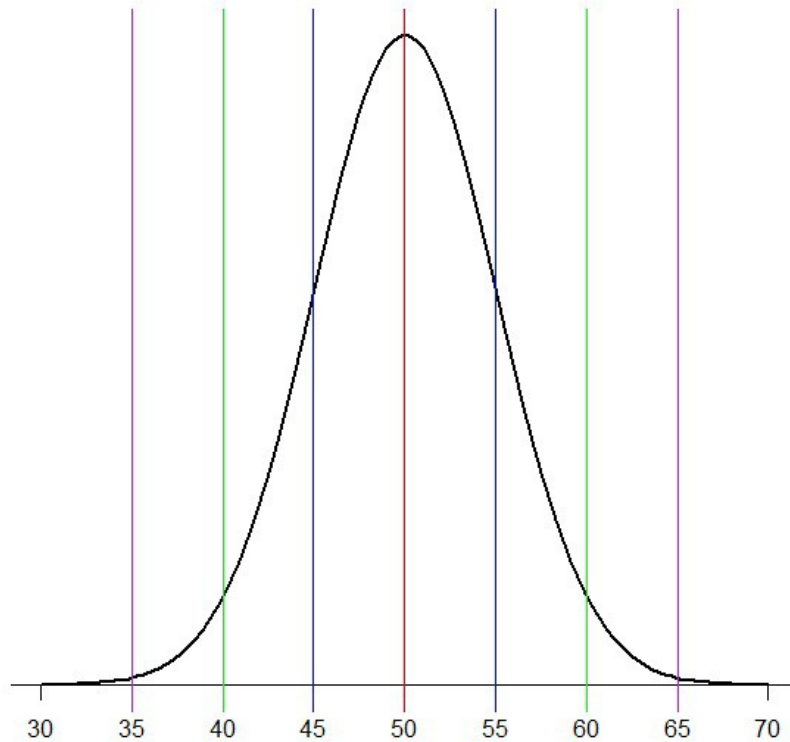


Assignment 1 (Unit 4)

- i. Get this graph in R/R Studio using the codes provided in the class slide of session 18:



#Define mean and SD

```
pop_mean<-50
```

```
pop_sd<-5
```

#Define lower and upper limits

```
LL <-pop_mean-pop_sd
```

```
UL <-pop_mean+ pop_sd
```

#Create a sequence of 100 x values based on pop mean and sd

```
x <-seq(-4,4, length=100)*pop_sd+pop_mean
```

```
y <-dnorm(x, pop_mean, pop_sd)
```

```
plot(x,y, type="l", lwd=2, axes=F, xlab="", ylab="")
```

```
sd_axis_bounds= 5
```

```
axis_bounds<-seq(-sd_axis_bounds*pop_sd + pop_mean, sd_axis_bounds*pop_sd+ pop_mean,
```

```
by=pop_sd)
```

```
axis(side=1, at=axis_bounds, pos=0)
```

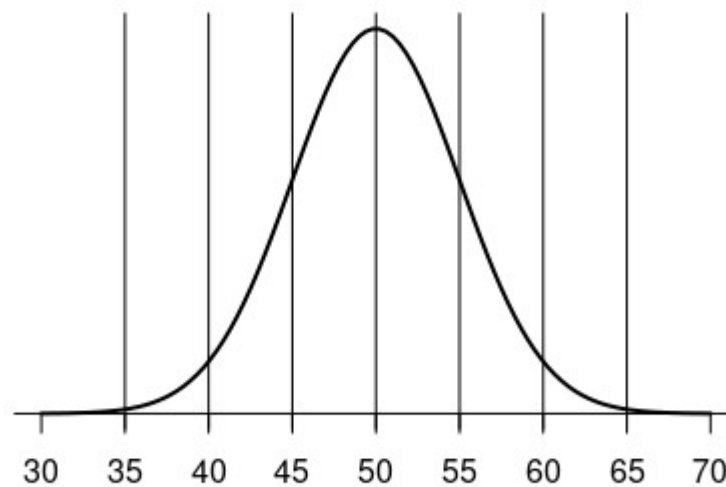
```
abline(v=50)
```

```
abline(v=50+pop_sd)
```

```

abline(v=50-pop_sd)
abline(v=50+2*pop_sd)
abline(v=50-2*pop_sd)
abline(v=50+3*pop_sd)
abline(v=50-3*pop_sd)

```



ii. Add annotation in this graph as follows:

45-55: mean \pm 1SD = 67% data with double edge arrow

40-60: mean \pm 2SD = 95% data with double edge arrow

35-65: mean \pm 3SD = 99% data with double edge arrow

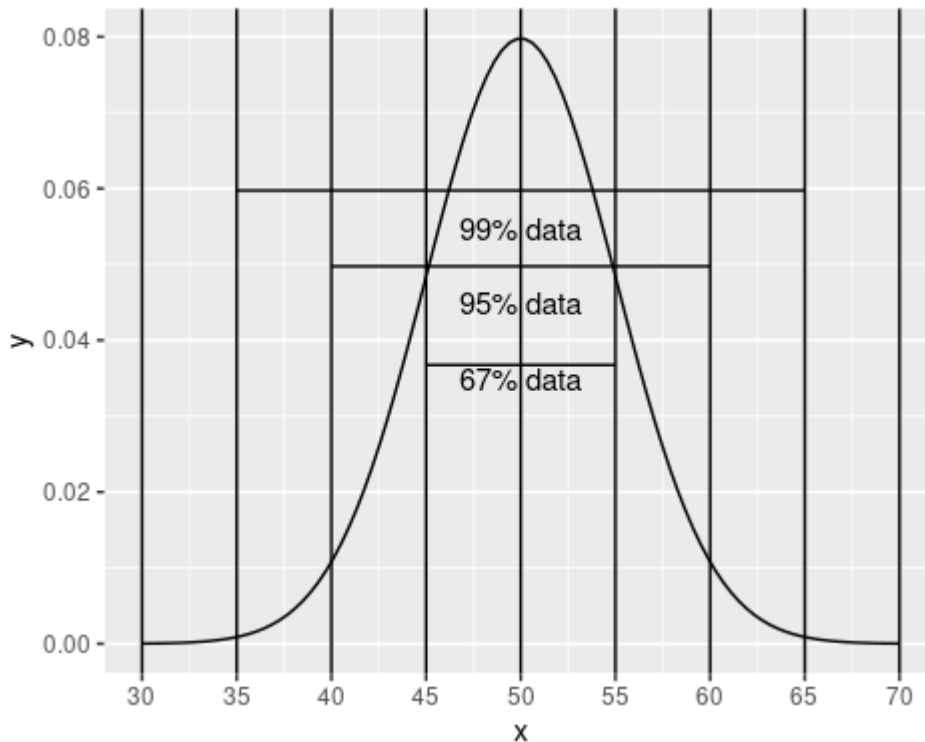
Note: You can use the “ggplot2” package if you want!

```

library(ggplot2)
ggplot(data=as.data.frame(x,y),aes(x=x,y=y))+geom_line()+scale_x_continuous(breaks =
axis_bounds)+geom_vline(xintercept = seq(30,70,by=pop_sd))+annotate(geom = "text",x =
50,y=mean(y)+0.01, label = "67% data")+annotate(geom = "segment",x = 45,y=mean(y)

```

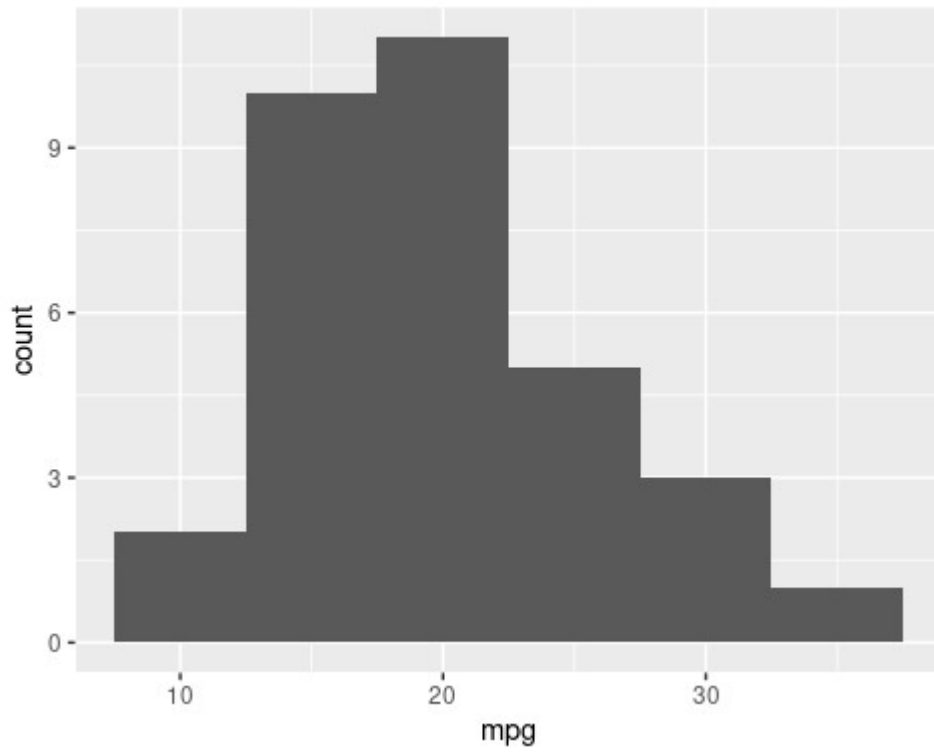
```
+0.012,xend = 55,yend = mean(y)+0.012)+annotate(geom = "text",x = 50,y=mean(y)+0.02, label
= "95% data")+annotate(geom = "segment",x = 40,y=mean(y)+0.025,xend = 60,yend = mean(y)
+0.025)+annotate(geom = "text",x = 50,y=mean(y)+0.03, label = "99% data")+annotate(geom =
"segment",x = 35,y=mean(y)+0.035,xend = 65,yend = mean(y)+0.035)
```



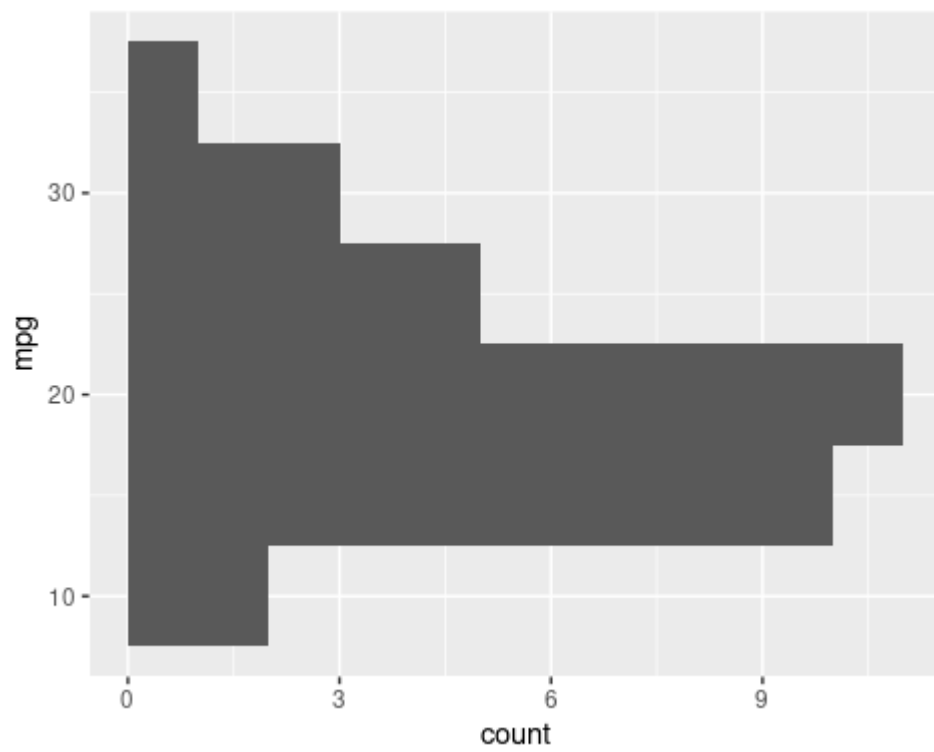
Assignment 2 (Unit 4):

- i. Get stem-leaf plot, histogram and normal q-q plot of mpg variable of the built-in “mtcars” data of R.

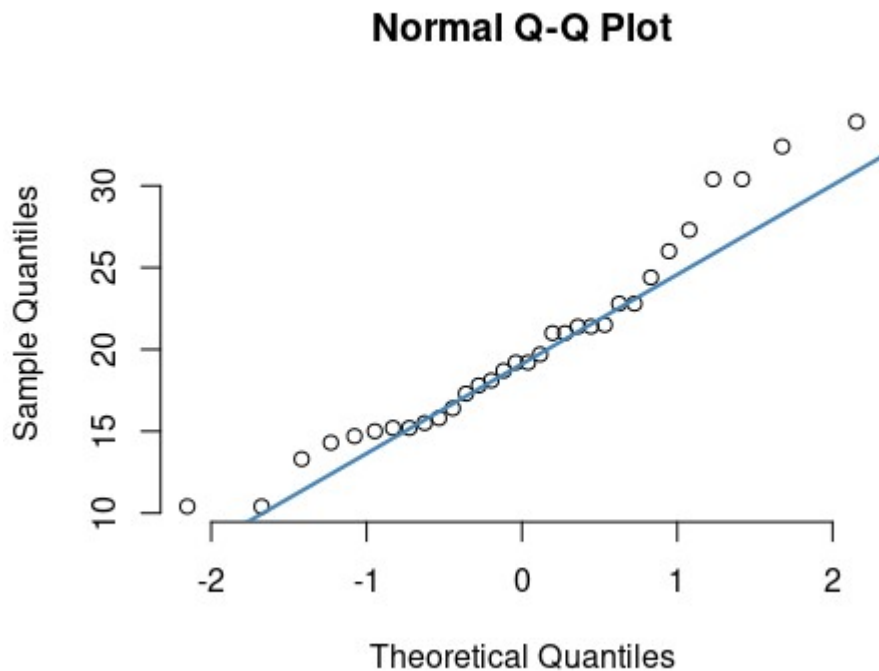
```
cars_data<-mtcars
ggplot(data=cars_data,aes(x=mpg))+geom_histogram(binwidth = 5)
```



```
ggplot(data = cars_data,aes(mpg))+geom_histogram(binwidth = 5)+coord_flip()
```



```
cars_data<-mtcars
qqnorm(cars_data$mpg, pch = 1, frame = FALSE)
qqline(cars_data$mpg, col = "steelblue", lwd = 2)
```



- ii. Test the normality of mpg variable of mtcars data using Shapiro-Wilk test of normality and take decision based on the p-value of this test for null and alternative hypothesis you wrote for this test. Explain why this test of normality must be used here!

```
shapiro.test(cars_data$mpg)

##
## Shapiro-Wilk normality test
##
## data: cars_data$mpg
## W = 0.94756, p-value = 0.1229
```

Null Hypothesis(H_0): The data follows normal distribution Alternative Hypothesis(H_1): The data does not follow normal distribution Since the p-value obtained from the test is not less than 0.05 we can say that the data comes from normal distribution i.e Null Hypothesis is accepted.

Since the number of samples ($32 < 50$) we cannot use Kolmogorov-Smirnov. Thus, we have to use Shapiro-Wilk normality test.