

**Technological Institute of Tijuana****Academic Subdirector****Systems and Computing Department****SEMESTER:** August - December 2021**CAREER:** Computer Systems Engineer**MATTER:** Data Mining**JOB NAME:** Unit 4 - Exam**STUDENT NAME AND CONTROL NUMBER:**

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TEACHER NAME: Jose Christian Romero Hernandez**DATE OF DELIVERY:** December 12, 2021**Development**

The first part is to decide where we want to work and we put the direction

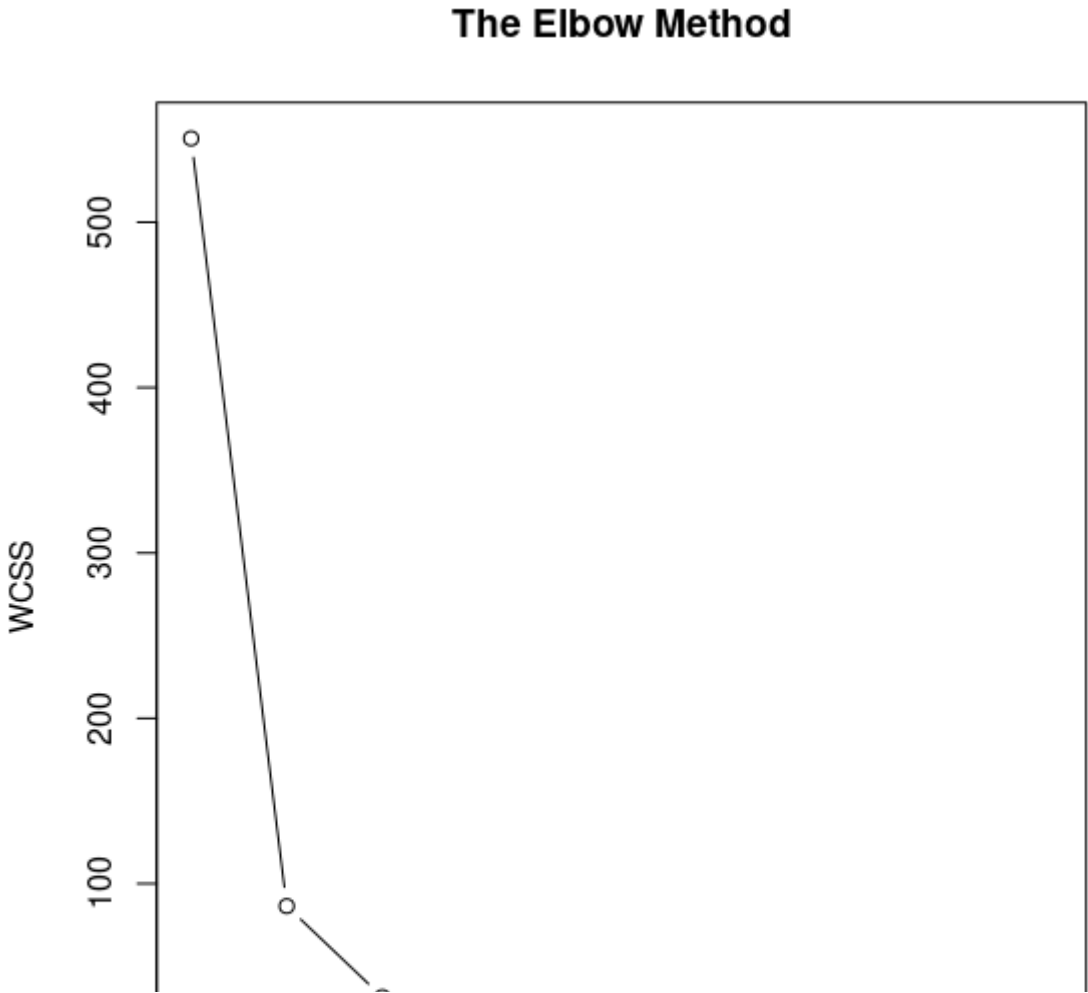
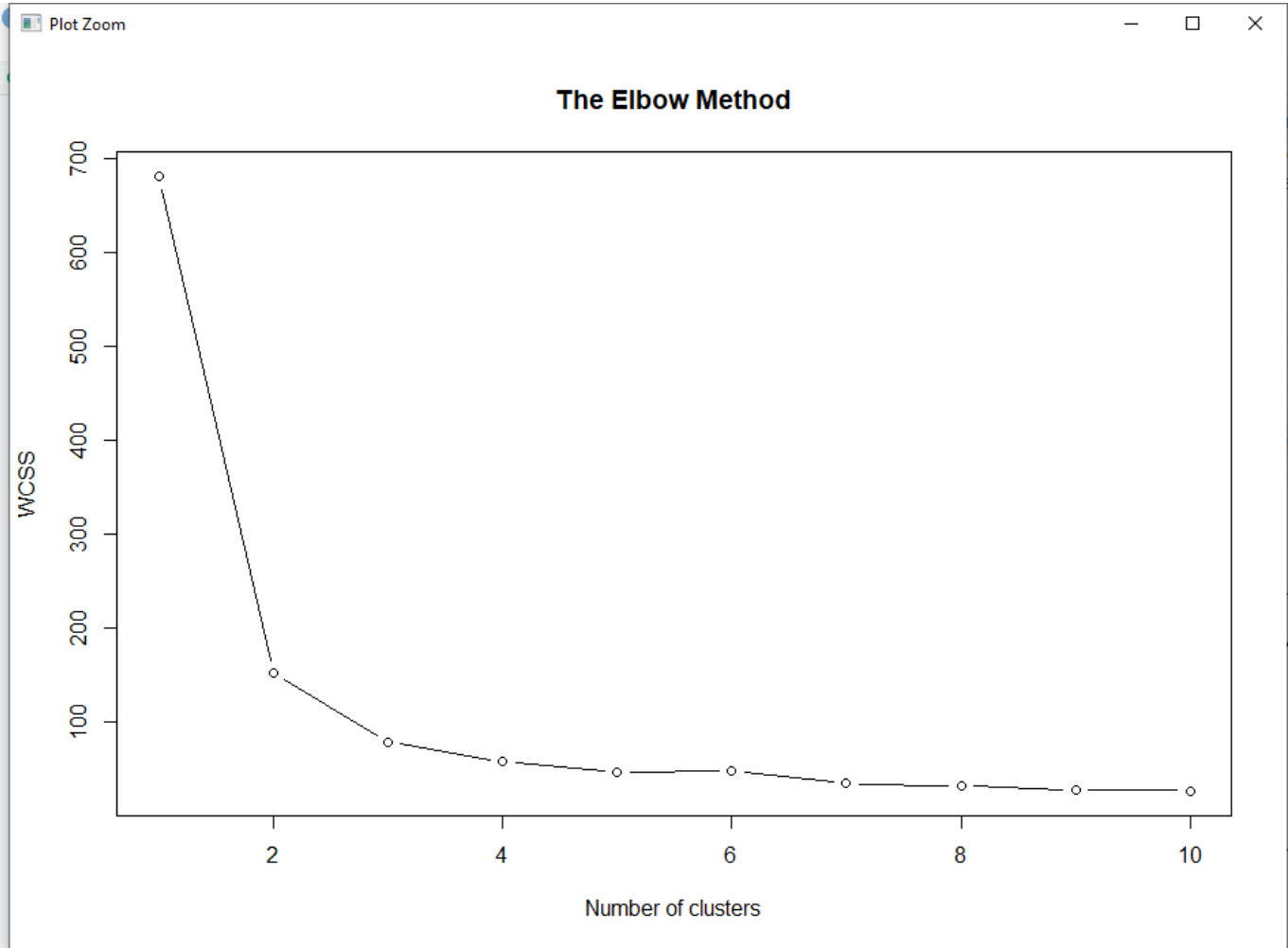
```
getwd()  
setwd("C:/Users/vanem/OneDrive/Documentos/9 SEMESTRE/Mineria/Repo  
mineria/DataMining/MachineLearning/LogisticRegression")  
getwd()
```

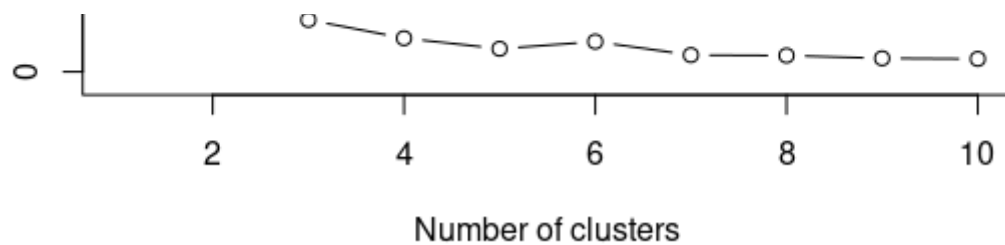
Now we import the data set and select the columns

```
dataset <- read.csv(file.choose())  
dat1 = dataset[1:4]  
dat2 = dataset[3:4]
```

Well here we used the the elbow method to find the optimal number of cluster, but these method is on the function because is more easy to visualizing the graphic, and after we call the function and send like param the data set and the resul is here.

```
TEM <- function(dataset){  
  set.seed(6)  
  wcss = vector()  
  for (i in 1:10) wcss[i] = sum(kmeans(dataset, i)$withinss)  
  plot(1:10,  
       wcss,  
       type = 'b',  
       main = paste('The Elbow Method'),  
       xlab = 'Number of clusters',  
       ylab = 'WCSS')  
}  
  
TEM(dat1)  
TEM(dat2)
```





Like we can see the breaking point is in the point number 3 so is these the number we used more later.

Here we used the model and fitting the dataset

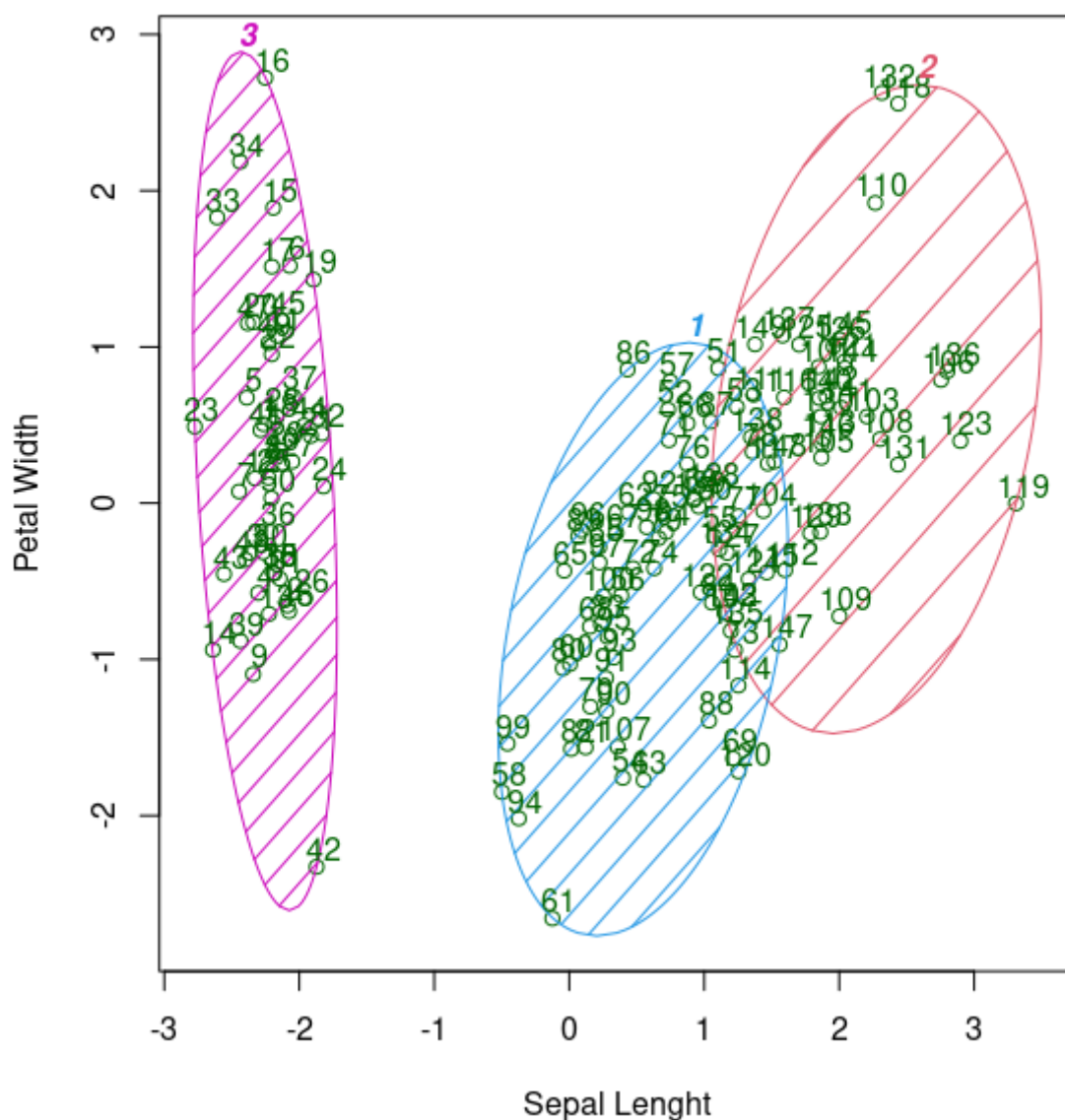
```
set.seed(29)
kmeans = kmeans(x = dataset, centers = 3)
y_kmeans = kmeans$cluster
```

Here we visualize the cluster, but first we need the cluster library in order to execute the code well

```
library(cluster)

#Sepal lenght and Petal width
clusplot(dat1,
          y_kmeans,
          lines = 0,
          shade = TRUE,
          color = TRUE,
          labels = 2,
          plotchar = FALSE,
          span = TRUE,
          main = paste('Clusters of Iris'),
          xlab = 'Sepal lenght',
          ylab = 'Petal width')
```

Clusters of Iris



These two components explain 95.8 % of the point variability.

```
#Petal lenght and Petal width
clusplot(dat2,
  y_kmeans,
  lines = 0,
  shade = TRUE,
  color = TRUE,
  labels = 2,
  plotchar = FALSE,
  span = TRUE,
  main = paste('Clusters of Iris'),
  xlab = 'Petal lenght',
  ylab = 'Petal width')
```

