

Unsupervised Learning - K-mean Clustering

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11/06/2021

R Markdown

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```
set.seed(256)
# Load iris dataset into a new variable iris2
data(iris)
str(iris)
```

```
## 'data.frame':  150 obs. of  5 variables:
## $ Sepal.Length: num  5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
## $ Sepal.Width : num  3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
## $ Petal.Length: num  1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
## $ Petal.Width : num  0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
## $ Species      : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
#Remove the initial label of Species from original dataset
iris2 <- iris[,-5]
iris2
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1           5.1           3.5           1.4           0.2
## 2           4.9           3.0           1.4           0.2
## 3           4.7           3.2           1.3           0.2
## 4           4.6           3.1           1.5           0.2
## 5           5.0           3.6           1.4           0.2
## 6           5.4           3.9           1.7           0.4
## 7           4.6           3.4           1.4           0.3
## 8           5.0           3.4           1.5           0.2
## 9           4.4           2.9           1.4           0.2
## 10          4.9           3.1           1.5           0.1
## 11          5.4           3.7           1.5           0.2
## 12          4.8           3.4           1.6           0.2
## 13          4.8           3.0           1.4           0.1
## 14          4.3           3.0           1.1           0.1
## 15          5.8           4.0           1.2           0.2
## 16          5.7           4.4           1.5           0.4
```

## 17	5.4	3.9	1.3	0.4
## 18	5.1	3.5	1.4	0.3
## 19	5.7	3.8	1.7	0.3
## 20	5.1	3.8	1.5	0.3
## 21	5.4	3.4	1.7	0.2
## 22	5.1	3.7	1.5	0.4
## 23	4.6	3.6	1.0	0.2
## 24	5.1	3.3	1.7	0.5
## 25	4.8	3.4	1.9	0.2
## 26	5.0	3.0	1.6	0.2
## 27	5.0	3.4	1.6	0.4
## 28	5.2	3.5	1.5	0.2
## 29	5.2	3.4	1.4	0.2
## 30	4.7	3.2	1.6	0.2
## 31	4.8	3.1	1.6	0.2
## 32	5.4	3.4	1.5	0.4
## 33	5.2	4.1	1.5	0.1
## 34	5.5	4.2	1.4	0.2
## 35	4.9	3.1	1.5	0.2
## 36	5.0	3.2	1.2	0.2
## 37	5.5	3.5	1.3	0.2
## 38	4.9	3.6	1.4	0.1
## 39	4.4	3.0	1.3	0.2
## 40	5.1	3.4	1.5	0.2
## 41	5.0	3.5	1.3	0.3
## 42	4.5	2.3	1.3	0.3
## 43	4.4	3.2	1.3	0.2
## 44	5.0	3.5	1.6	0.6
## 45	5.1	3.8	1.9	0.4
## 46	4.8	3.0	1.4	0.3
## 47	5.1	3.8	1.6	0.2
## 48	4.6	3.2	1.4	0.2
## 49	5.3	3.7	1.5	0.2
## 50	5.0	3.3	1.4	0.2
## 51	7.0	3.2	4.7	1.4
## 52	6.4	3.2	4.5	1.5
## 53	6.9	3.1	4.9	1.5
## 54	5.5	2.3	4.0	1.3
## 55	6.5	2.8	4.6	1.5
## 56	5.7	2.8	4.5	1.3
## 57	6.3	3.3	4.7	1.6
## 58	4.9	2.4	3.3	1.0
## 59	6.6	2.9	4.6	1.3
## 60	5.2	2.7	3.9	1.4
## 61	5.0	2.0	3.5	1.0
## 62	5.9	3.0	4.2	1.5
## 63	6.0	2.2	4.0	1.0
## 64	6.1	2.9	4.7	1.4
## 65	5.6	2.9	3.6	1.3
## 66	6.7	3.1	4.4	1.4
## 67	5.6	3.0	4.5	1.5
## 68	5.8	2.7	4.1	1.0
## 69	6.2	2.2	4.5	1.5
## 70	5.6	2.5	3.9	1.1

## 71	5.9	3.2	4.8	1.8
## 72	6.1	2.8	4.0	1.3
## 73	6.3	2.5	4.9	1.5
## 74	6.1	2.8	4.7	1.2
## 75	6.4	2.9	4.3	1.3
## 76	6.6	3.0	4.4	1.4
## 77	6.8	2.8	4.8	1.4
## 78	6.7	3.0	5.0	1.7
## 79	6.0	2.9	4.5	1.5
## 80	5.7	2.6	3.5	1.0
## 81	5.5	2.4	3.8	1.1
## 82	5.5	2.4	3.7	1.0
## 83	5.8	2.7	3.9	1.2
## 84	6.0	2.7	5.1	1.6
## 85	5.4	3.0	4.5	1.5
## 86	6.0	3.4	4.5	1.6
## 87	6.7	3.1	4.7	1.5
## 88	6.3	2.3	4.4	1.3
## 89	5.6	3.0	4.1	1.3
## 90	5.5	2.5	4.0	1.3
## 91	5.5	2.6	4.4	1.2
## 92	6.1	3.0	4.6	1.4
## 93	5.8	2.6	4.0	1.2
## 94	5.0	2.3	3.3	1.0
## 95	5.6	2.7	4.2	1.3
## 96	5.7	3.0	4.2	1.2
## 97	5.7	2.9	4.2	1.3
## 98	6.2	2.9	4.3	1.3
## 99	5.1	2.5	3.0	1.1
## 100	5.7	2.8	4.1	1.3
## 101	6.3	3.3	6.0	2.5
## 102	5.8	2.7	5.1	1.9
## 103	7.1	3.0	5.9	2.1
## 104	6.3	2.9	5.6	1.8
## 105	6.5	3.0	5.8	2.2
## 106	7.6	3.0	6.6	2.1
## 107	4.9	2.5	4.5	1.7
## 108	7.3	2.9	6.3	1.8
## 109	6.7	2.5	5.8	1.8
## 110	7.2	3.6	6.1	2.5
## 111	6.5	3.2	5.1	2.0
## 112	6.4	2.7	5.3	1.9
## 113	6.8	3.0	5.5	2.1
## 114	5.7	2.5	5.0	2.0
## 115	5.8	2.8	5.1	2.4
## 116	6.4	3.2	5.3	2.3
## 117	6.5	3.0	5.5	1.8
## 118	7.7	3.8	6.7	2.2
## 119	7.7	2.6	6.9	2.3
## 120	6.0	2.2	5.0	1.5
## 121	6.9	3.2	5.7	2.3
## 122	5.6	2.8	4.9	2.0
## 123	7.7	2.8	6.7	2.0
## 124	6.3	2.7	4.9	1.8

```
## 125      6.7      3.3      5.7      2.1
## 126      7.2      3.2      6.0      1.8
## 127      6.2      2.8      4.8      1.8
## 128      6.1      3.0      4.9      1.8
## 129      6.4      2.8      5.6      2.1
## 130      7.2      3.0      5.8      1.6
## 131      7.4      2.8      6.1      1.9
## 132      7.9      3.8      6.4      2.0
## 133      6.4      2.8      5.6      2.2
## 134      6.3      2.8      5.1      1.5
## 135      6.1      2.6      5.6      1.4
## 136      7.7      3.0      6.1      2.3
## 137      6.3      3.4      5.6      2.4
## 138      6.4      3.1      5.5      1.8
## 139      6.0      3.0      4.8      1.8
## 140      6.9      3.1      5.4      2.1
## 141      6.7      3.1      5.6      2.4
## 142      6.9      3.1      5.1      2.3
## 143      5.8      2.7      5.1      1.9
## 144      6.8      3.2      5.9      2.3
## 145      6.7      3.3      5.7      2.5
## 146      6.7      3.0      5.2      2.3
## 147      6.3      2.5      5.0      1.9
## 148      6.5      3.0      5.2      2.0
## 149      6.2      3.4      5.4      2.3
## 150      5.9      3.0      5.1      1.8
```

```
# Apply K-mean clustering to understand the Species from other attributes
?kmeans
```

```
## starting httpd help server ... done
```

```
kmeans.result <- kmeans(iris2, centers = 3, nstart = 20)
kmeans.result
```

```
## K-means clustering with 3 clusters of sizes 50, 38, 62
```

```
##
```

```
## Cluster means:
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1    5.006000    3.428000    1.462000    0.246000
## 2    6.850000    3.073684    5.742105    2.071053
## 3    5.901613    2.748387    4.393548    1.433871
```

```
##
```

```
## Clustering vector:
```

```
##   [1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
##  [38] 1 1 1 1 1 1 1 1 1 1 1 1 1 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
##  [75] 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 3 2 2 2 3 2 2 2
## [112] 2 2 3 3 2 2 2 2 3 2 3 2 3 2 2 3 3 2 2 2 2 3 2 2 2 3 2 2 3 2 2 3 2
## [149] 2 3
```

```
##
```

```
## Within cluster sum of squares by cluster:
```

```
## [1] 15.15100 23.87947 39.82097
```

```
## (between_SS / total_SS = 88.4 %)
```

```
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"       "
```

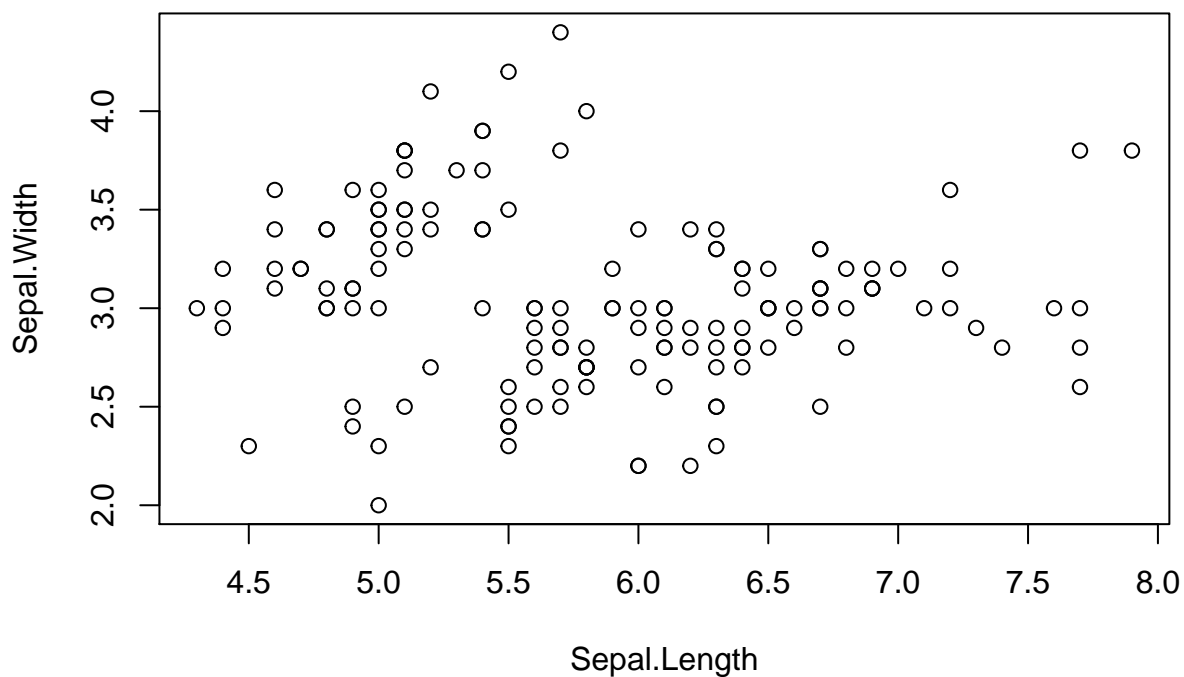
```
# See the cluster identification for each observation
kmeans.result$cluster
```

```
## [1] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [38] 1 1 1 1 1 1 1 1 1 1 1 1 1 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [75] 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 3 2 2 2 2 3 2 2 2
## [112] 2 2 3 3 2 2 2 2 3 2 3 2 3 2 2 2 2 2 3 2 2 2 2 3 2 2 2 3 2 2 2 3 2
## [149] 2 3
```

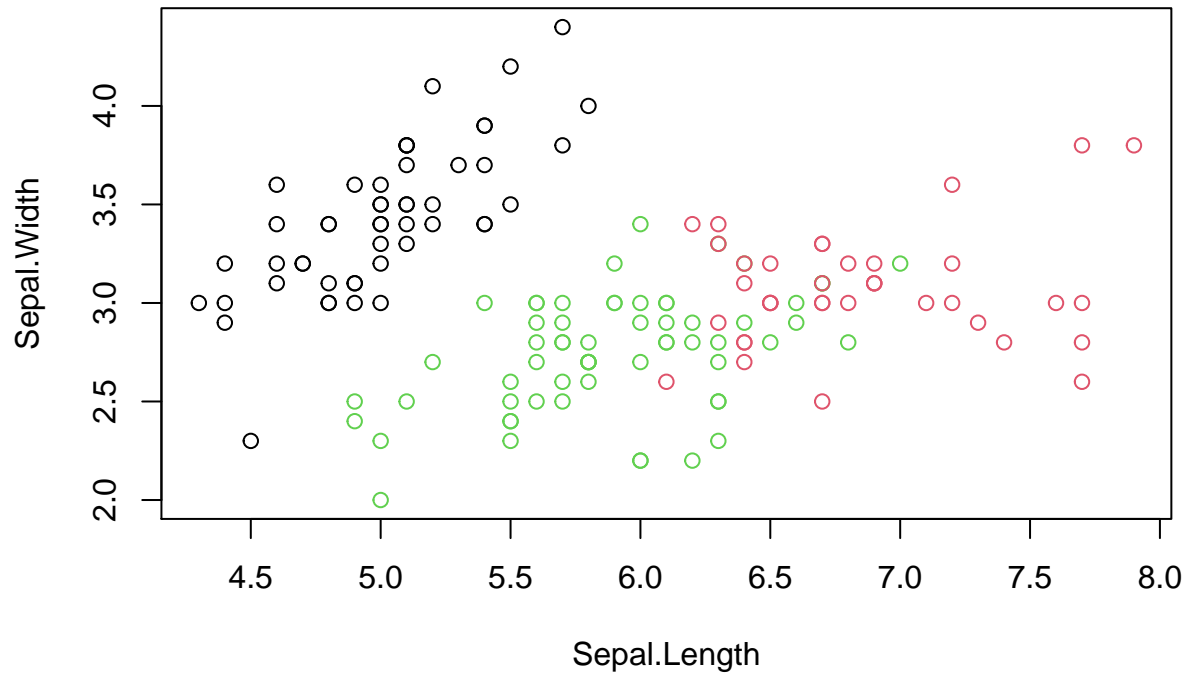
```
# Compare with original label
table(iris$Species, kmeans.result$cluster)
```

```
##
##           1  2  3
## setosa    50  0  0
## versicolor 0  2 48
## virginica  0 36 14
```

```
# Visualizing and interpreting results of k-means()
plot(iris2[c("Sepal.Length", "Sepal.Width")])
```

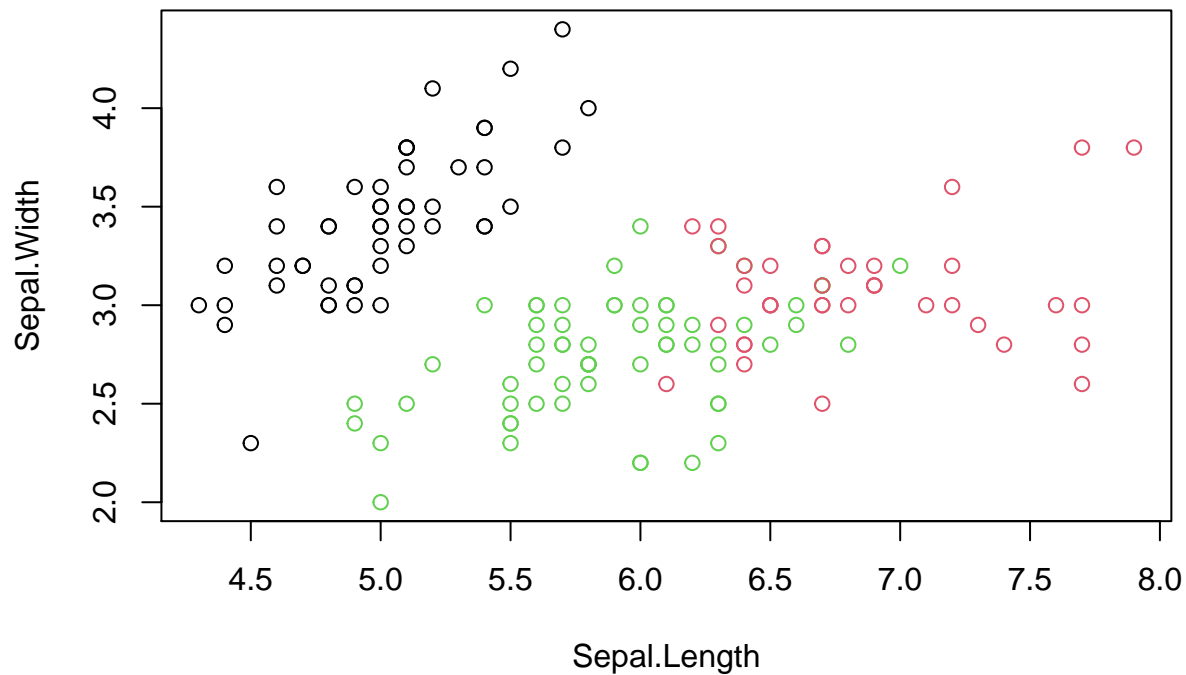


```
plot(iris2[c("Sepal.Length", "Sepal.Width")], col = kmeans.result$cluster)
```



```
plot(iris2[c("Sepal.Length", "Sepal.Width")], col = kmeans.result$cluster, main = "K-Means with 3 clusters")
```

K-Means with 3 clusters



```
# plot cluster centers
kmeans.result$centers
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1    5.006000    3.428000    1.462000    0.246000
## 2    6.850000    3.073684    5.742105    2.071053
## 3    5.901613    2.748387    4.393548    1.433871
```

```
kmeans.result$centers[,c("Sepal.Length", "Sepal.Width")]
```

```
##   Sepal.Length Sepal.Width
## 1    5.006000    3.428000
## 2    6.850000    3.073684
## 3    5.901613    2.748387
```

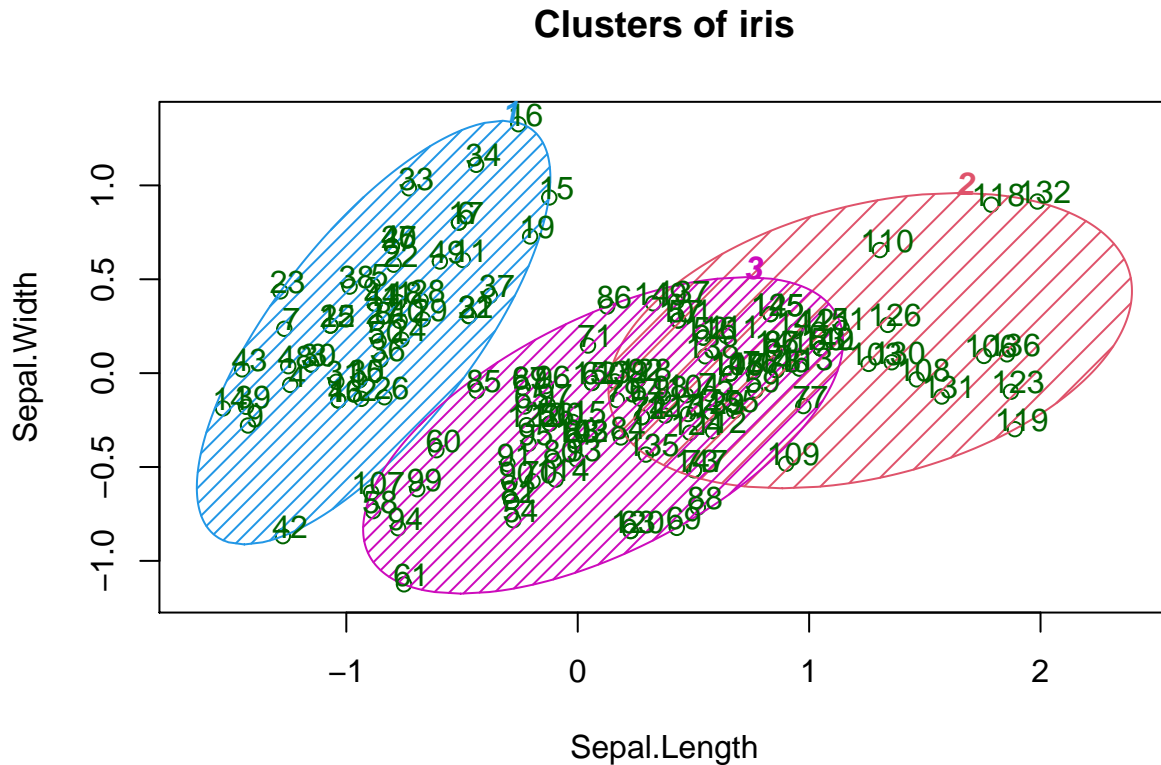
```
#points(kmeans.result$centers[,c("Sepal.Length", "Sepal.Width")], col = 1:3, pch = 8, cex=3)
```

```
# Visualising the clusters
library(cluster)
y_kmeans <- kmeans.result$cluster
?clusplot()
clusplot(iris2[,c("Sepal.Length", "Sepal.Width")],
         y_kmeans,
         lines = 0,
```

```

shade = TRUE,
color = TRUE,
labels = 2,
plotchar = FALSE,
span = TRUE,
main = paste('Clusters of iris'),
xlab = 'Sepal.Length',
ylab = 'Sepal.Width')

```



These two components explain 100 % of the point variability.

```

# Determining number of clusters
# See Total within sum of square error
kmeans.result$tot.withinss

```

```
## [1] 78.85144
```

```

# Initialize total within cluster sum of squares error: wcss
set.seed(6)
wcss = vector()

```

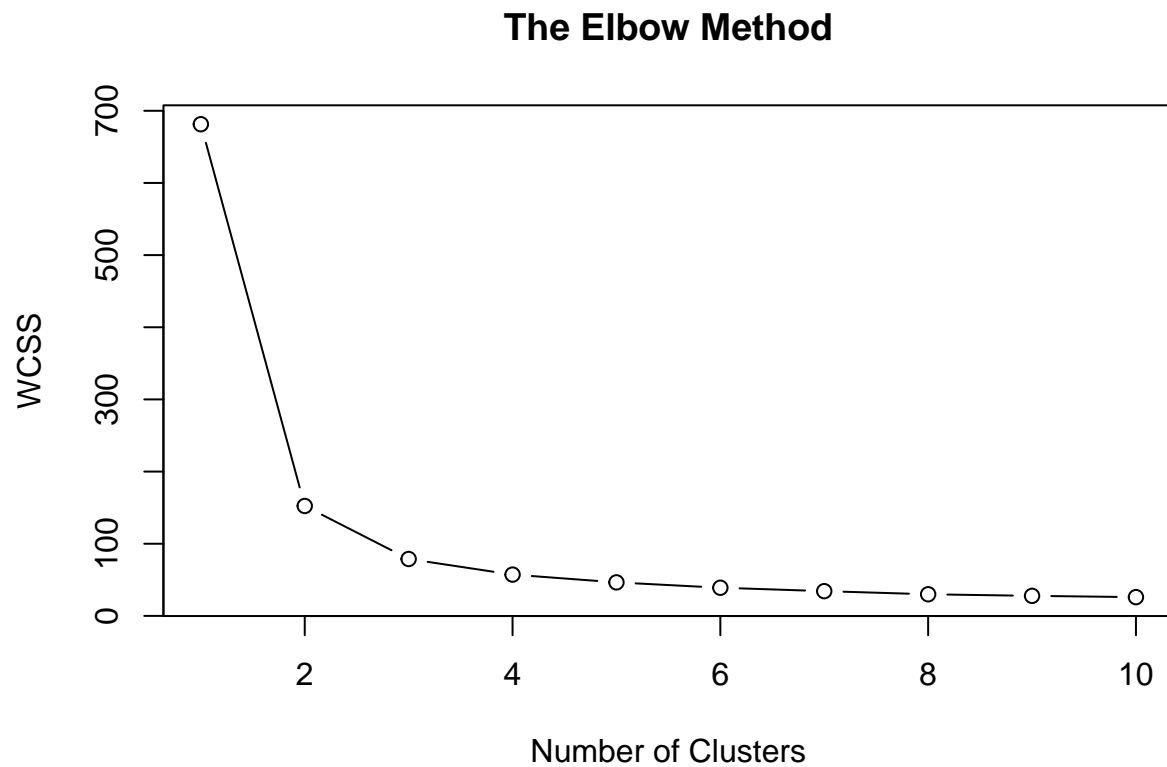
```

# For 1 to 15 cluster centers check the WCSS
for (i in 1:10) wcss[i] = sum(kmeans(iris2, centers = i, nstart = 20)$withinss)
# Plot WSS vs. number of clusters
plot(1:10,
     wcss,

```



```
type = "b",  
main = paste('The Elbow Method'),  
xlab = "Number of Clusters",  
ylab = "WCSS")
```



```
# Set k equal to the number of clusters corresponding to the elbow location  
k <- 3  
  
# Fitting K-Means to the dataset  
set.seed(29)  
kmeans.result <- kmeans(iris2, centers = 3, nstart = 20)  
y_kmeans = kmeans.result$cluster
```