## Ex.No.6 k Means Clustering Algorithm

**Date**: 5-09-23

#### **Aim**

To implement k Means Clustering Algorithm under unsupervised machine learning by grouping the data by similarity through R programming.

#### **Procedure**

- 1. To do programming in R, first install "RStudio" and "R" in the system. RStudio is an integrated development environment [IDE] for R and python.
- 2. Select the File in taskbar  $\rightarrow$  open New file  $\rightarrow$ R script or use shortcut "ctrl+shift+N"
- 3. Install the 'ClusterR, cluster' package and load it in R.
- 4. Import the built-in dataset 'iris'
- 5. Apply the k Means Algorithm on the iris dataset.
- 6. Write the program in the script and save it using the extension R.
- 7. Run the program by clicking Run option or use the shortcut "ctrl+enter".
- 8. See the output in the console tab.

### **Concepts Involved**

• Applying the k Means Clustering algorithm on the data set.

### k-Means Clustering Algorithm

k Means Clustering in R programming is an unsupervised Non-linear algorithm that cluster data based on similarity. Segmentation of data takes place to assign each training example to a segment called as cluster. It is used in a variety of fields like Banking, healthcare, retail, Media etc.,

#### Algorithm:

- 1. Choose the number K clusters.
- 2. Select at random K points, the centroids (Not necessarily from the given data).
- 3. Assign each data point to closest centroid that forms K clusters.
- 4. Compute and place the new centroid of each centroid.
- 5. Reassign each data point to new cluster.
- 6. After final reassignment, name the cluster as Final cluster.

#### **Dataset**

Here, Iris dataset consisting of 50 samples from each of 3 species Iris (Iris setosa, Iris virginica, Iris versicolor) is used. Four features were measured from each sample i.e., length and width of the sepals and petals and based on a combination of these four features.

#### **Script**

```
# Installing Packages
install.packages("ClusterR")
install.packages("cluster")
# Loading package
library(ClusterR)
library(cluster)
# Removing initial label of
# Species from original dataset
iris 1 < -iris[, -5]
# Fitting K-Means clustering Model
# to training dataset
set.seed(240) # Setting seed
kmeans.re <- kmeans(iris 1, centers = 3, nstart = 20)
kmeans.re
Output
K-means clustering with 3 clusters of sizes 50, 62, 38
Cluster means:
  Sepal.Length Sepal.Width Petal.Length Petal.Width
      5.006000
                               1.462000
                                           0.246000
                  3.428000
2
      5.901613
                  2.748387
                               4.393548
                                           1.433871
3
      6.850000
                  3.073684
                               5.742105
                                           2.071053
Clustering vector:
  1 1 1 1 1 1 1 1 1 1
                   1
                     1
                        1
                          1
                           1
                 2
                     2
           1 1
               2
                   3
                        2
                          [46]
     1
       1 1
2 3 2 3 2 3 3 2
[136] 3 3 3 2 3
                2
                 3
                    3
                      3
                        3
                          3
                           2
                3
                 3 2
                      3
                       3 3 2
Within cluster sum of squares by cluster:
[1] 15.15100 39.82097 23.87947
 (between_SS / total_SS = 88.4 %)
Available components:
```

```
[1] "cluster" "centers" "totss" "withinss"
"tot.withinss" "betweenss"
[7] "size" "iter" "ifault"
```

# **Script**

# Cluster identification for

# each observation

kmeans.re\$cluster

#### Output

# **Script**

# Confusion Matrix

cm <- table(iris\$Species, kmeans.re\$cluster)

cm

#### Output

#### Script

```
# Model Evaluation and visualization
plot(iris_1[c("Sepal.Length", "Sepal.Width")])
plot(iris_1[c("Sepal.Length", "Sepal.Width")],
    col = kmeans.re$cluster)
plot(iris_1[c("Sepal.Length", "Sepal.Width")],
    col = kmeans.re$cluster,
```

```
main = "K-means with 3 clusters")
```

## Ploting cluster centers

kmeans.re\$centers

## Output

```
Sepal.Length Sepal.Width Petal.Length Petal.Width15.0060003.4280001.4620000.24600025.9016132.7483874.3935481.43387136.8500003.0736845.7421052.071053
```

# Script

kmeans.re\$centers[, c("Sepal.Length", "Sepal.Width")]

#### Output

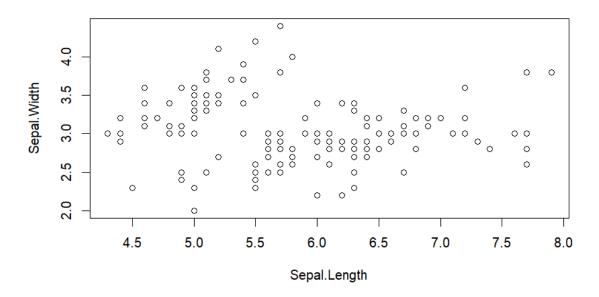
```
Sepal.Length Sepal.Width
1 5.006000 3.428000
2 5.901613 2.748387
3 6.850000 3.073684
```

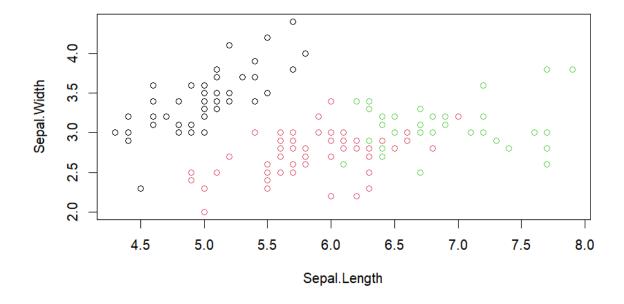
# **Script**

```
# cex is font size, pch is symbol
points(kmeans.re$centers[, c("Sepal.Length", "Sepal.Width")],
        col = 1:3, pch = 8, cex = 3)
## Visualizing clusters
y_kmeans <- kmeans.re$cluster
clusplot(iris_1[, c("Sepal.Length", "Sepal.Width")],
        y_kmeans,
        lines = 0,
        shade = TRUE,
        color = TRUE,
        labels = 2,
        plotchar = FALSE,
        span = TRUE,
        main = paste("Cluster iris"),</pre>
```

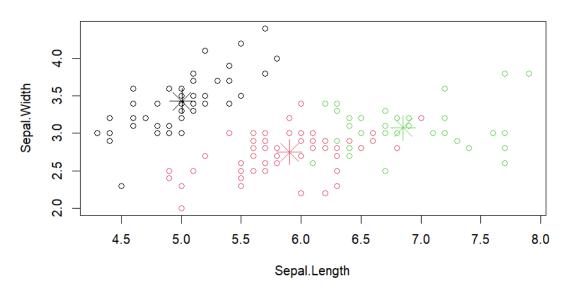
xlab = 'Sepal.Length',
ylab = 'Sepal.Width')

# Output

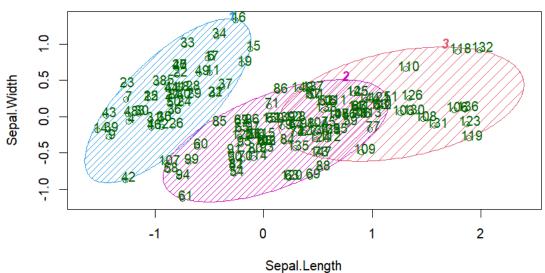




## K-means with 3 clusters



# **Cluster iris**



These two components explain 100 % of the point variability.

# Result

Thus the k Means Clustering Algorithm has been successfully implemented on Iris data set using R programming.