

Exp. No : 9

## Implement clustering techniques – Hierarchical and K-Means

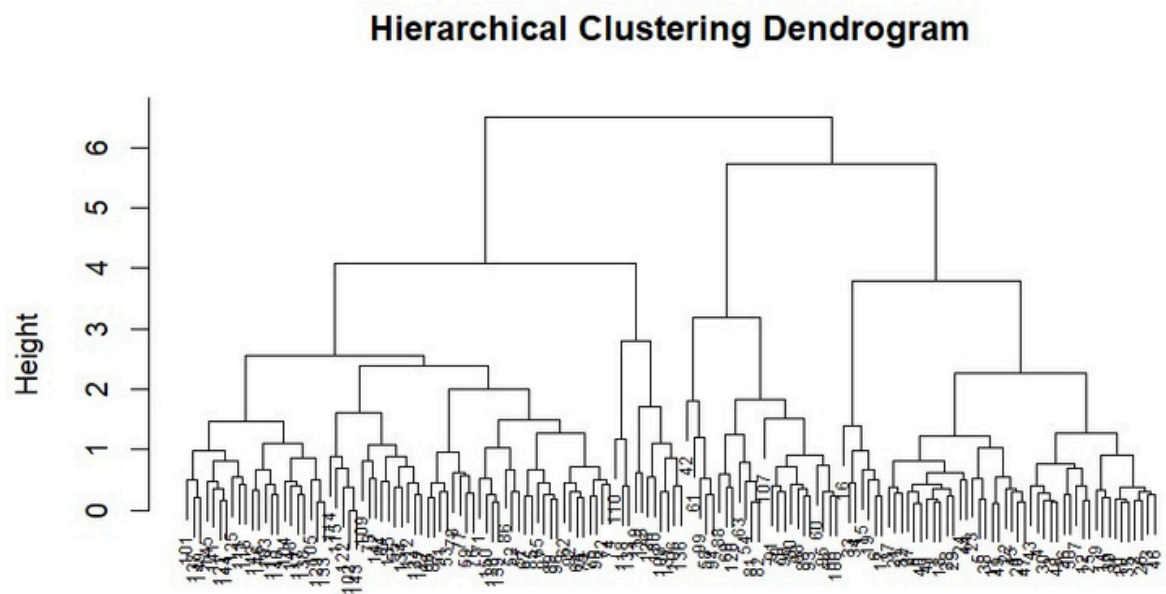
### a) Hierarchical Clustering

```
# Load the iris dataset data(iris) # Use only the numeric columns for clustering (exclude the
Species column) iris_data <- iris[, -5] # Standardize the data iris_scaled <- scale(iris_data) #
Compute the distance matrix distance_matrix <- dist(iris_scaled, method = "euclidean") #
Perform hierarchical clustering using the "complete" linkage method hc_complete <-
hclust(distance_matrix, method = "complete") # Plot the dendrogram plot(hc_complete, main =
"Hierarchical Clustering Dendrogram", xlab = "", sub = "", cex =
0.6)
# Cut the tree to form 3 clusters
clusters <- cutree(hc_complete, k = 3)
# Print the cluster memberships
print(clusters)
# Add the clusters to the original dataset
iris$Cluster <- as.factor(clusters)
# Display the first few rows of the updated dataset
head(iris)
```

**Output :**

```
> # Load the iris dataset  
> data(iris)  
> # Use only the numeric columns for clustering (exclude the species column)  
> iris_data <- iris[, -5]  
> # Standardize the data  
> iris_scaled <- scale(iris_data)  
> # Compute the distance matrix  
> distance_matrix <- dist(iris_scaled, method = "euclidean")  
> # Perform hierarchical clustering using the "complete" linkage method  
> hc_complete <- hclust(distance_matrix, method = "complete")  
> # Plot the dendrogram  
> plot(hc_complete, main = "Hierarchical Clustering Dendrogram", xlab = "", sub  
"", cex =  
+      0.6)  
> # Cut the tree to form 3 clusters  
> clusters <- cutree(hc_complete, k = 3)  
> # Print the cluster memberships  
> print(clusters)  
[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  
[37] 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 3 3 3 2 3 2 3 2 3 2 2 3 2 3 3 3 3 2 2 2 3 3  
[73] 3 3 3 3 3 3 3 3 2 2 2 2 3 3 3 3 2 3 2 2 3 2 2 2 3 3 3 2 2 3 3 3 3 3 3 2 3  
[109] 3 3 3 3 3 3 3 3 3 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3  
[145] 3 3 3 3 3 3  
> # Add the clusters to the original dataset  
> iris$Cluster <- as.factor(clusters)  
> # Display the first few rows of the updated dataset  
> head(iris)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	Cluster
1	5.1	3.5	1.4	0.2	setosa	1
2	4.9	3.0	1.4	0.2	setosa	1
3	4.7	3.2	1.3	0.2	setosa	1
4	4.6	3.1	1.5	0.2	setosa	1
5	5.0	3.6	1.4	0.2	setosa	1
6	5.4	3.9	1.7	0.4	setosa	1



#### b) K-Means Clustering

```
# Load the iris dataset
data(iris)

# Use only the numeric columns for clustering (exclude the Species column)
iris_data <- iris[, -5]

# Standardize the data
iris_scaled <- scale(iris_data)

# Set the number of clusters
set.seed(123) # For reproducibility
k <- 3 # Number of clusters

# Perform K-Means clustering
kmeans_result <- kmeans(iris_scaled, centers = k, nstart = 25)

# Print the K-Means result
print(kmeans_result)

# Print the cluster centers
```

```

print(kmeans_result$centers)

# Add the cluster assignments to the original dataset
iris$Cluster <- as.factor(kmeans_result$cluster)

# Display the first few rows of the updated dataset
head(iris)

# Plot the clusters

library(ggplot2)

ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Cluster)) +
geom_point(size = 3) +
labs(title = "K-Means Clustering of Iris Dataset", x = "Sepal Length", y = "Sepal Width")

```

### Output :

```

_143_ 3 3 4 3 3 4

```

```

within cluster sum of squares by cluster:

```

```

[1] 47.35062 44.08754 47.45019
(between_SS / total_SS = 76.7 %)

```

```

\available components:

```

```

[1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
[6] "betweenss"    "size"         "iter"         "ifault"       "tot.withinss"
> # Print the cluster centers
> print(kmeans_result$centers)
  Sepal.Length Sepal.Width Petal.Length Petal.Width
1 -1.01119138  0.85041372  -1.3006301  -1.2507035
2 -0.05005221 -0.88042696   0.3465767   0.2805873
3  1.13217737  0.08812645   0.9928284   1.0141287
> # Add the cluster assignments to the original dataset
> iris$Cluster <- as.factor(kmeans_result$cluster)
> # Display the first few rows of the updated dataset
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species Cluster
1           5.1          3.5          1.4          0.2  setosa        1
2           4.9          3.0          1.4          0.2  setosa        1
3           4.7          3.2          1.3          0.2  setosa        1
4           4.6          3.1          1.5          0.2  setosa        1
5           5.0          3.6          1.4          0.2  setosa        1
6           5.4          3.9          1.7          0.4  setosa        1
> # Plot the clusters
> library(ggplot2)
> ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Cluster)) +
+   geom_point(size = 3) +
+   labs(title = "K-Means Clustering of Iris Dataset", x = "Sepal Length", y = "S
epal width")

```

