ROLL NO: 210701296

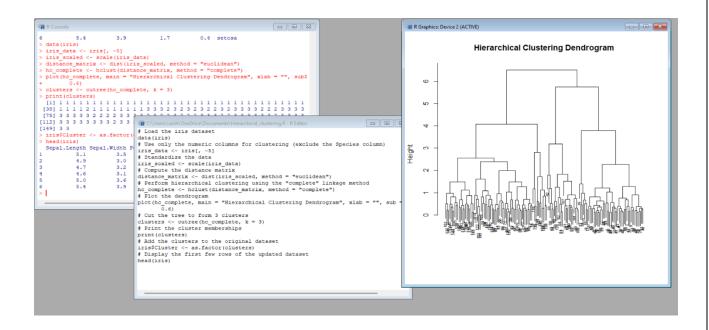
Exp:9

<u>Implement clustering techniques – Hierarchical and K-Means</u>

a) HIERARCHIAL 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)</pre>
# Compute the distance matrix
distance_matrix <- dist(iris_scaled, method = "euclidean")</pre>
# Perform hierarchical clustering using the "complete" linkage method
hc_complete <- hclust(distance_matrix, method = "complete")</pre>
# 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)</pre>
# Display the first few rows of the updated dataset
head(iris)
```

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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)

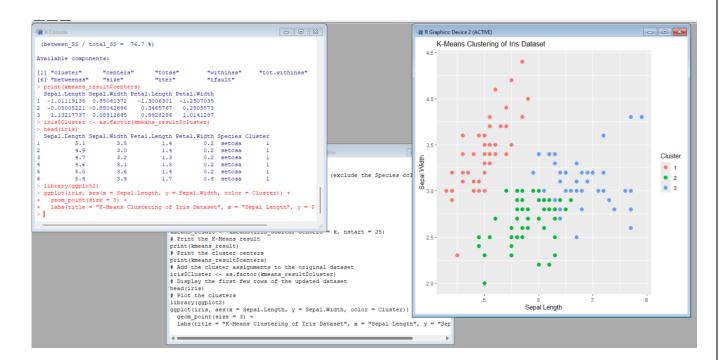
Print the K-Means result print(kmeans_result)

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")
```



RESULT:

Thus the implementation of clustering techniques like hierarchical and kmeans is successfully executed using R.