Ex No 9

Implement clustering techniques – Hierarchical and K-Means

AIM:

To Implement clustering techniques – Hierarchical and K-Means using R.

PROCEDURE:

- Collect and load the dataset from sources like CSV files or databases.
- Clean and preprocess the data, including handling missing values and scaling features.
- Determine the number of clusters (K) for K-Means, or decide on the stopping criterion for Hierarchical Clustering.
- Choose the appropriate clustering algorithm: K-Means for partitioning, Hierarchical for nested clustering.
- Apply the K-Means algorithm using fit_predict to assign data points to clusters.
- Apply the Hierarchical Clustering algorithm using Agglomerative Clustering for hierarchical clusters.
- Visualize the clusters with scatter plots for K-Means, and dendrograms for Hierarchical Clustering.
- Evaluate clustering performance using metrics like silhouette score or inertia (for KMeans).
- Fine-tune the clustering by adjusting the number of clusters or linkage criteria.
- Interpret the results to understand the structure and relationships within the data.

CODE:

Hierarchical Clustering.R:

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)
```

K-Means Clustering.R:

```
# 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) #

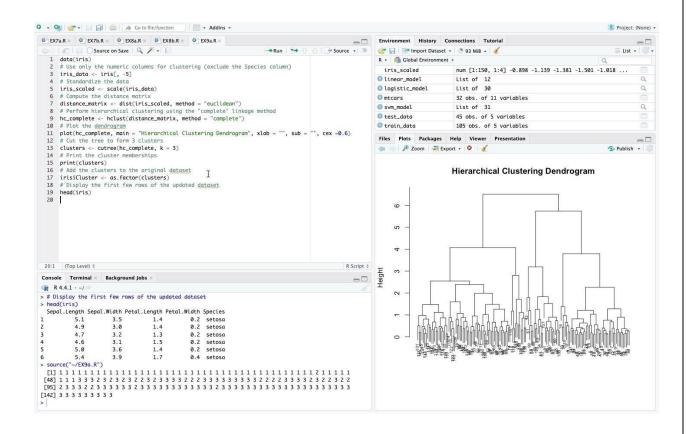
Set the number of clusters set.seed(123) # For reproducibility k <- 3 # Number

of clusters # Perform K-Means clustering kmeans_result <-

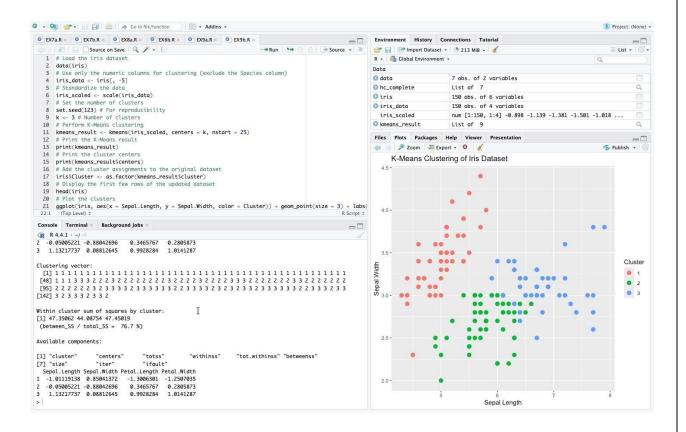
kmeans(iris_scaled, centers = k, nstart = 25)
```

```
# 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:
```

Hierarchical Clustering:



K-Means Clustering:



RESULT:

Thus, to implement clustering techniques – Hierarchical and K-Means using R has been successfully executed.