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 K-Means).
- 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)

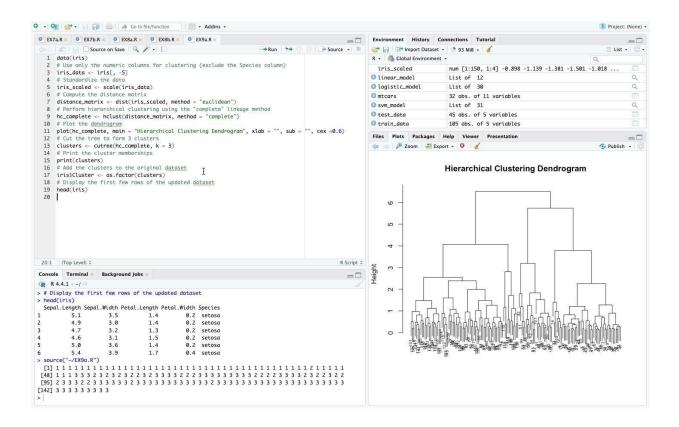
Compute the distance matrix

```
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$Cluster <- as.factor(clusters)</pre>
# Display the first few rows of the updated dataset head(iris)
K-Means Clustering.R:
# Load the iris dataset data(iris)
# Use only the numeric columns for clustering (exclude the Species column)
iris_data <- iris[, -5] # Standardize the data iris_scaled <- scale(iris_data) #
Set the number of clusters set.seed(123) \# For reproducibility k <- 3 \#
Number of clusters # Perform K-Means clustering
kmeans_result <- kmeans(iris_scaled, centers = k, nstart = 25)
# Print the K-Means result
print(kmeans_result) # Print
the cluster centers
print(kmeans_result$centers)
```

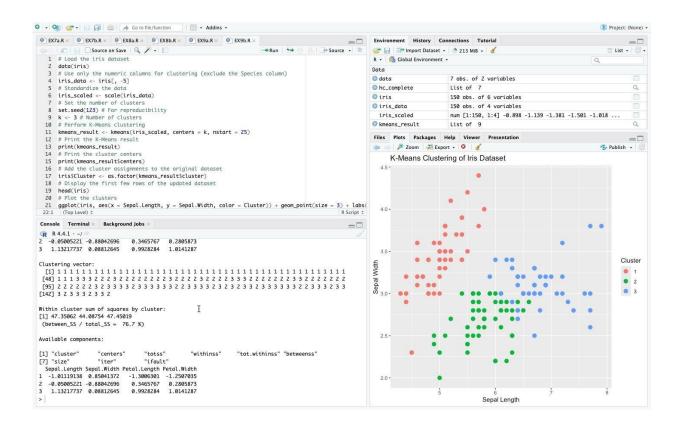
```
# Add the cluster assignments to the original dataset
iris$Cluster <- as.factor(kmeans_result$cluster) #
Display the first few rows of the updated dataset
head(iris) # Plot the clusters library(ggplot2)
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Cluster)) +
geom_point(size = 3) +
labs(title = "K-Means Clustering of Iris Dataset", x = "Sepal Length", y = "Sepal Width")
```

OUTPUT:

Hierarchical Clustering:



K-Means Clustering:



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	210701206
RESULT:	
Thus, to implement clustering techniques — Hierarchical successfully executed.	l and K-Means using R has been