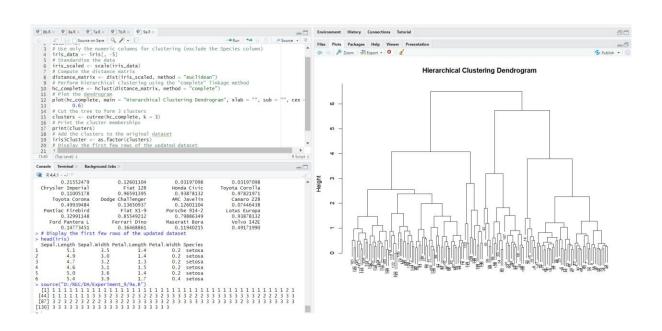
EXP NO: 9

IMPLEMENT CLUSTERING TECHNIQUES – HIERARCHICAL AND K-MEANS

a) Hierarchial Clustering

```
# Load the iris dataset
data(iris)
# Use only the numeric columns for clustering (exclude the Species
column)
iris_data <- iris[, -5]</pre>
# 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)</pre>
# 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)
```

Output:



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]</pre>
# Standardize the data
iris_scaled <- scale(iris_data)</pre>
# 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)</pre>
# 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)</pre>
# 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:

