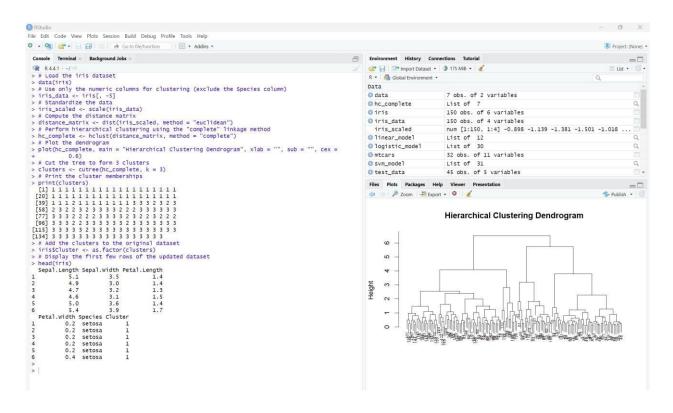
Exp:9

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] # 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)
```



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)</pre>

Set the number of clusters

set.seed(123) # For reproducibility

 $k \le 3 \# Number of clusters$

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)</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")

