Implement clustering techniques – Hierarchical and K-Means

# HIERARCHIAL CLUSTERING

**CODE:**

# 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

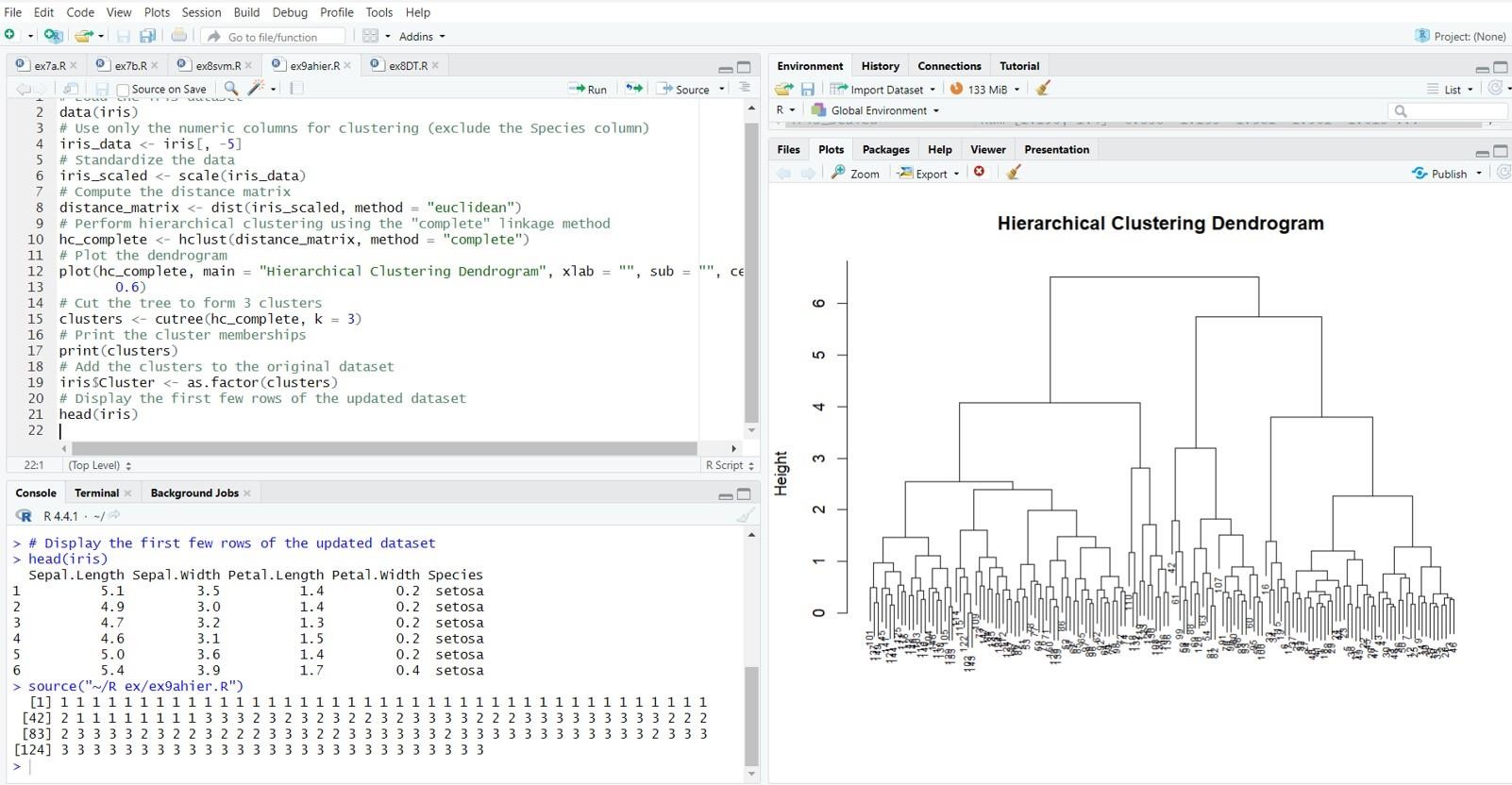
plot(hc\_complete, main = "Hierarchical Clustering Dendrogram", xlab = "", sub = "", cex = 0.6) # Cut the tree to form 3 clusters

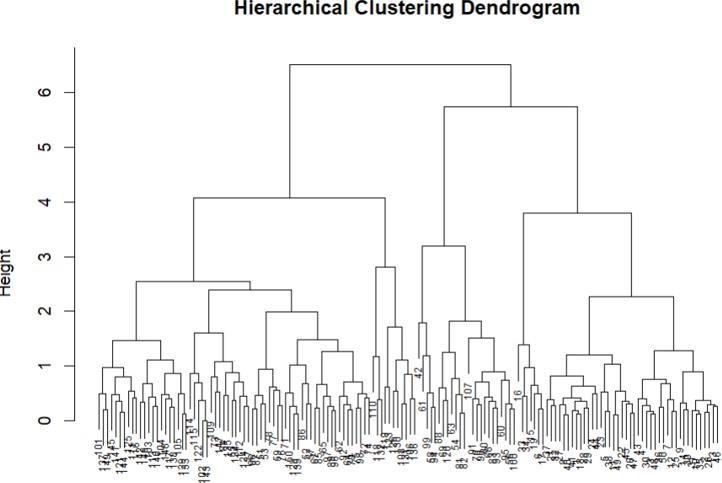
clusters <- cutree(hc\_complete, k = 3) # Print the cluster memberships print(clusters)

# Add the clusters to the original dataset iris$Cluster <- as.factor(clusters)

# Display the first few rows of the updated dataset head(iris)

# OUTPUT:





1. **K-MEANS CLUSTERING**

# CODE:

# 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)

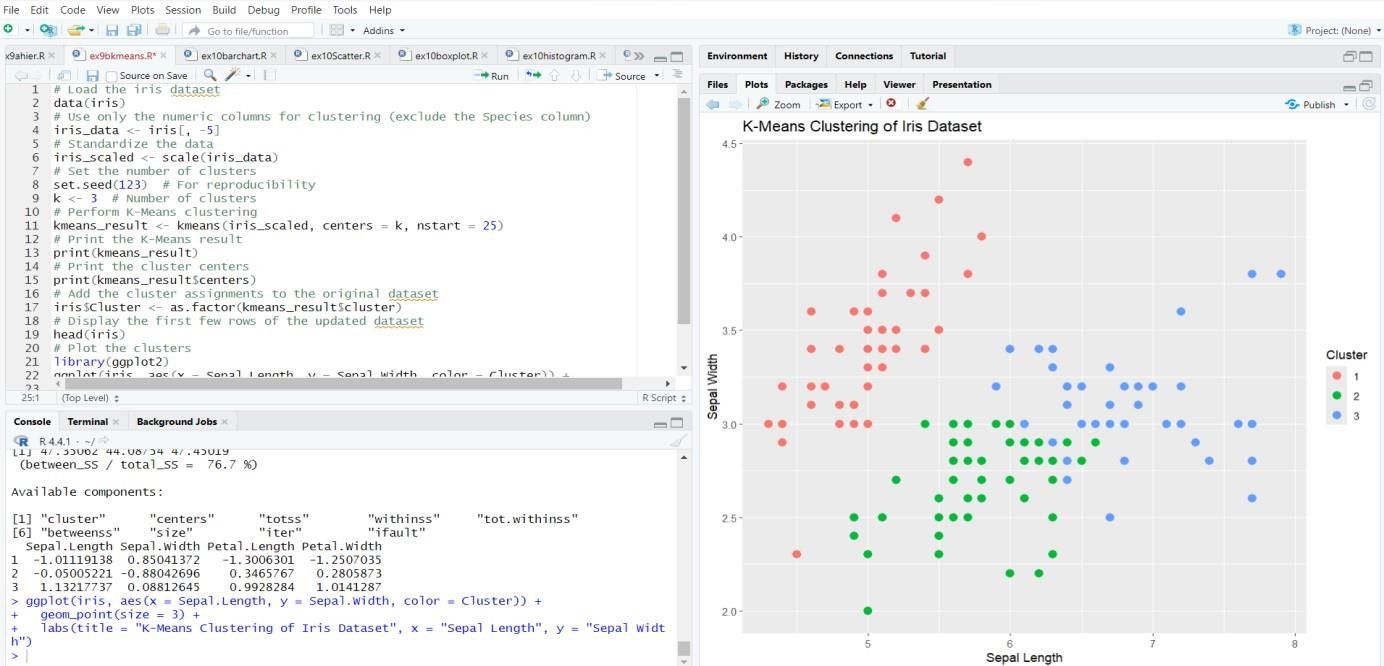
# Add the cluster assignments to the original dataset iris$Cluster <- as.factor(kmeans\_result$cluster)

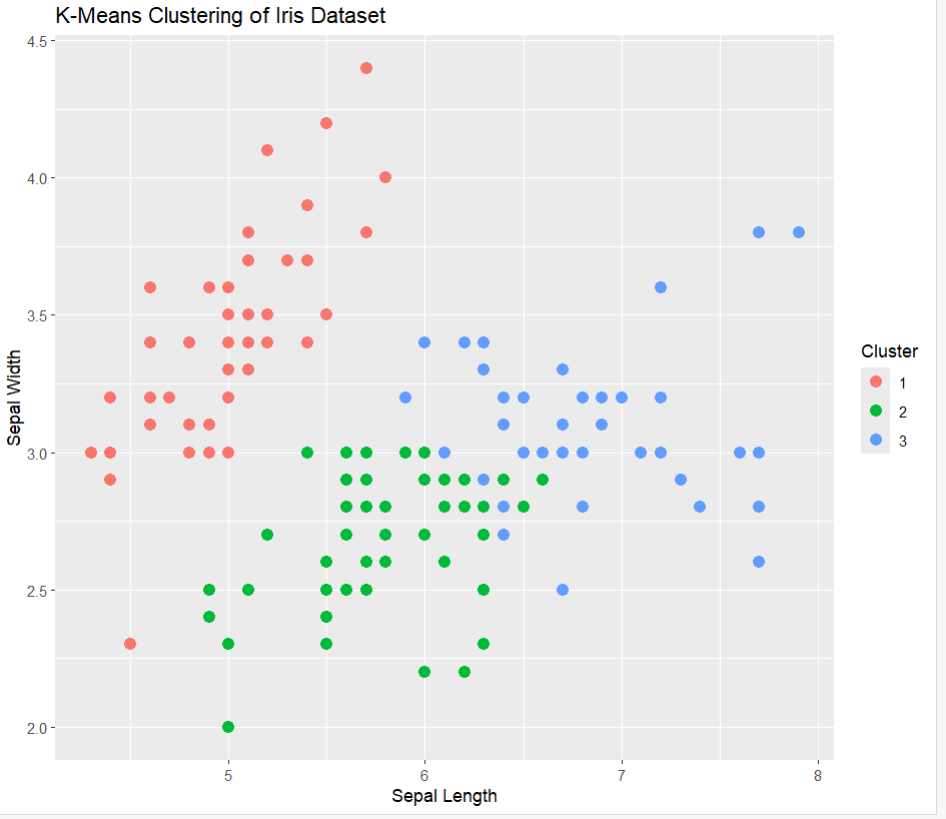
# Display the first few rows of the updated dataset head(iris)

# 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 clustering techniques such as Hierarchical and K-Means clustering implemented successfully.