$\mathbf{Q9}.\mathbf{R}$

Manoj

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```
# Load necessary libraries
library(datasets)
library(cluster)
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.3.3

# Load the iris dataset
data(iris)

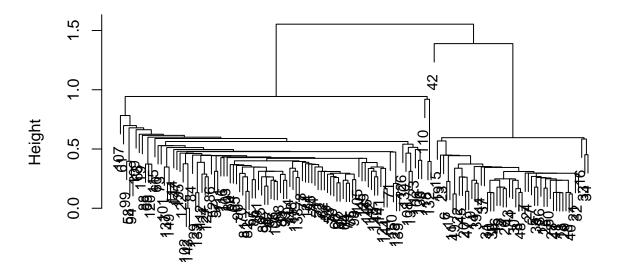
# Extract the first four variables
iris_data <- iris[, 1:4]

# Standardize the data
iris_data <- scale(iris_data)

# a) Fit hierarchical clustering model using single linkage
hc_single <- hclust(dist(iris_data), method = "single")

# Plot the dendrogram for single linkage
plot(hc_single, main = "Dendrogram (Single Linkage)", xlab = "", sub = "", cex = 0.9)</pre>
```

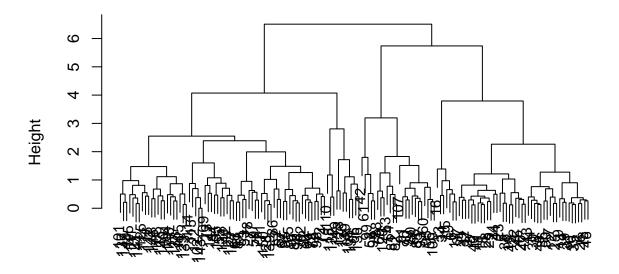
Dendrogram (Single Linkage)



```
# b) Fit hierarchical clustering model using complete linkage
hc_complete <- hclust(dist(iris_data), method = "complete")

# Plot the dendrogram for complete linkage
plot(hc_complete, main = "Dendrogram (Complete Linkage)", xlab = "", sub = "", cex = 0.9)</pre>
```

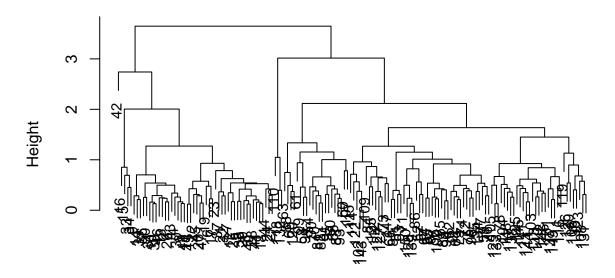
Dendrogram (Complete Linkage)



```
# c) Fit hierarchical clustering model using average linkage
hc_average <- hclust(dist(iris_data), method = "average")

# Plot the dendrogram for average linkage
plot(hc_average, main = "Dendrogram (Average Linkage)", xlab = "", sub = "", cex = 0.9)</pre>
```

Dendrogram (Average Linkage)



```
# d) Find the best hierarchical clustering model and locate the number of clusters
# Use the silhouette method to find the optimal number of clusters for each model
sil_width_single <- silhouette(cutree(hc_single, k = 3), dist(iris_data))
sil_width_complete <- silhouette(cutree(hc_complete, k = 3), dist(iris_data))
sil_width_average <- silhouette(cutree(hc_average, k = 3), dist(iris_data))

# Calculate average silhouette width for each method
avg_sil_width_single <- mean(sil_width_single[, 3])
avg_sil_width_complete <- mean(sil_width_complete[, 3])
avg_sil_width_average <- mean(sil_width_average[, 3])

# Determine the best model
best_model <- which.max(c(avg_sil_width_single, avg_sil_width_complete, avg_sil_width_average))
# Display the best model and the corresponding number of clusters
cat("Best hierarchical clustering model is: ")</pre>
```

```
if (best_model == 1) {
  cat("Single Linkage\n")
} else if (best_model == 2) {
  cat("Complete Linkage\n")
} else {
```

```
cat("Average Linkage\n")
}
```

Single Linkage

```
# Determine the optimal number of clusters for the best model using silhouette method

optimal_clusters <- function(hc) {
    sil_widths <- sapply(2:10, function(k) {
        mean(silhouette(cutree(hc, k), dist(iris_data))[, 3])
    })
    which.max(sil_widths) + 1
}

optimal_clusters_single <- optimal_clusters(hc_single)
optimal_clusters_complete <- optimal_clusters(hc_complete)
optimal_clusters_average <- optimal_clusters(hc_average)

cat("Optimal number of clusters for the best model: ")</pre>
```

Optimal number of clusters for the best model:

```
if (best_model == 1) {
   cat(optimal_clusters_single, "\n")
} else if (best_model == 2) {
   cat(optimal_clusters_complete, "\n")
} else {
   cat(optimal_clusters_average, "\n")
}
```

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