#### **WEEK 10**

# PROBLEM DEFINATION:

**CLUSTERING MODEL** 

e. Clustering algorithms for unsupervised classification.

Plot the cluster data using R visualizations

#### **SOURCE CODE:**

1. Clustering algorithms for unsupervised classification.

library(cluster)

- > set.seed(20)
- > irisCluster <- kmeans(iris[, 3:4], 3, nstart = 20)
- # nstart = 20. This means that R will try 20 different random starting assignments and then select the one with the lowest within cluster variation.
- > irisCluster

### **OUTPUT:**

Petal.Length Petal.Width

- 1 1.462000 0.246000
- 2 4.269231 1.342308
- 3 5.595833 2.037500

# **Clustering vector:**

### Within cluster sum of squares by cluster:

```
[1] 2.02200 13.05769 16.29167 (between_SS / total_SS = 94.3 %)
```

# **Available components:**

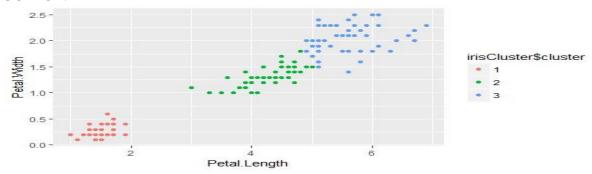
- [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
- [6] "betweenss" "size" "iter" "ifault"

### **SOURCE CODE:**

> irisCluster\$cluster <- as.factor(irisCluster\$cluster)

> ggplot(iris, aes(Petal.Length, Petal.Width, color = irisCluster\$cluster)) + geom\_point()

# **OUTPUT:**



### **SOURCE CODE:**

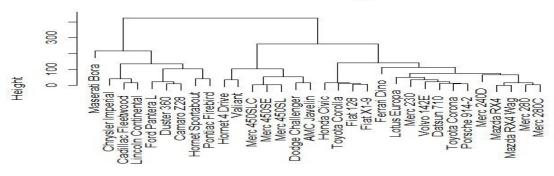
> d <- dist(as.matrix(mtcars)) # find distance matrix

> hc <- hclust(d) # apply hirarchical clustering

> plot(hc) # plot the dendrogram

# **OUTPUT:**

### Cluster Dendrogram



d hclust (\*, "complete")

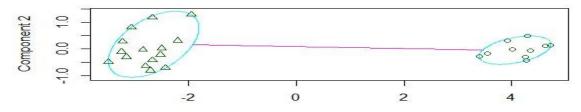
# 2. Plot the cluster data using R visualizations.

### **SOURCE CODE**:

## generate 25 objects, divided into 2 clusters.  $x \leftarrow rbind(cbind(rnorm(10,0,0.5), rnorm(10,0,0.5)), cbind(rnorm(15,5,0.5), rnorm(15,5,0.5))) clusplot(pam(x, 2))$ 

# **OUTPUT:**

# clusplot(pam(x = x, k = 2))



Component 1
These two components explain 100 % of the point variability.

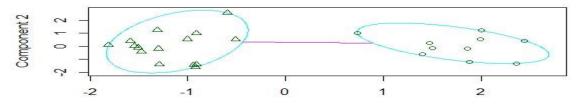
# **SOURCE CODE:**

## add noise, and try again:

x4 <- cbind(x, rnorm(25), rnorm(25))
clusplot(pam(x4, 2))</pre>

# **OUTPUT:**

# clusplot(pam(x = x4, k = 2))



Component 1
These two components explain 81.17 % of the point variability.