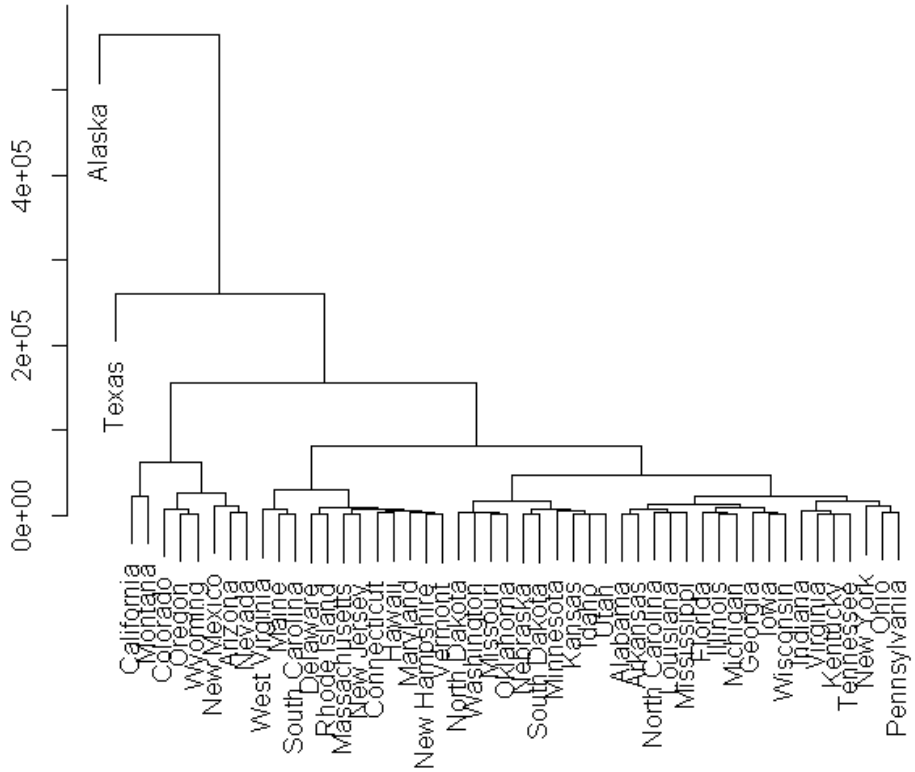
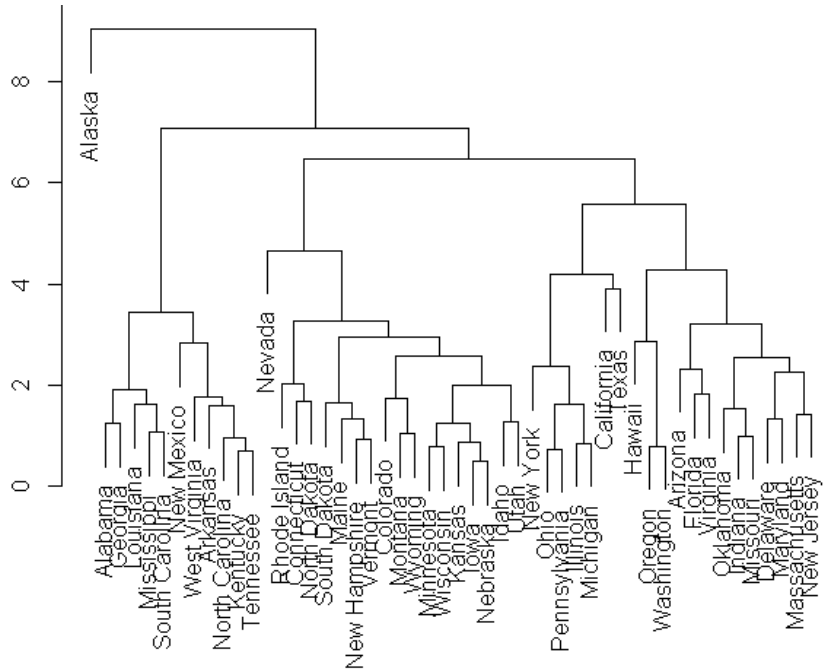
10 PROVE: CLUSTERING THE STATES

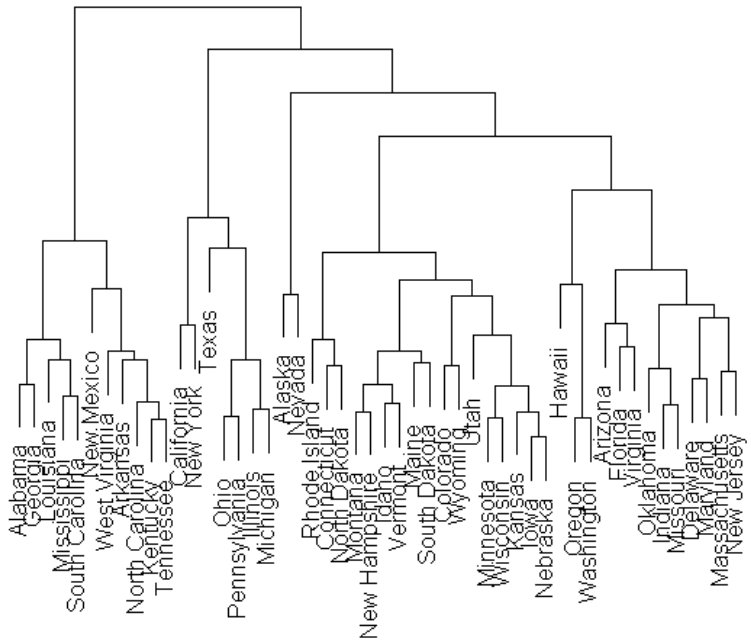
```
library(tidyverse)
library(datasets)
library(cluster)


data <- state.x77



# Computing the distance matrix
distance <- dist(as.matrix(data))

# Clustering
hc <- hclust(distance)

# Plotting a dendogram
plot(hc)
```



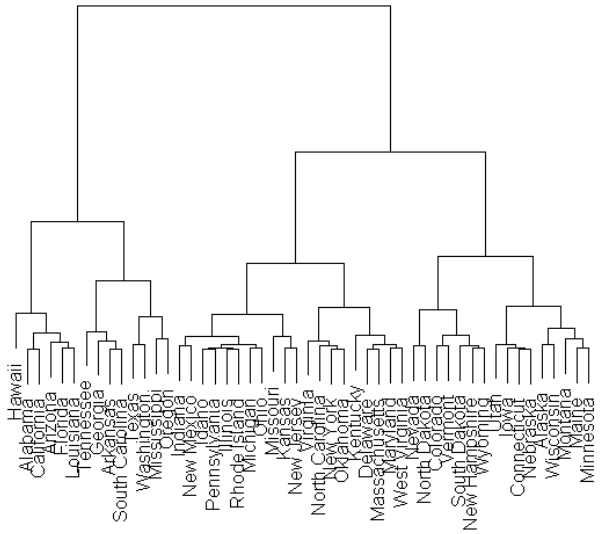
```r
# Scaling data (normalization)
data.scaled <- scale(data)

# dendogram with normalized data
distance <- dist(as.matrix(data.scaled))
hc <- hclust(distance)
plot(hc)
```



```r
# dendogram without Area
data.scaled_noArea <- scale(data[,1:7])
distance <- dist(as.matrix(data.scaled_noArea))
hc <- hclust(distance)
plot(hc)
```



```r
# dendogram only with Frost
data.scaled_Frost <- scale(data[,7])
distance <- dist(as.matrix(data.scaled_Frost))
hc <- hclust(distance)
plot(hc)
```




```r
# K-means CLustering
clust <- kmeans(data.scaled, 3)    # k = 5
summary(clust)
```

```r
# Centers of the clusters (mean values)
clust$centers
```

```
             Length Class  Mode
cluster      50     -none- numeric
centers      24     -none- numeric
totss         1     -none- numeric
withinss      3     -none- numeric
tot.withinss  1     -none- numeric
betweenss     1     -none- numeric
size          3     -none- numeric
iter          1     -none- numeric
ifault        1     -none- numeric
> clust$centers
  Population     Income   Illiteracy   Life Exp     Murder    HS Grad
1 -0.2269956 -1.3014617  1.391527063 -1.1773136  1.0919809 -1.4157826
2 -0.4873370  0.1329601 -0.641201154  0.7422562 -0.8552439  0.5515044
3  0.9462026  0.7416690  0.005468667 -0.3242467  0.5676042  0.1558335
       Frost       Area
1 -0.7206500 -0.2340290
2  0.4528591 -0.1729366
3 -0.1960979  0.4483198
```

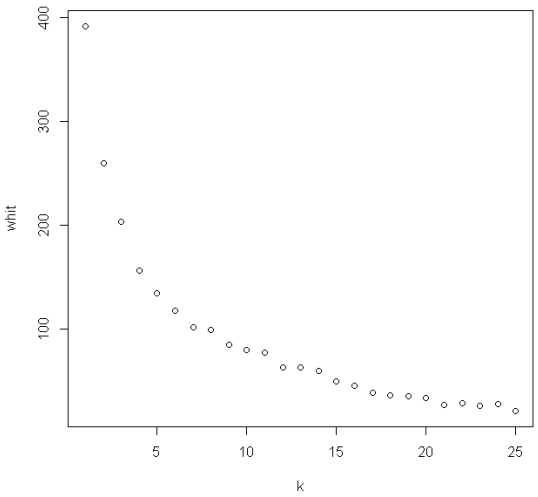
```r
# Clusters
clust$cluster
```

```
     Alabama         Alaska        Arizona       Arkansas     California
           1              3              3              1              3
    Colorado    Connecticut       Delaware        Florida        Georgia
           2              2              2              3              1
      Hawaii          Idaho       Illinois        Indiana           Iowa
           2              2              3              2              2
      Kansas       Kentucky      Louisiana          Maine       Maryland
           2              1              1              2              3
Massachusetts       Michigan      Minnesota    Mississippi       Missouri
           2              3              2              1              3
     Montana       Nebraska         Nevada  New Hampshire     New Jersey
           2              2              3              2              3
  New Mexico       New York North Carolina   North Dakota           Ohio
           1              3              1              2              3
    Oklahoma         Oregon   Pennsylvania   Rhode Island South Carolina
           2              2              3              2              1
South Dakota      Tennessee          Texas           Utah        Vermont
           2              1              3              2              2
    Virginia     Washington  West Virginia      Wisconsin        Wyoming
           3              2              1              2              2
```

```r
# Within-cluster sum of squares
clust$withinss
# Total sum of squares across clusters
clust$tot.withinss
```

```
> clust$withinss
[1]  23.62227  67.72742 111.66951
> clust$tot.withinss
[1] 203.0192
>
```

```r
# Plotting k-means clusters
clusplot(data.scaled, clust$cluster, color = T, shaed = T, labels = 2, lines = 0)

# CHOOSING NUMBER OF K USING ELBOW METHOD
whit <- c()
k <- c()
for (i in 1:25) {
    clust <- kmeans(data.scaled, i)
    whit[[i]] <- clust$tot.withinss
    k[[i]] <- i
}

elbow <- cbind(k, whit) %>%
    data.frame()
plot(elbow)
```




```
clust <- kmeans(data.scaled, 5)
clust$cluster
clusplot(data.scaled, clust$cluster, color = T, shaed = T, labels = 2, lines = 0)
```