

# DSA 8070 R Session 12: Cluster Analysis

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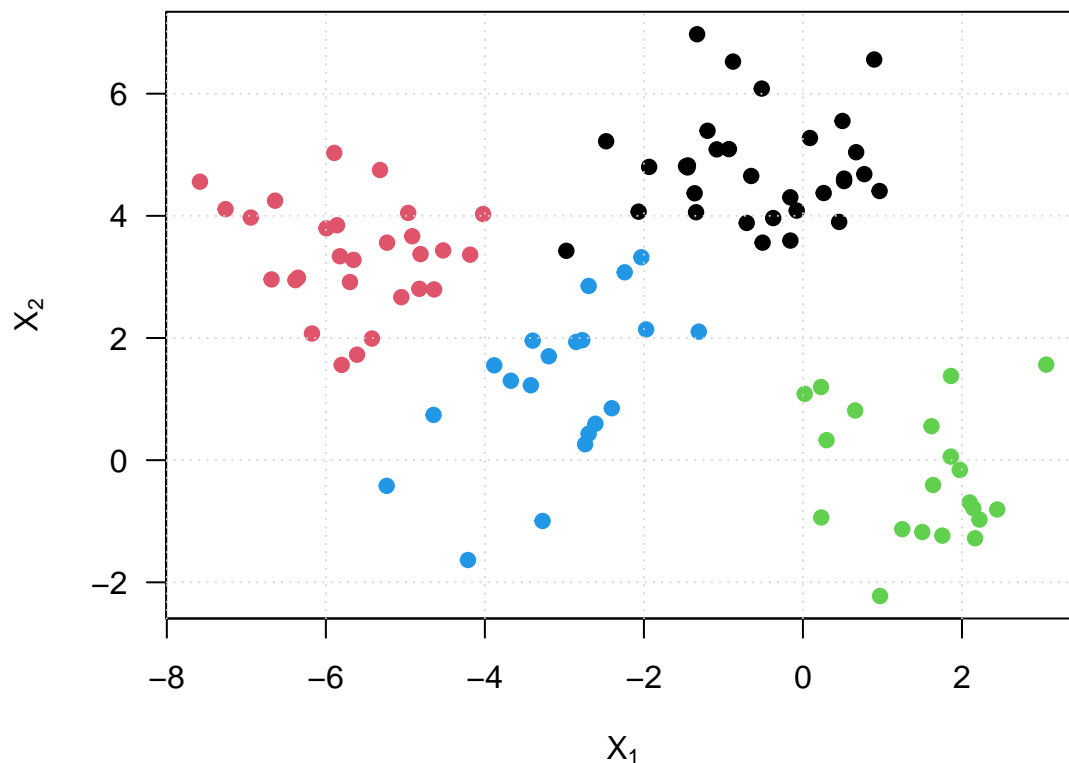
## Contents

K-Means Clustering . . . . .	1
Simulated Example . . . . .	1
Geyser Example . . . . .	3
US State Facts and Figures Example . . . . .	4
Hierarchical Clustering . . . . .	6
US State Facts and Figures Example . . . . .	6
Model-based . . . . .	9
Geyser Example . . . . .	9
Fisher's Iris Data Example . . . . .	11

## K-Means Clustering

### Simulated Example

```
set.seed(101)
library(scales)
x <- matrix(rnorm(100 * 2), 100, 2)
xmean <- matrix(rnorm(8, sd = 4), 4, 2)
which <- sample(1:4, 100, replace = TRUE)
x = x + xmean[which,]
plot(x, col = which, pch = 19, xlab = expression(X[1]),
     ylab = expression(X[2]), las = 1)
grid()
```

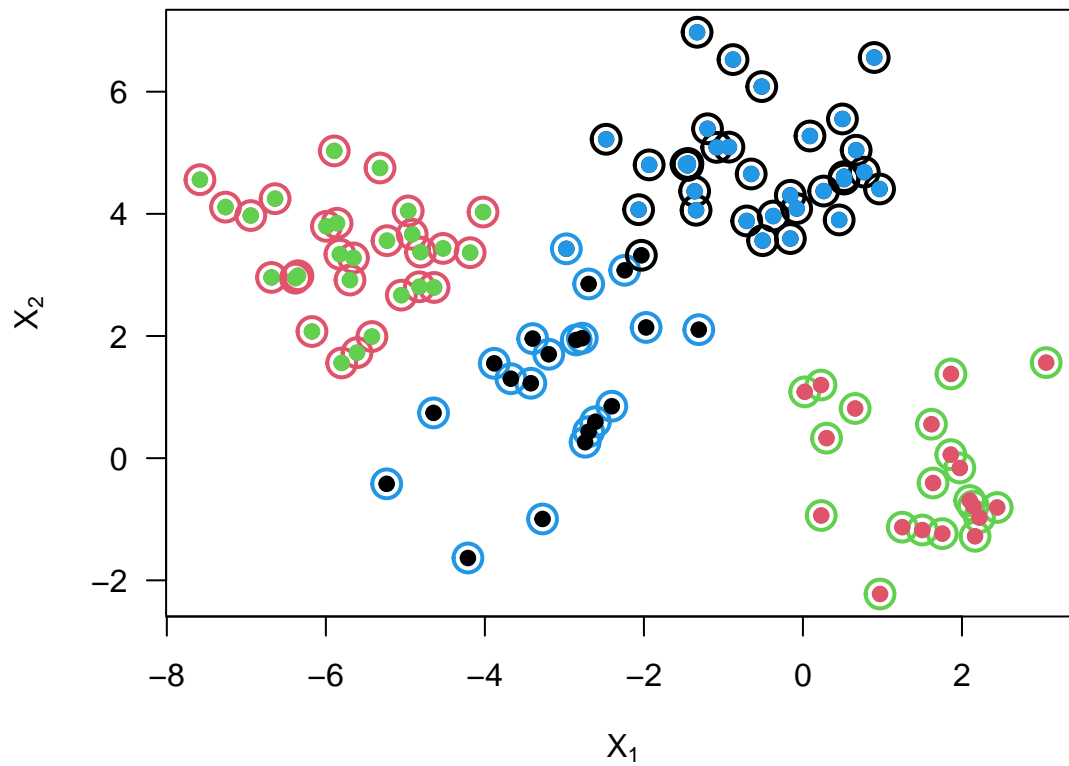


```
km.out <- kmeans(x, 4, nstart = 15)
km.out

## K-means clustering with 4 clusters of sizes 32, 28, 20, 20
##
## Cluster means:
##      [,1]      [,2]
## 1 -0.5787702  4.7639233
## 2 -5.6518323  3.3513316
## 3  1.4989983 -0.2412154
## 4 -3.1104142  1.2535711
##
## Clustering vector:
##  [1] 2 4 1 2 4 1 2 4 1 1 3 1 1 3 4 3 2 3 2 2 2 2 2 3 1 1 4 2 4 1 2 3 2 4 4 3 3
## [38] 4 3 3 2 4 4 2 2 3 2 1 2 4 2 1 1 3 3 4 3 1 1 1 4 2 2 2 4 4 1 1 3 2 2 1 1 3
## [75] 1 3 2 1 1 1 4 1 4 1 2 3 1 2 2 1 1 4 2 4 1 1 3 3 1 1
##
## Within cluster sum of squares by cluster:
## [1] 53.04203 42.40322 34.95921 48.52107
## (between_SS / total_SS =  85.7 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"

```

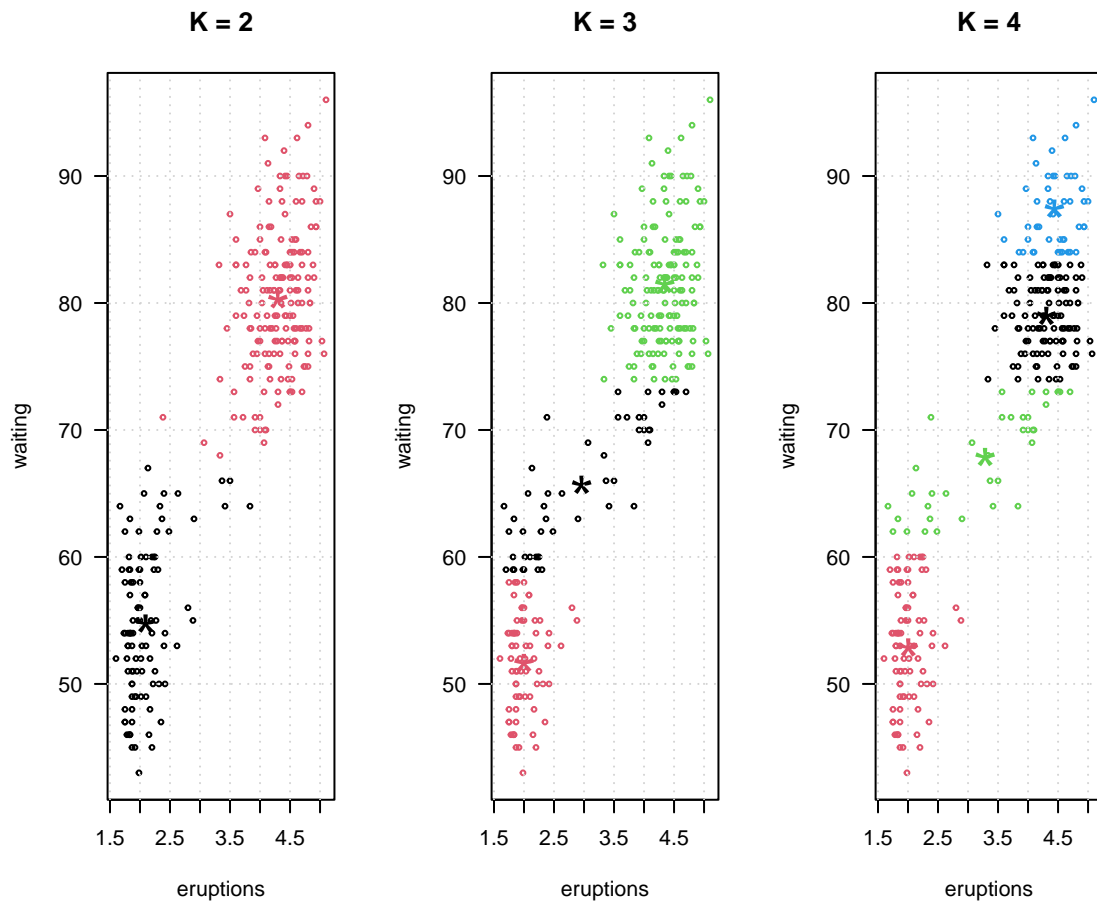
```
plot(x, col = km.out$cluster, cex = 2, pch = 1, lwd = 2,
     xlab = expression(X[1]), ylab = expression(X[2]), las = 1)
points(x, col = which, pch = 19)
points(x, col = c(4, 3, 2, 1)[which], pch = 19)
```



### Geyser Example

```
km3.faithful <- kmeans(faithful, 3)
km2.faithful <- kmeans(faithful, 2)
km4.faithful <- kmeans(faithful, 4)

par(las = 1, mfrow = c(1, 3))
plot(faithful, col = km2.faithful$cluster, cex = 0.5, main = "K = 2")
points(km2.faithful$centers, cex = 3, pch = "*", col = 1:2)
grid()
plot(faithful, col = km3.faithful$cluster, cex = 0.5, main = "K = 3")
points(km3.faithful$centers, cex = 3, pch = "*", col = 1:3)
grid()
plot(faithful, col = km4.faithful$cluster, cex = 0.5, main = "K = 4")
grid()
points(km4.faithful$centers, cex = 3, pch = "*", col = 1:4)
```



## US State Facts and Figures Example

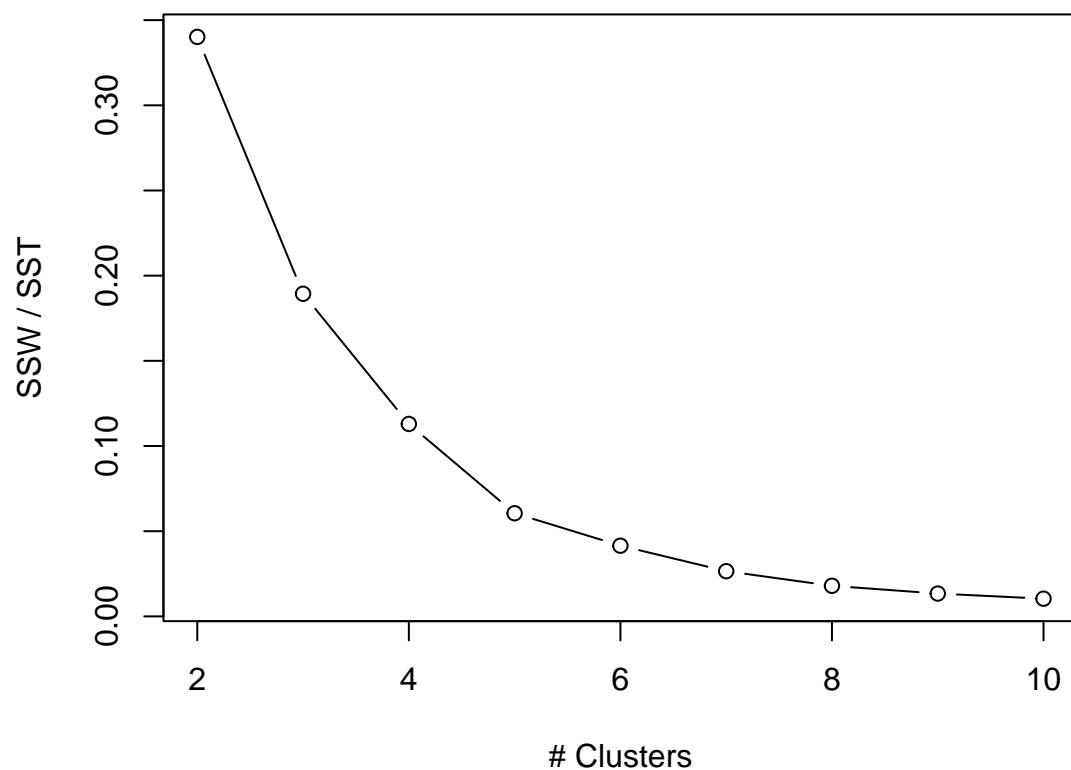
```
# look at states data
vars <- c("Income", "Illiteracy", "Life Exp", "HS Grad")
head(state.x77[, vars])

##           Income Illiteracy Life Exp HS Grad
## Alabama      3624         2.1   69.05   41.3
## Alaska       6315         1.5   69.31   66.7
## Arizona      4530         1.8   70.55   58.1
## Arkansas     3378         1.9   70.66   39.9
## California   5114         1.1   71.71   62.6
## Colorado     4884         0.7   72.06   63.9

# fit k means for k = 2, ..., 10 (raw data)
kmlist <- vector("list", 9)
for(k in 2:10){
  set.seed(1)
  kmlist[[k-1]] <- kmeans(state.x77[, vars], k, nstart = 5000)
}

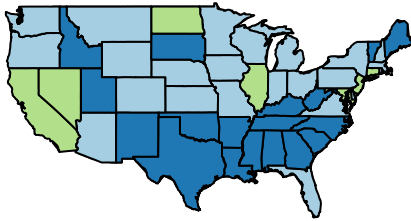
# scree plot (raw data)
tot.withinss <- sapply(kmlist, function(x) x$tot.withinss)
plot(2:10, tot.withinss / kmlist[[1]]$totss, type = "b", xlab = "# Clusters",
     ylab = "SSW / SST", main = "Scree Plot: Raw Data")
```

## Scree Plot: Raw Data

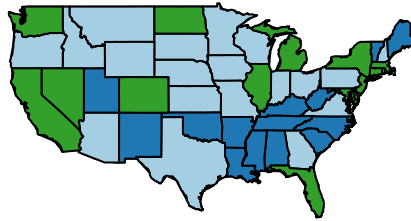


```
# plot results (raw data)
library(maps)
library(RColorBrewer)
par(mfrow = c(2, 2))
for(k in 3:6){
  map(database = "state")
  title(paste0("K=", k, " Clusters: Raw Data"))
  cols <- brewer.pal(k, "Paired")
  for(j in 1:k){
    ix <- names(which(kmlist[[k-1]]$cluster==j))
    if(length(ix) > 1) map(database = "state", regions = ix, col = cols[j],
                          fill = T, add = TRUE)
  }
}
```

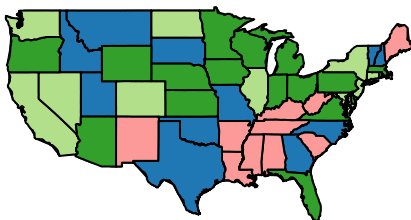
**K=3 Clusters: Raw Data**



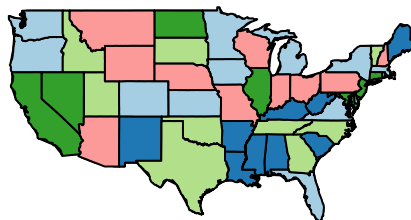
**K=4 Clusters: Raw Data**



**K=5 Clusters: Raw Data**



**K=6 Clusters: Raw Data**



## Hierarchical Clustering

### US State Facts and Figures Example

```
apply(state.x77[, vars], 2, mean)

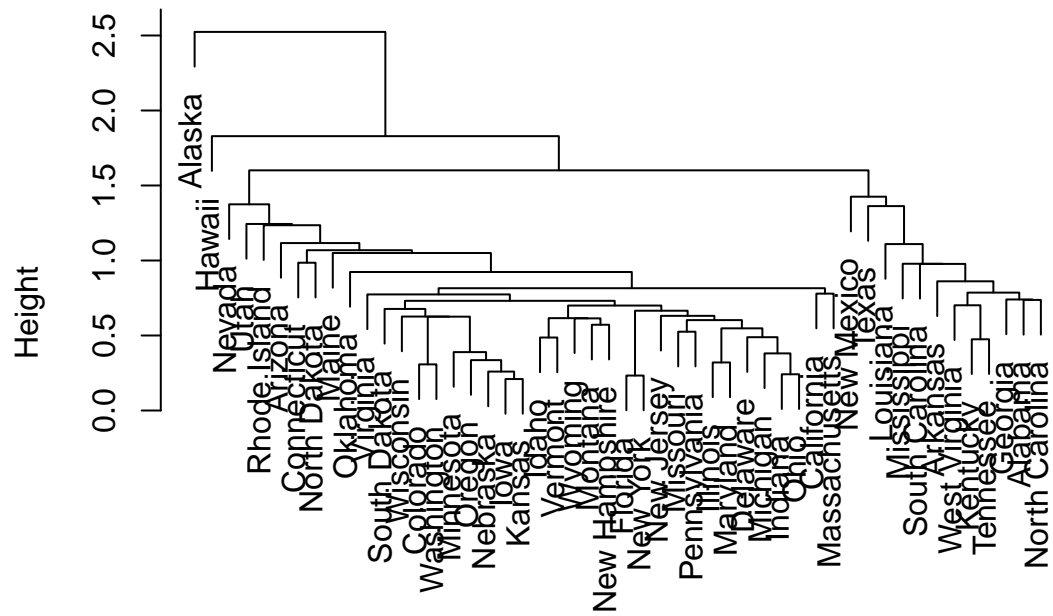
##      Income Illiteracy   Life Exp   HS Grad
## 4435.8000    1.1700    70.8786    53.1080

apply(state.x77[, vars], 2, sd)

##      Income  Illiteracy   Life Exp   HS Grad
## 614.4699392  0.6095331   1.3423936   8.0769978

# create distance (raw and standardized)
distrw <- dist(state.x77[, vars])
diststd <- dist(scale(state.x77[, vars]))
# hierarchical clustering (standardized data)
hcstdSL <- hclust(diststd, method = "single")
hcstdCL <- hclust(diststd, method = "complete")
hcstdAL <- hclust(diststd, method = "average")
# plot results (standardized data)
plot(hcstdSL)
```

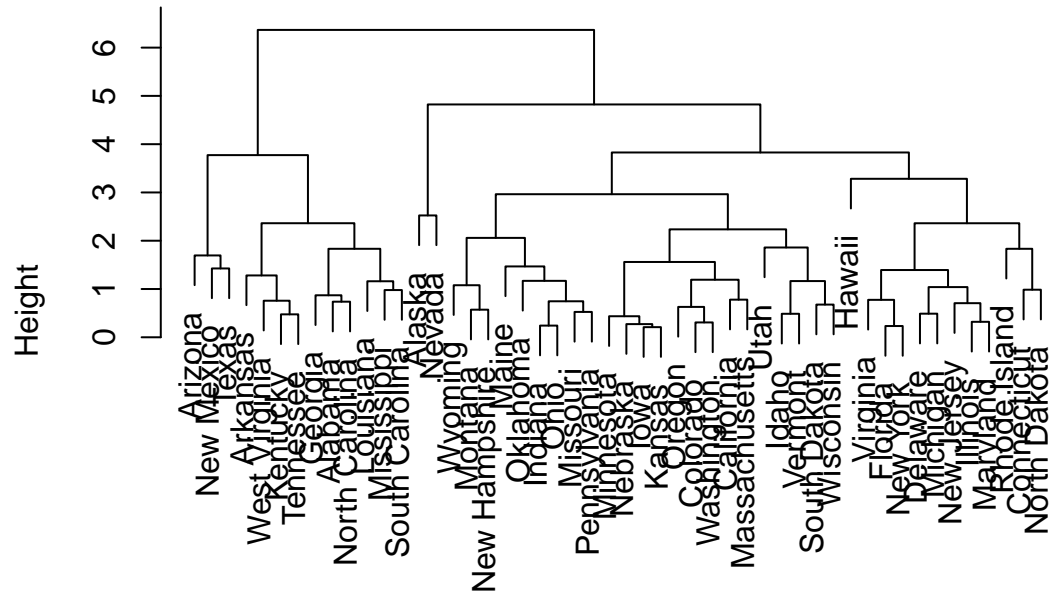
## Cluster Dendrogram



diststd  
hclust (\*, "single")

```
plot(hcstdCL)
```

## Cluster Dendrogram

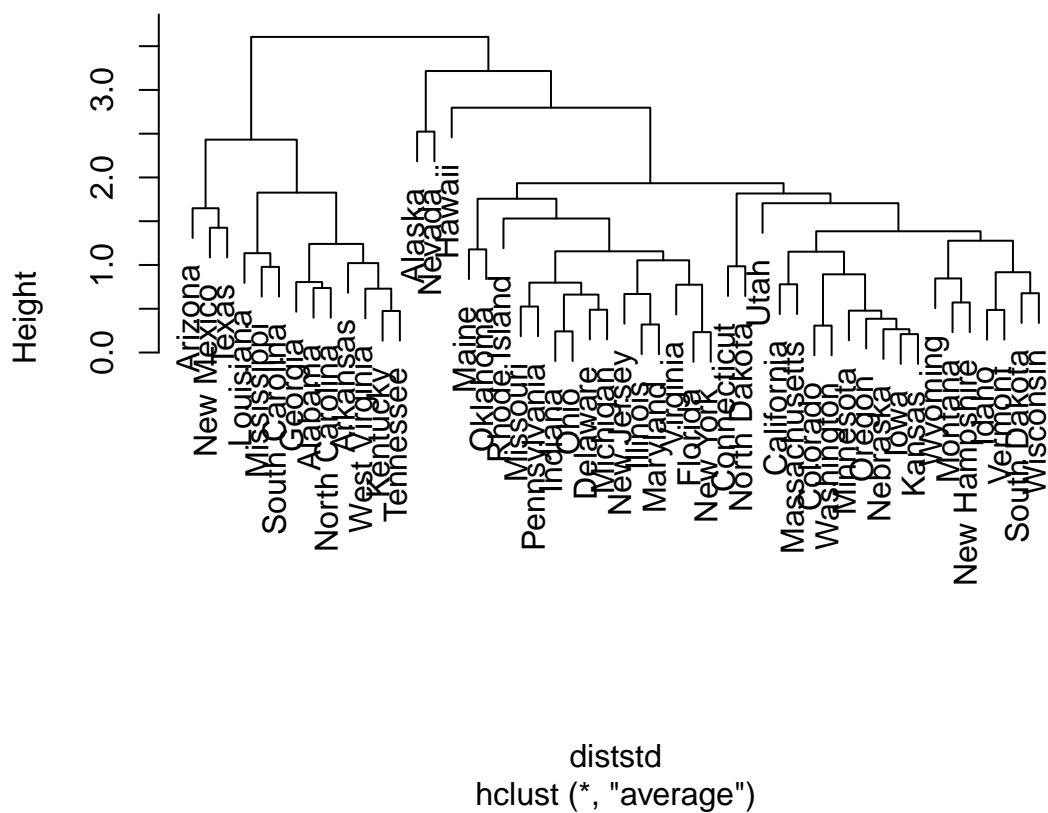


dlistd  
hclust (\*, "complete")

```
plot(hcstdAL)
```



## Cluster Dendrogram

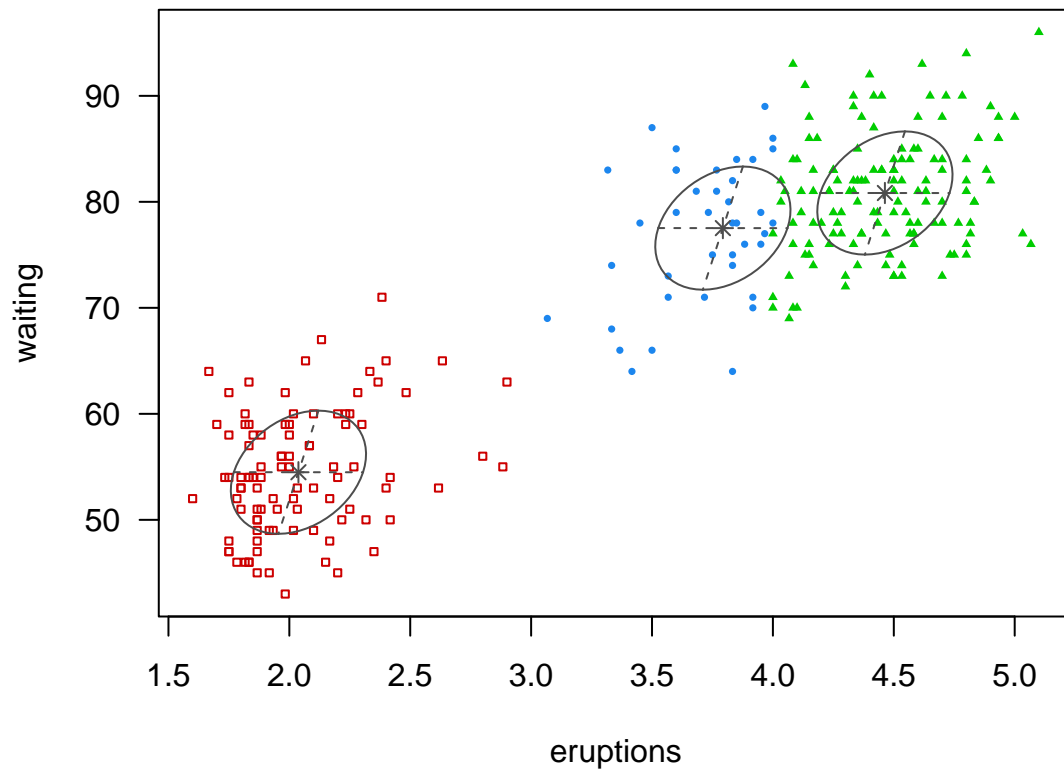


## Model-based

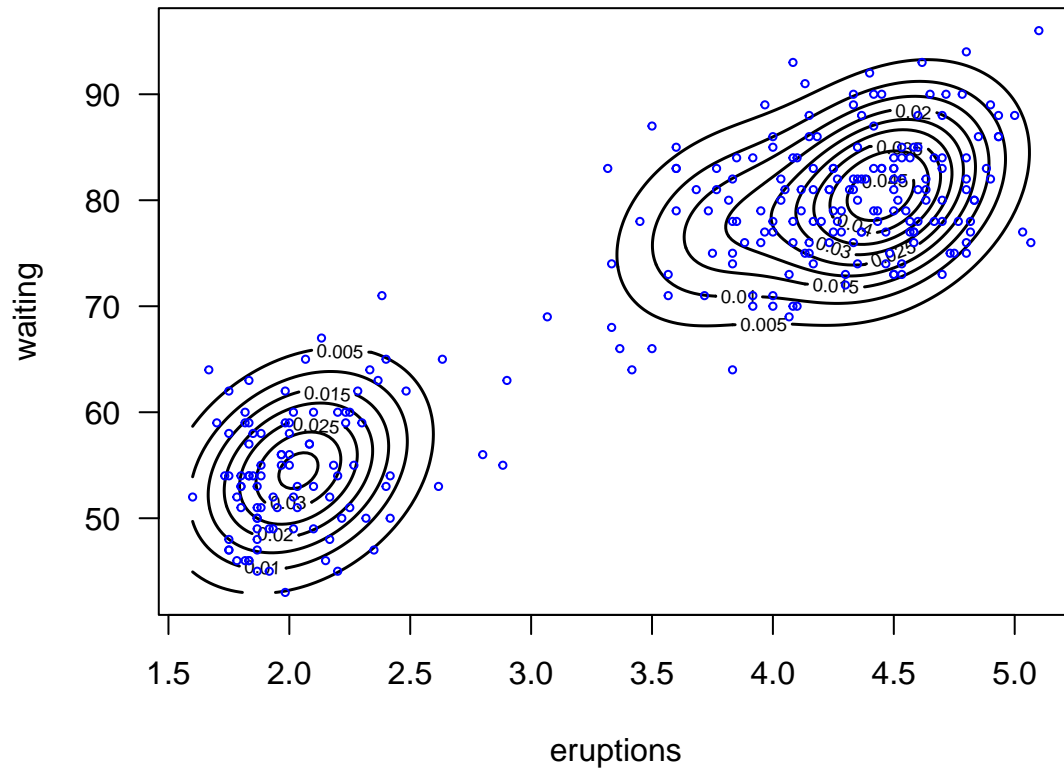
### Geyser Example

```
library(mclust)
BIC <- mclustBIC(faithful)
model1 <- Mclust(faithful, x = BIC)

plot(model1, what = "classification", cex = 0.5, las = 1)
```



```
plot(model1, what = "density", col = "black", lwd = 1.5, las = 1)
points(faithful, col = "blue", cex = 0.5)
```



```
(LRT <- mclustBootstrapLRT(faithful, modelName = "VVV"))
```

```
## -----
## Bootstrap sequential LRT for the number of mixture components
## -----
## Model          = VVV
## Replications = 999
##               LRTS bootstrap p-value
## 1 vs 2      319.065354      0.001
## 2 vs 3       6.130516      0.559
```

### Fisher's Iris Data Example

```
data(iris)
attach(iris)
iris$Species <- factor(iris$Species)
dat <- iris[, 1:4]
BIC <- mclustBIC(dat)
model2 <- Mclust(dat, x = BIC)

par(las = 1)
plot(model2, what = "classification", cex = 0.5, col = c("green", "blue"))
```

