MichaelBasta_Assignment_5

Michael Basta

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```
Cereals <- read.csv("/Users/michaelbasta/Documents/Fundmentals of Machine Learning /Module 8/Cereals.cs
df <- Cereals
row.names(df) <- df$name</pre>
df <- na.omit(df)</pre>
# eliminating all the non
df <- df[,4:16]
df <- scale(df)
head(df)
                              calories
                                          protein
                                                         fat
## 100%_Bran
                            -1.8659155 1.3817478 0.0000000 -0.3910227
## 100%_Natural_Bran
                             ## All-Bran
                            -1.8659155 1.3817478 0.0000000 1.1795987
## All-Bran_with_Extra_Fiber -2.8737823 1.3817478 -0.9932203 -0.2702057
## Apple_Cinnamon_Cheerios
                             0.1498180 -0.4773310 0.9932203 0.2130625
## Apple_Jacks
                             0.1498180 - 0.4773310 - 0.9932203 - 0.4514312
##
                                             carbo
                                                       sugars
## 100%_Bran
                             3.22866747 -2.5001396 -0.2542051 2.5605229
## 100% Natural Bran
                            -0.07249167 -1.7292632 0.2046041
                                                              0.5147738
## All-Bran
                             2.81602258 -1.9862220 -0.4836096 3.1248675
## All-Bran_with_Extra_Fiber 4.87924705 -1.7292632 -1.6306324 3.2659536
## Apple_Cinnamon_Cheerios
                            -0.27881412 -1.0868662 0.6634132 -0.4022862
## Apple_Jacks
                            -0.48513656 -0.9583868 1.5810314 -0.9666308
##
                              vitamins
                                            shelf
                                                      weight
## 100%_Bran
                            -0.1818422 0.9419715 -0.2008324 -2.0856582
## 100%_Natural_Bran
                            -1.3032024 0.9419715 -0.2008324 0.7567534
## All-Bran
                            -0.1818422 0.9419715 -0.2008324 -2.0856582
## All-Bran_with_Extra_Fiber -0.1818422 0.9419715 -0.2008324 -1.3644493
## Apple_Cinnamon_Cheerios
                            -0.1818422 -1.4616799 -0.2008324 -0.3038480
## Apple_Jacks
                            -0.1818422 -0.2598542 -0.2008324 0.7567534
                                rating
```

```
d <- dist(df, method = "euclidean")
# Hierarchical clustering using Complete Linkage</pre>
```

1.8549038

-0.5977113

1.2151965

-0.9165248

-0.6553998

100%_Bran

All-Bran

Apple_Jacks

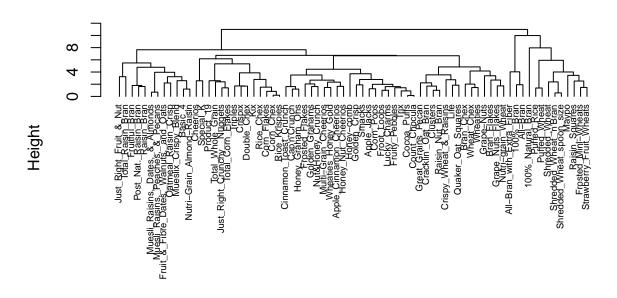
100%_Natural_Bran

Apple_Cinnamon_Cheerios

All-Bran_with_Extra_Fiber 3.6578436

```
hc1 <- hclust(d, method = "complete")

# Plot the obtained dendogram
plot(hc1, cex = 0.6, hang = -1)</pre>
```



d hclust (*, "complete")

```
library(cluster)

# Compute with agnes and with different Linkage methods
hc_single <- agnes(df, method = "single")
hc_complete <- agnes(df, method = "complete")
hc_average <- agnes(df, method = "average")

# Compare Agglomerative coefficients
print(hc_single$ac)

## [1] 0.6067859

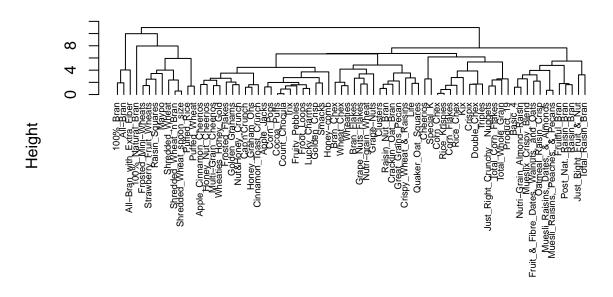
print(hc_complete$ac)

## [1] 0.8353712

print(hc_average$ac)</pre>
```

```
pltree(hc_complete, cex = 0.6, hang = -1, main = "Dedrogram of agnes")
```

Dedrogram of agnes



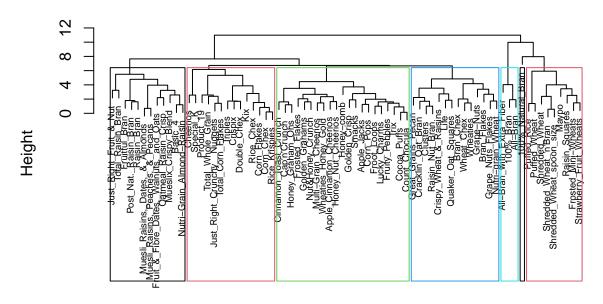
df agnes (*, "complete")

```
paste("Best Linkage method is the complete with the highest cofficient")
```

[1] "Best Linkage method is the complete with the highest cofficient"

```
# Marking clusters to see the best number
hc_complete <- hclust(d, method = "complete")

# plot dendrogram
plot(hc_complete, cex = 0.6)
rect.hclust(hc_complete, k = 7, border = 1:5)</pre>
```



d hclust (*, "complete")

```
paste("The choice for clusters is 7 based on the height chosen")
```

[1] "The choice for clusters is 7 based on the height chosen"

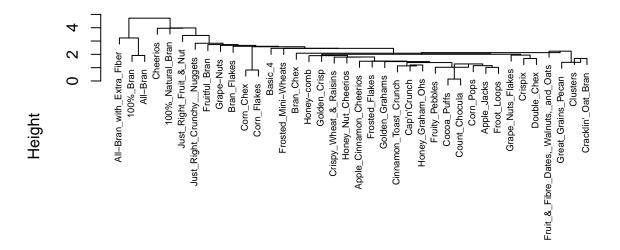
```
# In order to check the stability of the clusters we need to
# partition data and then see how well clusters formed based on part

Index <- 1:38
# Partition A
Train <- df[Index,]

# Partition B
Test <- df[-Index,]

# Applying clustering on partition A
d <- dist(Train, method = "euclidean")
hc_complete <- hclust(d, method = "centroid")

plot(hc_complete, cex = 0.6)</pre>
```



d hclust (*, "centroid")

```
hc_single <- agnes(Train, method = "single")
hc_complete <- agnes(Train, method = "complete")
hc_average <- agnes(Train, method = "average")

# Compare Agglomerative coefficients
print(hc_single$ac)

## [1] 0.5853558

print(hc_complete$ac)

## [1] 0.7904243

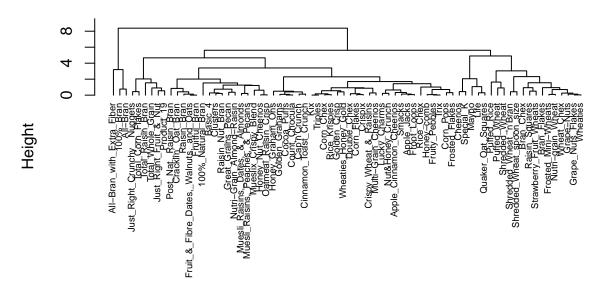
print(hc_average$ac)

## [1] 0.7235901

## From what is seen on the partition dendrogram and
##the agglomerative coefficients we see the structure is less stable

# To be able to cluster "Healthy Cereals" We will
#consider only (Protien, Fiber, Vitamins, rating)
```

```
## Yes data should be normalized As we don't want the
##clustering algorithm to depend to an arbitrary variable unit
df <- Cereals
row.names(df) <- df$name
df <- na.omit(df)</pre>
# Only Selecting the columns contributing to healthy cereals
df \leftarrow df[,c(5,8,12,16)]
df <- scale(df)</pre>
head(df)
##
                                 protein
                                                fiber
                                                        vitamins
                                                                      rating
## 100%_Bran
                               1.3817478 3.22866747 -0.1818422 1.8549038
## 100%_Natural_Bran
                               0.4522084 -0.07249167 -1.3032024 -0.5977113
## All-Bran
                               1.3817478 2.81602258 -0.1818422 1.2151965
## All-Bran_with_Extra_Fiber 1.3817478 4.87924705 -0.1818422 3.6578436
## Apple_Cinnamon_Cheerios
                              -0.4773310 -0.27881412 -0.1818422 -0.9165248
## Apple_Jacks
                              -0.4773310 -0.48513656 -0.1818422 -0.6553998
d <- dist(df, method = "euclidean")</pre>
# Hierarchical clustering using Complete Linkage
hc1 <- hclust(d, method = "complete")</pre>
# Plot the obtained dendogram
plot(hc1, cex = 0.6, hang = -1)
```



d hclust (*, "complete")