Assignment 5.R

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```
# Assignment 5 Fundamentals of Machine Learing
# Data comes from cereals.csv
library(utils)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(class)
library(caret)
## Loading required package: lattice
## Loading required package: ggplot2
library(FNN)
## Attaching package: 'FNN'
## The following objects are masked from 'package:class':
##
       knn, knn.cv
library(e1071)
library(reshape2)
library(cluster)
```

Warning: package 'cluster' was built under R version 4.0.5

library(factoextra) ## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa WD<-setwd("C:/Users/Jason/Documents/MSBA/Fundamentals for Machine Learning/Assignment5") Cereal<-read.csv("Cereals.csv", header = TRUE)</pre> Cereal2 <- Cereal[complete.cases(Cereal),]</pre> row.names(Cereal2) <- Cereal2[,1]</pre> Cereal2 <- Cereal2[,-1:-3]</pre> NC <- sapply(Cereal2, scale)</pre> CD <- dist(NC, method = 'euclidean')</pre> Hsingle <- agnes(CD, method = 'single')</pre> Hcomp <- agnes(CD, method = 'complete')</pre> Havg <- agnes(CD, method = 'average')</pre> Hward <- agnes(CD, method = 'ward')</pre> Hsingle agnes(x = CD, method = "single") ## Agglomerative coefficient: 0.6067859 ## Order of objects: ## [1] 1 3 4 2 5 35 6 14 18 71 41 23 28 17 10 34 12 64 46 74 47 8 72 73 30 ## [26] 24 29 36 7 48 50 26 27 51 56 13 57 19 55 33 40 21 31 49 20 22 70 32 15 60 ## [51] 16 59 9 25 66 58 42 61 62 63 39 45 11 65 43 44 37 67 69 52 38 68 53 54 ## Height (summary): Min. 1st Qu. Median Mean 3rd Qu. Max. ## 0.1431 1.3777 1.7695 1.8668 2.2787 4.0361 ## ## Available components: ## [1] "order" "height" "ac" "merge" "diss" "call" "method" Hcomp ## Call: agnes(x = CD, method = "complete") ## Agglomerative coefficient: 0.8353712 ## Order of objects: ## [1] 1 3 4 2 25 66 58 42 61 62 63 53 54 5 35 46 74 24 30 47 10 34 12 6 17 ## [26] 14 18 71 28 23 41 29 64 36 8 72 73 9 31 49 32 13 57 19 33 21 40 55 11 65 ## Height (summary):

Mean 3rd Qu.

##

##

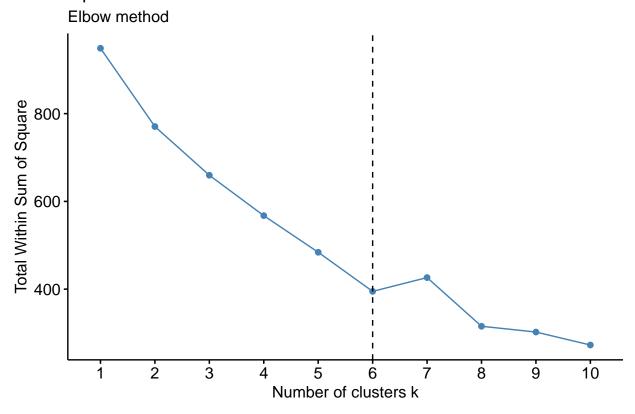
Min. 1st Qu. Median

0.1431 1.6076 2.3389 2.9321 3.7169 10.9839

```
## Available components:
## [1] "order" "height" "ac" "merge" "diss" "call"
                                                          "method"
Havg
## Call:
            agnes(x = CD, method = "average")
## Agglomerative coefficient: 0.7766075
## Order of objects:
## [1] 1 3 4 2 5 35 46 74 24 30 47 6 17 14 18 71 23 41 28 29 64 10 34 12 36
## [26] 8 72 73 9 32 20 22 70 31 49 13 57 19 33 40 55 21 15 60 16 59 39 25 66 58
## [51] 42 61 62 63 7 48 50 45 26 27 51 56 43 44 37 67 69 52 38 68 11 65 53 54
## Height (summary):
     Min. 1st Qu. Median Mean 3rd Qu.
## 0.1431 1.4633 2.0666 2.4461 2.9445 7.7243
## Available components:
## [1] "order" "height" "ac" "merge" "diss" "call"
                                                          "method"
Hward
           agnes(x = CD, method = "ward")
## Call:
## Agglomerative coefficient: 0.9046042
## Order of objects:
## [1] 1 3 4 2 43 44 13 57 19 33 21 40 55 7 48 45 26 50 27 51 56 38 68 5 35
## [26] 46 74 24 30 47 10 34 12 6 17 29 64 14 18 71 28 23 41 36 8 72 73 9 31 49
## [51] 32 20 22 70 11 65 15 60 16 59 39 37 67 69 52 25 66 58 42 61 62 63 53 54
## Height (summary):
     Min. 1st Qu. Median Mean 3rd Qu.
## 0.1431 1.5858 2.3422 3.6092 4.1559 18.5749
## Available components:
## [1] "order" "height" "ac" "merge" "diss" "call"
                                                          "method"
# WARD clustering method is the best, as it's agglomerative coefficient is
# closest to 1
fviz_nbclust(NC, kmeans, method = "wss") +
 geom vline(xintercept = 6, linetype = 2)+
```

labs(subtitle = "Elbow method")

Optimal number of clusters



```
# the plot elbow is at k=6, therefore 6 clusters is ideal

Partition <- createDataPartition(NC[,1], p = .5, list = FALSE )
A <- NC[Partition,]
B <- NC[-Partition,]

Award <- agnes(A, method = 'ward')
Bward <- agnes(B, method = 'ward')

Aclus <- cutree(Award, k = 6)
Bclus <- cutree(Bward, k = 6)

Aclus</pre>
```

[1] 1 2 2 3 3 2 3 3 3 4 5 3 3 2 2 2 3 6 3 4 5 5 3 3 5 4 2 5 2 4 4 6 6 6 3 3 3

Bclus

```
 \hbox{\tt ##} \quad \hbox{\tt [1]} \ 1 \ 2 \ 1 \ 2 \ 2 \ 3 \ 2 \ 2 \ 4 \ 2 \ 2 \ 5 \ 5 \ 2 \ 6 \ 3 \ 2 \ 2 \ 2 \ 5 \ 2 \ 2 \ 3 \ 3 \ 6 \ 3 \ 5 \ 6 \ 5 \ 6 \ 6 \ 2 \ 4 \ 6 \ 2 \ 2
```

```
# data should not be normalized for evaluating healthy cereal options
# data should be looked at to see which cereal has lowest sugar
# and highest density of vitamins per calorie
```