

Assignment 5

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04-15-2023

```
#Displaying the required libraries
```

```
library(cluster)
```

```
library(caret)
```

```
## Loading required package: ggplot2
```

```
## Loading required package: lattice
```

```
library(dendextend)
```

```
##
```

```
## -----
```

```
## Welcome to dendextend version 1.16.0
```

```
## Type citation('dendextend') for how to cite the package.
```

```
##
```

```
## Type browseVignettes(package = 'dendextend') for the package vignette.
```

```
## The github page is: https://github.com/talgalili/dendextend/
```

```
##
```

```
## Suggestions and bug-reports can be submitted at: https://github.com/talgalili/dendextend/issues
```

```
## You may ask questions at stackoverflow, use the r and dendextend tags:
```

```
## https://stackoverflow.com/questions/tagged/dendextend
```

```
##
```

```
## To suppress this message use: suppressPackageStartupMessages(library(dendextend))
```

```
## -----
```

```
##
```

```
## Attaching package: 'dendextend'
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
## cutree
```

```
library(knitr)
```

```
library(factoextra)
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(readr)
```

```
#Creating a data set with only numerical data by importing a dataset
```

```
library(readr)
```

```
sl_Cereals <- read.csv("~/Downloads/Cereals.csv")
```

```
Num_data <- data.frame(sl_Cereals[,4:16])
```

```
#Deleting all cereals with missing values
```

```
Num_data <- na.omit(Num_data)
```

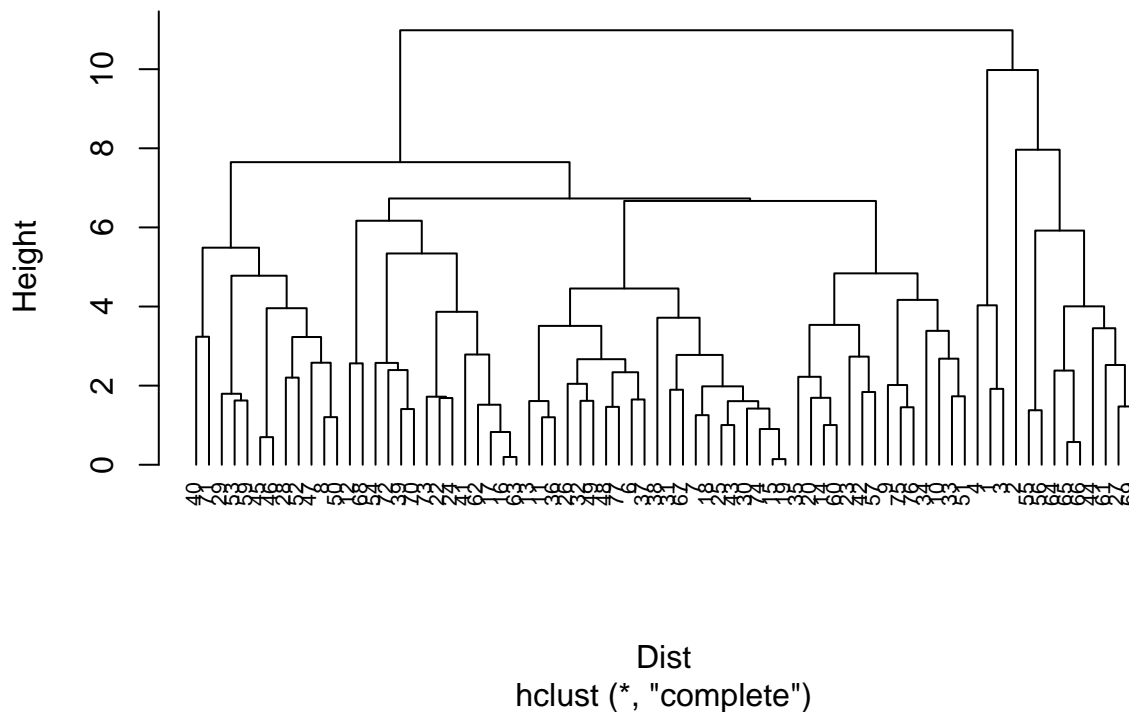
```
#Normalizing data
sl_Cereals_normalise <- scale(Num_data)
```

Apply hierarchical clustering to the adjusted data using the Euclidean distance technique

```
Dist <- dist(sl_Cereals_normalise, method = "euclidean")
H_clust <- hclust(Dist, method = "complete")
```

```
#the dendrogram plotting process.
plot(H_clust, cex = 0.7, hang = -1)
```

Cluster Dendrogram



Comparing the Clustering with single linkage, complete linkage, average linkage and Ward using the Agnes function/

```
single_Hclust <- agnes(sl_Cereals_normalise, method = "single")
complete_Hclust <- agnes(sl_Cereals_normalise, method = "complete")
average_Hclust <- agnes(sl_Cereals_normalise, method = "average")
ward_Hclust <- agnes(sl_Cereals_normalise, method = "ward")
```

Choosing the best method

```
#Choosing the best method
print(single_Hclust$ac)
```

```
## [1] 0.6067859
```

```
print(complete_Hclust$ac)
```

```
## [1] 0.8353712
```

```
print(average_Hclust$ac)
```

```
## [1] 0.7766075
```

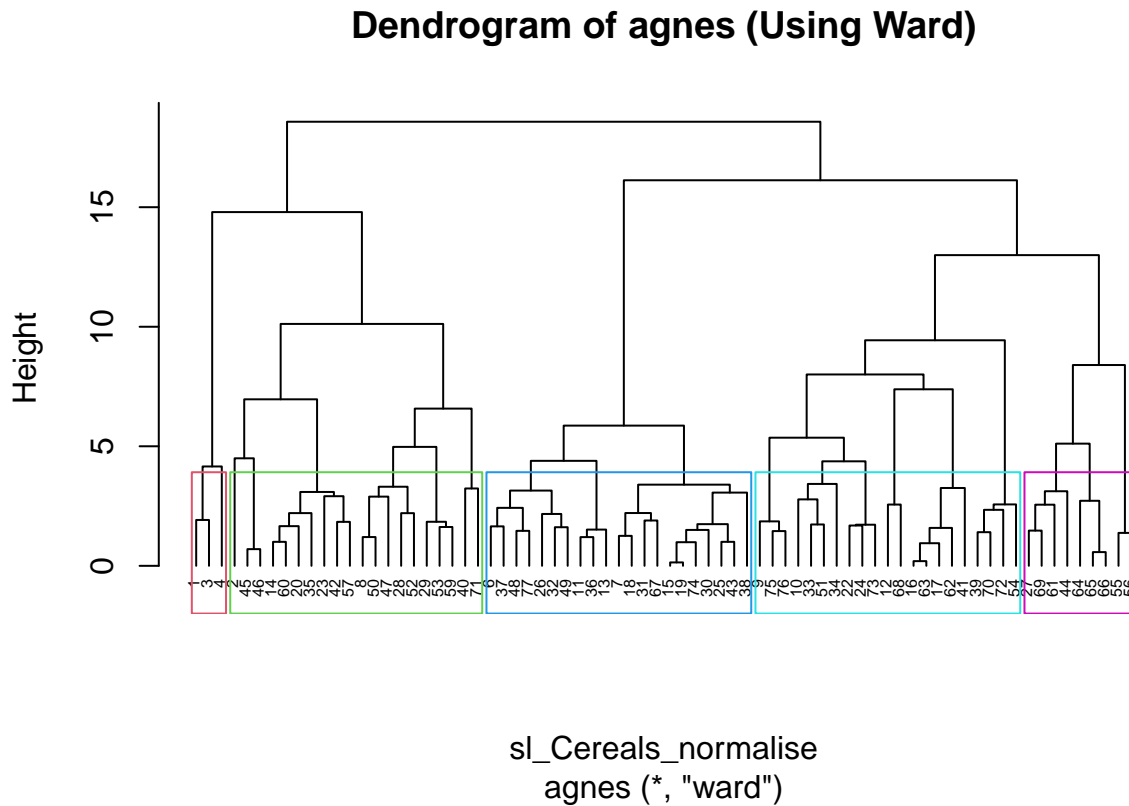
```
print(ward_Hclust$ac)
```

```
## [1] 0.9046042
```

#The ward strategy is the most successful one, as shown by its value of 0.9046042, which is evident given the facts provided.

Choosing the clusters

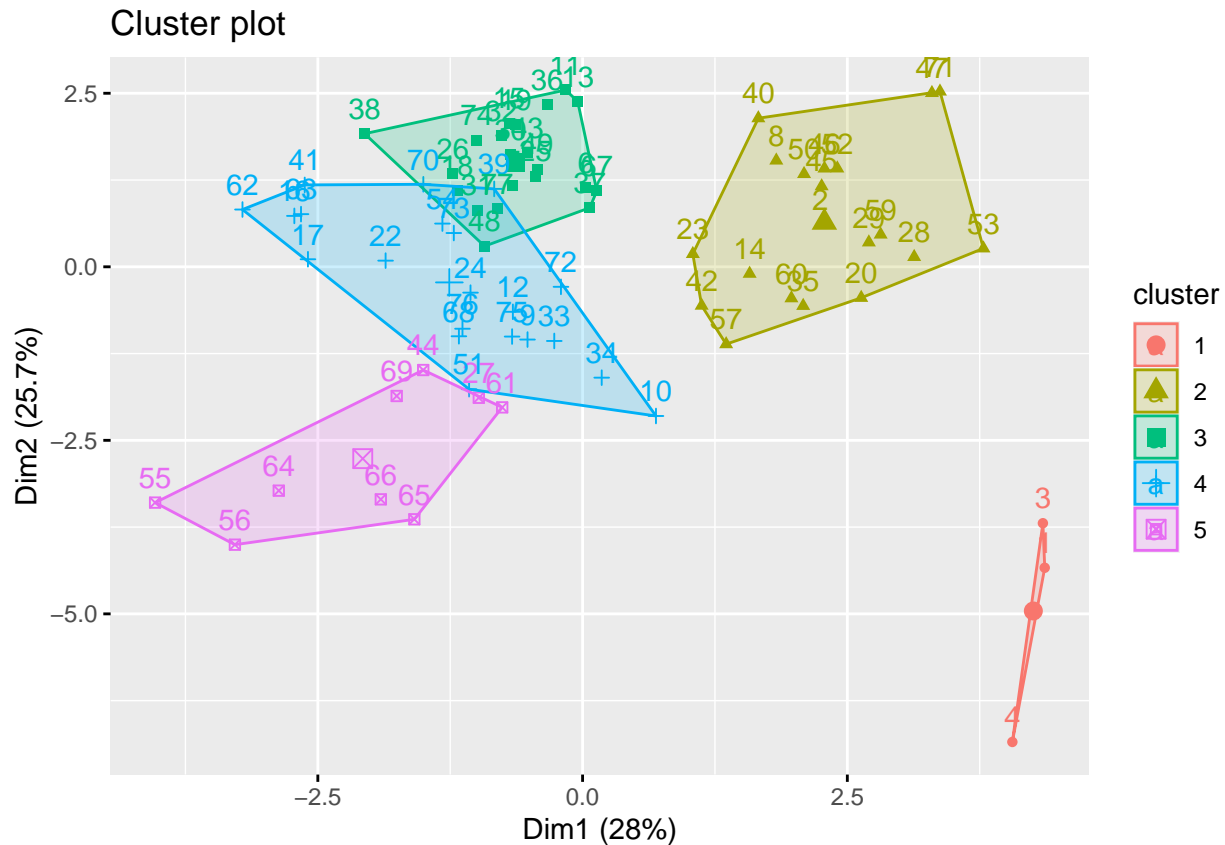
```
pltree(ward_Hclust, cex = 0.5, hang = -1, main = "Dendrogram of agnes (Using Ward)")  
rect.hclust(ward_Hclust, k = 5, border = 2:7)
```



```
S_Group <- cutree(ward_Hclust, k=5)
```

```
D_frame_2 <- as.data.frame(cbind(sl_Cereals_normalise,S_Group))
```

```
fviz_cluster(list(data = D_frame_2, cluster = S_Group))
```



#From the observation mentioned above, clusters can be selected.

Determining the stability and structure of the clusters

#Partitioning the data into A and B

```
set.seed(123)
partition_A <- Num_data[1:55,]
partition_B <- Num_data[56:74,]
```

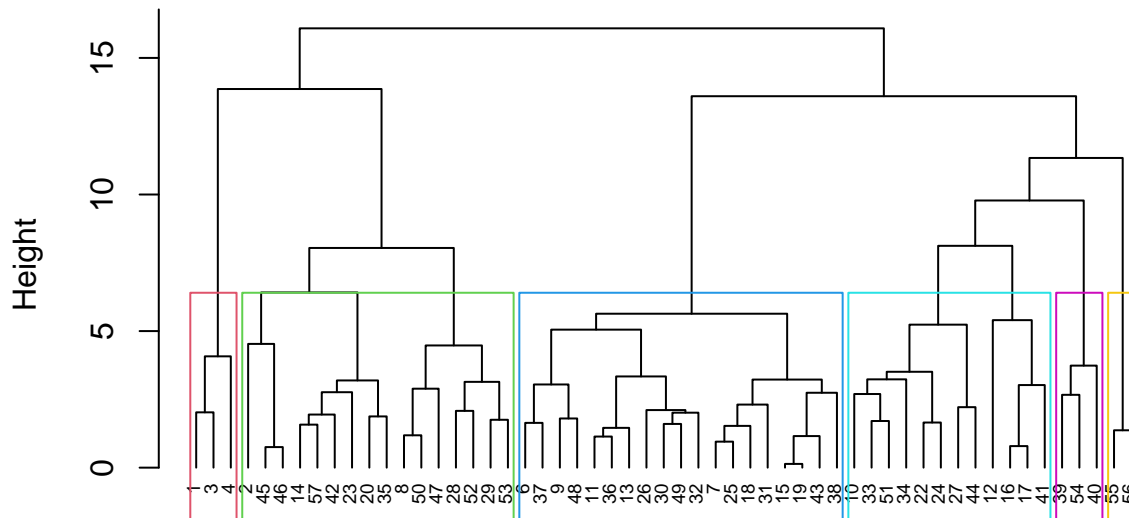
#Performing the Hierarchical Clustering of each partition while considering k = 6.

```
single_sl <- agnes(scale(partition_A), method = "single")
complete_sl <- agnes(scale(partition_A), method = "complete")
average_sl <- agnes(scale(partition_A), method = "average")
ward_sl <- agnes(scale(partition_A), method = "ward")
cbind(single=single_sl$ac , complete=complete_sl$ac , average= average_sl$ac , ward= ward_sl$ac)
```

```
##           single complete average      ward
## [1,] 0.6564842 0.8120228 0.7449303 0.8808195
```

```
pltree(ward_sl, cex = 0.6, hang = -1, main = "Dendrogram of Agnes with Partitioned Data (Using Ward)")
rect.hclust(ward_sl, k = 6, border = 2:7)
```

Dendrogram of Agnes with Partitioned Data (Using Ward)



```
scale(partition_A)
agnes (*, "ward")
```

```
cut_2 <- cutree(ward_sl, k = 6)
```

```
#the centroids are calculated.
```

```
sl_result <- as.data.frame(cbind(partition_A, cut_2))
```

```
sl_result[sl_result$cut_2==1,]
```

```
##   calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
## 1      70      4  1   130    10    5      6    280      25      3      1
## 3      70      4  1   260     9    7      5    320      25      3      1
## 4      50      4  0   140    14    8      0    330      25      3      1
##   cups   rating cut_2
## 1 0.33 68.40297     1
## 3 0.33 59.42551     1
## 4 0.50 93.70491     1
```

```
one_centroid <- colMeans(sl_result[sl_result$cut_2==1,])
```

```
sl_result[sl_result$cut_2==2,]
```

```
##   calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
## 2      120      3  5    15    2.0   8.0      8    135         0      3    1.00
## 8      130      3  2   210    2.0  18.0      8    100        25      3    1.33
## 14     110      3  2   140    2.0  13.0      7    105        25      3    1.00
## 20     110      3  3   140    4.0  10.0      7    160        25      3    1.00
## 23     100      2  1   140    2.0  11.0     10    120        25      3    1.00
## 28     120      3  2   160    5.0  12.0     10    200        25      3    1.25
## 29     120      3  0   240    5.0  14.0     12    190        25      3    1.33
## 35     120      3  3    75    3.0  13.0      4    100        25      3    1.00
## 42     100      4  2   150    2.0  12.0      6     95        25      2    1.00
## 45     150      4  3    95    3.0  16.0     11    170        25      3    1.00
## 46     150      4  3   150    3.0  16.0     11    170        25      3    1.00
```

```
## 47      160      3  2    150   3.0  17.0     13    160     25    3   1.50
## 50      140      3  2    220   3.0  21.0      7    130     25    3   1.33
## 52      130      3  2    170   1.5  13.5     10    120     25    3   1.25
## 53      120      3  1    200   6.0  11.0     14    260     25    3   1.33
## 57      100      4  1    135   2.0  14.0      6    110     25    3   1.00
##      cups   rating cut_2
## 2   1.00 33.98368      2
## 8   0.75 37.03856      2
## 14  0.50 40.40021      2
## 20  0.50 40.44877      2
## 23  0.75 36.17620      2
## 28  0.67 40.91705      2
## 29  0.67 41.01549      2
## 35  0.33 45.81172      2
## 42  0.67 45.32807      2
## 45  1.00 37.13686      2
## 46  1.00 34.13976      2
## 47  0.67 30.31335      2
## 50  0.67 40.69232      2
## 52  0.50 30.45084      2
## 53  0.67 37.84059      2
## 57  0.50 49.51187      2
```

```
two_centroid <- colMeans(sl_result[sl_result$cut_2==2,])
sl_result[sl_result$cut_2==3,]
```

```
##      calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
## 6      110      2  2    180   1.5  10.5     10     70      25     1     1
## 7      110      2  0    125   1.0  11.0     14     30      25     2     1
## 9       90      2  1    200   4.0  15.0      6    125      25     1     1
## 11     120      1  2    220   0.0  12.0     12     35      25     2     1
## 13     120      1  3    210   0.0  13.0      9     45      25     2     1
## 15     110      1  1    180   0.0  12.0     13     55      25     2     1
## 18     110      1  0     90   1.0  13.0     12     20      25     2     1
## 19     110      1  1    180   0.0  12.0     13     65      25     2     1
## 25     110      2  1    125   1.0  11.0     13     30      25     2     1
## 26     110      1  0    200   1.0  14.0     11     25      25     1     1
## 30     110      1  1    135   0.0  13.0     12     25      25     2     1
## 31     100      2  0     45   0.0  11.0     15     40      25     1     1
## 32     110      1  1    280   0.0  15.0      9     45      25     2     1
## 36     120      1  2    220   1.0  12.0     11     45      25     2     1
## 37     110      3  1    250   1.5  11.5     10     90      25     1     1
## 38     110      1  0    180   0.0  14.0     11     35      25     1     1
## 43     110      2  1    180   0.0  12.0     12     55      25     2     1
## 48     100      2  1    220   2.0  15.0      6     90      25     1     1
## 49     120      2  1    190   0.0  15.0      9     40      25     2     1
##      cups   rating cut_2
## 6   0.75 29.50954      3
## 7   1.00 33.17409      3
## 9   0.67 49.12025      3
## 11  0.75 18.04285      3
## 13  0.75 19.82357      3
## 15  1.00 22.73645      3
## 18  1.00 35.78279      3
## 19  1.00 22.39651      3
```

```
## 25 1.00 32.20758      3
## 26 0.75 31.43597      3
## 30 0.75 28.02576      3
## 31 0.88 35.25244      3
## 32 0.75 23.80404      3
## 36 1.00 21.87129      3
## 37 0.75 31.07222      3
## 38 1.33 28.74241      3
## 43 1.00 26.73451      3
## 48 1.00 40.10596      3
## 49 0.67 29.92429      3
```

```
three_centroid <- colMeans(sl_result[sl_result$cut_2==3,])
sl_result[sl_result$cut_2==4,]
```

```
##      calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
## 10          90       3  0   210     5   13      5    190       25      3      1
## 12         110       6  2   290     2   17      1    105       25      1      1
## 16         110       2  0   280     0   22      3     25       25      1      1
## 17         100       2  0   290     1   21      2     35       25      1      1
## 22         110       2  0   220     1   21      3     30       25      3      1
## 24         100       2  0   190     1   18      5     80       25      3      1
## 27         100       3  0     0     3   14      7    100       25      2      1
## 33         100       3  1   140     3   15      5     85       25      3      1
## 34         110       3  0   170     3   17      3     90       25      3      1
## 41         110       2  1   260     0   21      3     40       25      2      1
## 44         100       4  1     0     0   16      3     95       25      2      1
## 51          90       3  0   170     3   18      2     90       25      3      1
```

```
##      cups   rating cut_2
## 10 0.67 53.31381      4
## 12 1.25 50.76500      4
## 16 1.00 41.44502      4
## 17 1.00 45.86332      4
## 22 1.00 46.89564      4
## 24 0.75 44.33086      4
## 27 0.80 58.34514      4
## 33 0.88 52.07690      4
## 34 0.25 53.37101      4
## 41 1.50 39.24111      4
## 44 1.00 54.85092      4
## 51 1.00 59.64284      4
```

```
four_centroid <- colMeans(sl_result[sl_result$cut_2==4,])
centroids <- rbind(one_centroid, two_centroid, three_centroid, four_centroid)
x2 <- as.data.frame(rbind(centroids[,-14], partition_B))
```

#finding the Distance

```
Dist_1 <- get_dist(x2)
Matrix_1 <- as.matrix(Dist_1)
dataframe1 <- data.frame(data=seq(1,nrow(partition_B),1), Clusters =rep(0,nrow(partition_B)))
for(i in 1:nrow(partition_B))
  {dataframe1[i,2] <- which.min(Matrix_1[i+4, 1:4])}
dataframe1
```

```
##      data Clusters
## 1      1          1
```

```
## 2      2      2
## 3      3      2
## 4      4      3
## 5      5      3
## 6      6      2
## 7      7      2
## 8      8      2
## 9      9      3
## 10     10     4
## 11     11     2
## 12     12     3
## 13     13     2
## 14     14     4
## 15     15     4
## 16     16     3
## 17     17     4
## 18     18     4
## 19     19     3
```

```
cbind(D_frame_2$S_Group[56:74], dataframe1$Clusters)
```

```
##      [,1] [,2]
## [1,]    2    1
## [2,]    2    2
## [3,]    5    2
## [4,]    4    3
## [5,]    4    3
## [6,]    5    2
## [7,]    5    2
## [8,]    5    2
## [9,]    3    3
## [10,]   4    4
## [11,]   5    2
## [12,]   4    3
## [13,]   2    2
## [14,]   4    4
## [15,]   4    4
## [16,]   3    3
## [17,]   4    4
## [18,]   4    4
## [19,]   3    3
```

```
table(D_frame_2$S_Group[56:74] == dataframe1$Clusters)
```

```
##
## FALSE  TRUE
##      9    10
```

#Our findings from the observation mentioned above are 9 False and 10 True. As a result, we may say that the model is just partially unstable.

#The elementary public schools would like to choose a set of `sl_Cereals` to include in their daily cafeterias. Every day a different cereal is offered, but all `sl_Cereals` should support a healthy diet. For this goal, you are requested to find a cluster of “healthy Cereals”

finding Clusters of Healthy Cereals

#Clustering Healthy sl_Cereals

```
Healthy_sl_Cereals <- sl_Cereals
Healthy_sl_Cereals_RD <- na.omit(Healthy_sl_Cereals)
clust <- cbind(Healthy_sl_Cereals_RD, S_Group)
clust[clust$S_Group==1,]
```

```
##               name mfr type calories protein fat sodium fiber carbo
## 1          100%_Bran   N   C       70        4  1   130    10     5
## 3             All-Bran   K   C       70        4  1   260     9     7
## 4 All-Bran_with_Extra_Fiber K   C       50        4  0   140    14     8
##   sugars potass vitamins shelf weight cups   rating S_Group
## 1      6     280       25    3      1 0.33 68.40297      1
## 3      5     320       25    3      1 0.33 59.42551      1
## 4      0     330       25    3      1 0.50 93.70491      1
```

```
clust[clust$S_Group==2,]
```

```
##               name mfr type calories protein fat sodium
## 2          100%_Natural_Bran   Q   C       120        3  5     15
## 8              Basic_4       G   C       130        3  2    210
## 14             Clusters       G   C       110        3  2    140
## 20      Cracklin'_Oat_Bran   K   C       110        3  3    140
## 23      Crispy_Wheat_&_Raisins G   C       100        2  1    140
## 28 Fruit_&_Fibre_Dates,_Walnuts,_and_Oats P   C       120        3  2    160
## 29             Fruitful_Bran   K   C       120        3  0    240
## 35      Great_Grains_Pecan   P   C       120        3  3     75
## 40      Just_Right_Fruit_&_Nut K   C       140        3  1    170
## 42              Life       Q   C       100        4  2    150
## 45      Muesli_Raisins,_Dates,_&_Almonds R   C       150        4  3     95
## 46      Muesli_Raisins,_Peaches,_&_Pecans R   C       150        4  3    150
## 47      Mueslix_Crispy_Blend   K   C       160        3  2    150
## 50      Nutri-Grain_Almond-Raisin   K   C       140        3  2    220
## 52      Oatmeal_Raisin_Crisp   G   C       130        3  2    170
## 53      Post_Nat._Raisin_Bran   P   C       120        3  1    200
## 57      Quaker_Oat_Squares   Q   C       100        4  1    135
## 59      Raisin_Bran   K   C       120        3  1    210
## 60      Raisin_Nut_Bran   G   C       100        3  2    140
## 71      Total_Raisin_Bran   G   C       140        3  1    190
##   fiber carbo sugars potass vitamins shelf weight cups   rating S_Group
## 2     2.0   8.0      8    135      0      3   1.00 1.00 33.98368      2
## 8     2.0  18.0      8    100     25      3   1.33 0.75 37.03856      2
## 14    2.0  13.0      7    105     25      3   1.00 0.50 40.40021      2
## 20    4.0  10.0      7    160     25      3   1.00 0.50 40.44877      2
## 23    2.0  11.0     10    120     25      3   1.00 0.75 36.17620      2
## 28    5.0  12.0     10    200     25      3   1.25 0.67 40.91705      2
## 29    5.0  14.0     12    190     25      3   1.33 0.67 41.01549      2
## 35    3.0  13.0      4    100     25      3   1.00 0.33 45.81172      2
## 40    2.0  20.0      9     95    100      3   1.30 0.75 36.47151      2
## 42    2.0  12.0      6     95     25      2   1.00 0.67 45.32807      2
## 45    3.0  16.0     11    170     25      3   1.00 1.00 37.13686      2
## 46    3.0  16.0     11    170     25      3   1.00 1.00 34.13976      2
## 47    3.0  17.0     13    160     25      3   1.50 0.67 30.31335      2
## 50    3.0  21.0      7    130     25      3   1.33 0.67 40.69232      2
## 52    1.5  13.5     10    120     25      3   1.25 0.50 30.45084      2
```

```
## 53  6.0 11.0   14   260      25    3   1.33 0.67 37.84059      2
## 57  2.0 14.0    6   110      25    3   1.00 0.50 49.51187      2
## 59  5.0 14.0   12   240      25    2   1.33 0.75 39.25920      2
## 60  2.5 10.5    8   140      25    3   1.00 0.50 39.70340      2
## 71  4.0 15.0   14   230     100    3   1.50 1.00 28.59278      2
```

```
clust[clust$S_Group==3,]
```

```
##              name mfr type calories protein fat sodium fiber carbo
## 6  Apple_Cinnamon_Cheerios G C    110      2  2   180   1.5  10.5
## 7              Apple_Jacks K C    110      2  0   125   1.0  11.0
## 11             Cap'n'Crunch Q C    120      1  2   220   0.0  12.0
## 13  Cinnamon_Toast_Crunch G C    120      1  3   210   0.0  13.0
## 15             Cocoa_Puffs G C    110      1  1   180   0.0  12.0
## 18             Corn_Pops K C    110      1  0    90   1.0  13.0
## 19             Count_Chocula G C    110      1  1   180   0.0  12.0
## 25             Froot_Loops K C    110      2  1   125   1.0  11.0
## 26             Frosted_Flakes K C    110      1  0   200   1.0  14.0
## 30             Fruity_Pebbles P C    110      1  1   135   0.0  13.0
## 31             Golden_Crisp P C    100      2  0    45   0.0  11.0
## 32             Golden_Grahams G C    110      1  1   280   0.0  15.0
## 36             Honey_Graham_Ohs Q C    120      1  2   220   1.0  12.0
## 37             Honey_Nut_Cheerios G C    110      3  1   250   1.5  11.5
## 38             Honey-comb P C    110      1  0   180   0.0  14.0
## 43             Lucky_Charms G C    110      2  1   180   0.0  12.0
## 48  Multi-Grain_Cheerios G C    100      2  1   220   2.0  15.0
## 49             Nut&Honey_Crunch K C    120      2  1   190   0.0  15.0
## 67             Smacks K C    110      2  1    70   1.0   9.0
## 74             Trix G C    110      1  1   140   0.0  13.0
## 77             Wheaties_Honey_Gold G C    110      2  1   200   1.0  16.0
##      sugars potass vitamins shelf weight cups   rating S_Group
## 6         10      70      25     1      1 0.75 29.50954      3
## 7         14      30      25     2      1 1.00 33.17409      3
## 11        12      35      25     2      1 0.75 18.04285      3
## 13         9      45      25     2      1 0.75 19.82357      3
## 15        13      55      25     2      1 1.00 22.73645      3
## 18        12      20      25     2      1 1.00 35.78279      3
## 19        13      65      25     2      1 1.00 22.39651      3
## 25        13      30      25     2      1 1.00 32.20758      3
## 26        11      25      25     1      1 0.75 31.43597      3
## 30        12      25      25     2      1 0.75 28.02576      3
## 31        15      40      25     1      1 0.88 35.25244      3
## 32         9      45      25     2      1 0.75 23.80404      3
## 36        11      45      25     2      1 1.00 21.87129      3
## 37        10      90      25     1      1 0.75 31.07222      3
## 38        11      35      25     1      1 1.33 28.74241      3
## 43        12      55      25     2      1 1.00 26.73451      3
## 48         6      90      25     1      1 1.00 40.10596      3
## 49         9      40      25     2      1 0.67 29.92429      3
## 67        15      40      25     2      1 0.75 31.23005      3
## 74        12      25      25     2      1 1.00 27.75330      3
## 77         8      60      25     1      1 0.75 36.18756      3
```

```
clust[clust$S_Group==4,]
```

```
##               name mfr type calories protein fat sodium fiber carbo
## 9           Bran_Chex R   C      90         2  1   200     4   15
## 10          Bran_Flakes P   C      90         3  0   210     5   13
## 12           Cheerios G   C     110         6  2   290     2   17
## 16           Corn_Chex R   C     110         2  0   280     0   22
## 17          Corn_Flakes K   C     100         2  0   290     1   21
## 22           Crispix K   C     110         2  0   220     1   21
## 24          Double_Chex R   C     100         2  0   190     1   18
## 33          Grape_Nuts P   C     100         3  1   140     3   15
## 34          Grape-Nuts P   C     110         3  0   170     3   17
## 39 Just_Right_Crunchy__Nuggets K   C     110         2  1   170     1   17
## 41              Kix G   C     110         2  1   260     0   21
## 51          Nutri-grain_Wheat K   C      90         3  0   170     3   18
## 54          Product_19 K   C     100         3  0   320     1   20
## 62           Rice_Chex R   C     110         1  0   240     0   23
## 63          Rice_Krispies K   C     110         2  0   290     0   22
## 68           Special_K K   C     110         6  0   230     1   16
## 70          Total_Corn_Flakes G   C     110         2  1   200     0   21
## 72          Total_Whole_Grain G   C     100         3  1   200     3   16
## 73              Triples G   C     110         2  1   250     0   21
## 75           Wheat_Chex R   C     100         3  1   230     3   17
## 76           Wheaties G   C     100         3  1   200     3   17
```

```
##      sugars potass vitamins shelf weight cups   rating S_Group
## 9         6    125      25     1      1 0.67 49.12025      4
## 10        5    190      25     3      1 0.67 53.31381      4
## 12        1    105      25     1      1 1.25 50.76500      4
## 16        3     25      25     1      1 1.00 41.44502      4
## 17        2     35      25     1      1 1.00 45.86332      4
## 22        3     30      25     3      1 1.00 46.89564      4
## 24        5     80      25     3      1 0.75 44.33086      4
## 33        5     85      25     3      1 0.88 52.07690      4
## 34        3     90      25     3      1 0.25 53.37101      4
## 39        6     60     100     3      1 1.00 36.52368      4
## 41        3     40      25     2      1 1.50 39.24111      4
## 51        2     90      25     3      1 1.00 59.64284      4
## 54        3     45     100     3      1 1.00 41.50354      4
## 62        2     30      25     1      1 1.13 41.99893      4
## 63        3     35      25     1      1 1.00 40.56016      4
## 68        3     55      25     1      1 1.00 53.13132      4
## 70        3     35     100     3      1 1.00 38.83975      4
## 72        3    110     100     3      1 1.00 46.65884      4
## 73        3     60      25     3      1 0.75 39.10617      4
## 75        3    115      25     1      1 0.67 49.78744      4
## 76        3    110      25     1      1 1.00 51.59219      4
```

```
#Mean ratings are used to select the best cluster.
```

```
mean(clust[clust$S_Group==1,"rating"])
```

```
## [1] 73.84446
```

```
mean(clust[clust$S_Group==2,"rating"])
```

```
## [1] 38.26161
```

```
mean(clust[clust$S_Group==3,"rating"])
```

```
## [1] 28.84825
```

```
mean(clust[clust$S_Group==4,"rating"])
```

```
## [1] 46.46513
```

#Cluster 1 may be chosen based on the data mentioned above because it is the highest. #Therefore, Group 1 may be considered of as the cluster for a healthy diet.