

# Assignment 5

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```
#Importing the necessary packages and 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)
```

```
#Importing a dataset and producing a data set that only contains numerical information
```

```
library(readr)
```

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

```
View(Cereals)
```

```
Numerical_data <- data.frame(Cereals[,4:16])
```

```
#Removing missing values from the data
```

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

```
#Data normalization
```

```
Data_Cereals_normalise <- scale(Numerical_data)
```

```
#Using the Euclidean distance algorithm on the normalized data, apply hierarchical clustering to the da
```

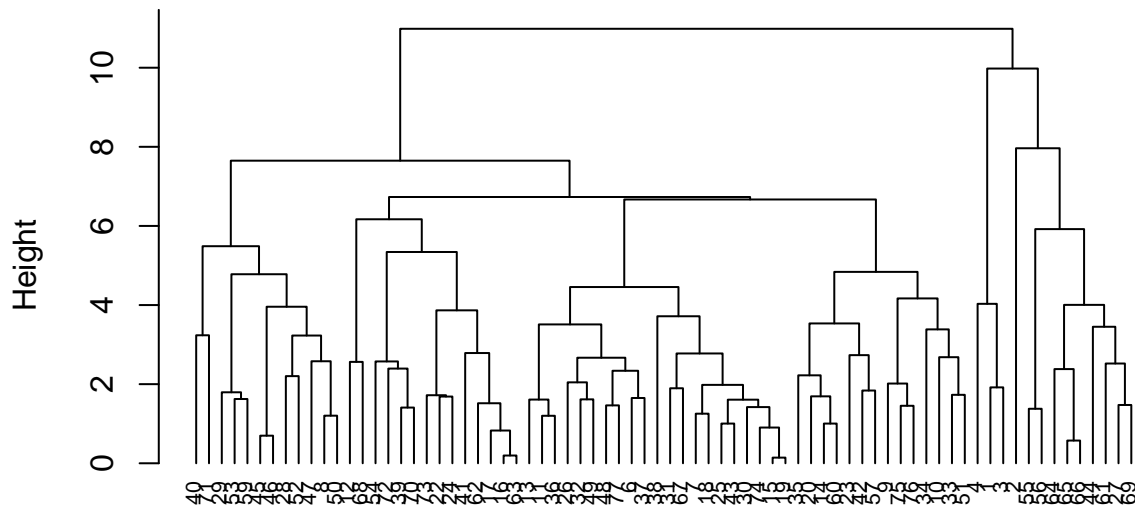
```
Distance <- dist(Data_Cereals_normalise, method = "euclidean")
```

```
Hierarchical_Clustering <- hclust(Distance, method = "complete")
```

```
#Creating a dendrogram plot.
```

```
plot(Hierarchical_Clustering, cex = 0.7, hang = -1)
```

## Cluster Dendrogram



Distance  
hclust (\*, "complete")

```
#Using Agnes function to perform clustering with single linkage, complete linkage  
#, average linkage and Ward.
```

```
HC_single <- agnes(Data_Cereals_normalise, method = "single")
```

```
HC_complete <- agnes(Data_Cereals_normalise, method = "complete")
```

```
HC_average <- agnes(Data_Cereals_normalise, method = "average")
```

```
HC_ward <- agnes(Data_Cereals_normalise, method = "ward")
```

```
#Choosing the most effective strategy
```

```
print(HC_single$ac)
```

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

```
print(HC_complete$ac)
```

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

```
print(HC_average$ac)
```

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

```
print(HC_ward$ac)
```

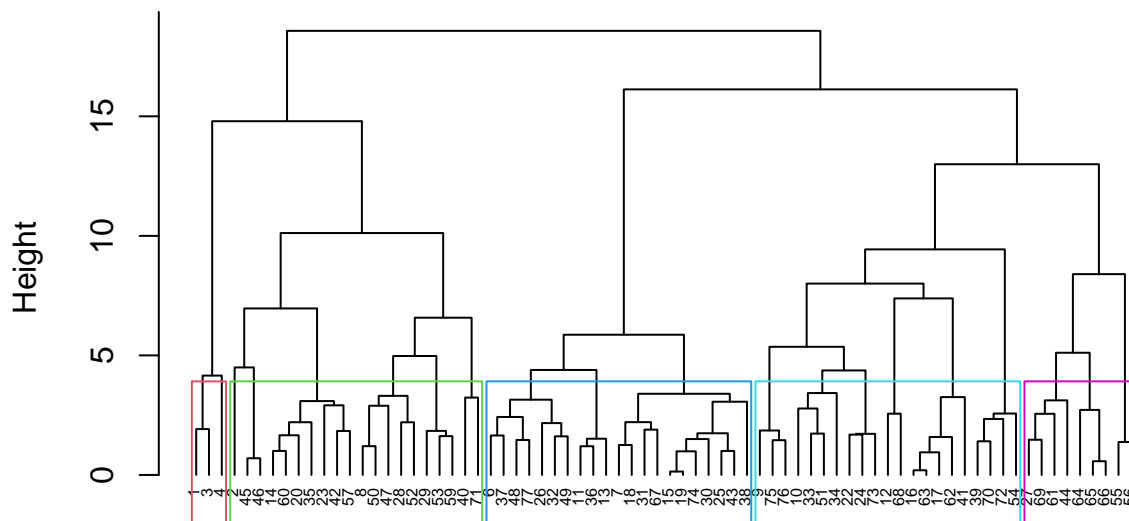
```
## [1] 0.9046042
```

#Given the information above, it is clear that the ward approach is the most effective because of its value of 0.9046042.

*#Task 2- Choosing the clusters:*

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

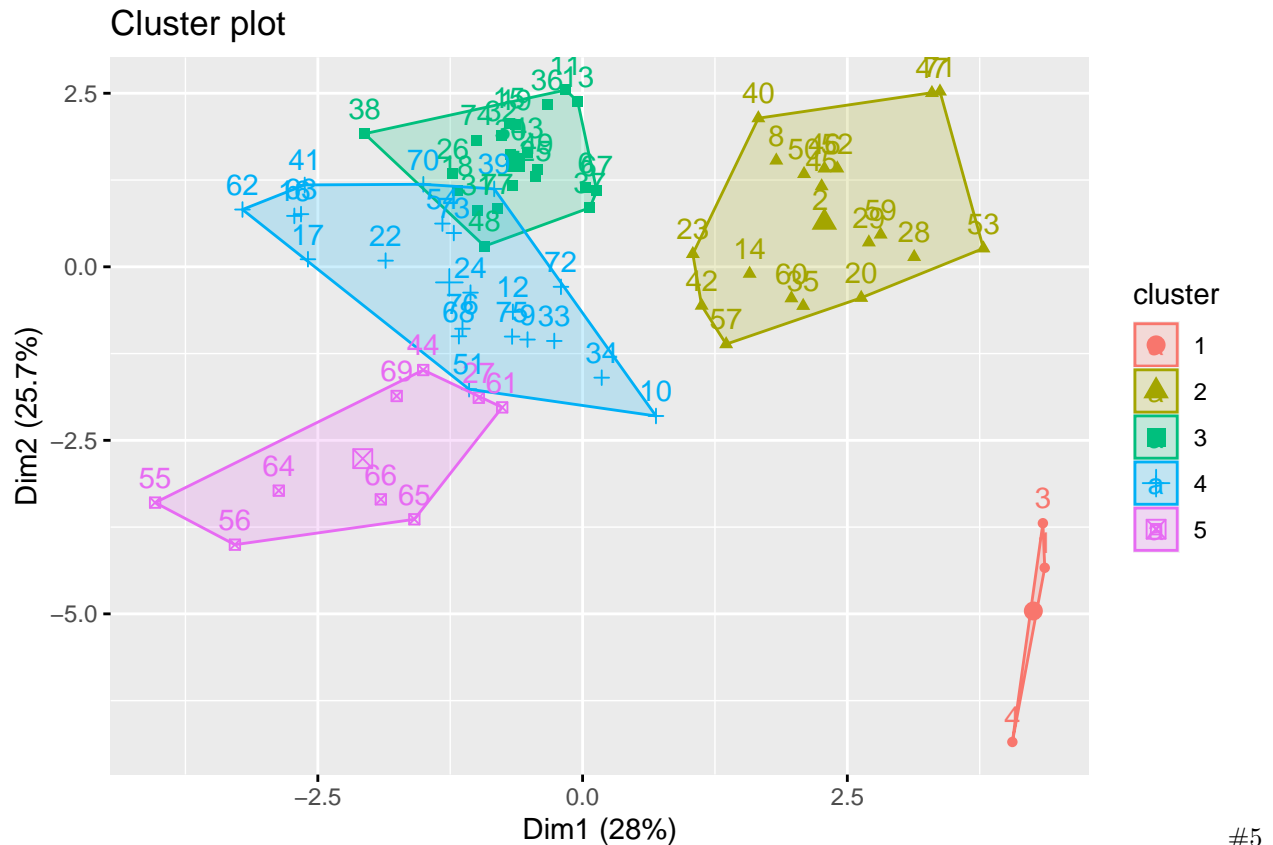
### Dendrogram of agnes (Using Ward)



Data\_Cereals\_normalise  
agnes (\*, "ward")

```
Group <- cutree(HC_ward, k=5)  
dataframe2 <- as.data.frame(cbind(Data_Cereals_normalise, Group))
```

```
fviz_cluster(list(data = dataframe2, cluster = Group))
```



clusters can be chosen from the observation above.. #figuring out the clusters stability and structure.

*#Constructing Partitions*

```
set.seed(123)
Partition_1 <- Numerical_data[1:50,]
Partition_2 <- Numerical_data[51:74,]
```

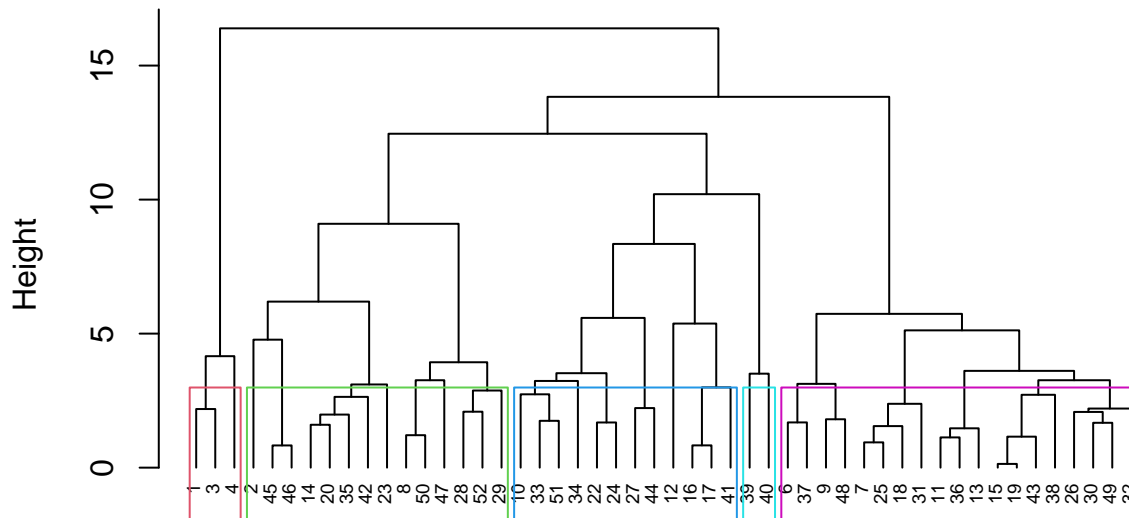
*#Performing Hierarchical Clustering, taking into account k = 5.*

```
RD_single <- agnes(scale(Partition_1), method = "single")
RD_complete <- agnes(scale(Partition_1), method = "complete")
RD_average <- agnes(scale(Partition_1), method = "average")
RD_ward <- agnes(scale(Partition_1), method = "ward")
cbind(single=RD_single$ac , complete=RD_complete$ac , average= RD_average$ac , ward= RD_ward$ac)
```

```
##          single complete average      ward
## [1,] 0.6393338 0.8138238 0.7408904 0.8764323
```

```
pltree(RD_ward, cex = 0.6, hang = -1, main = "Dendrogram of Agnes with Partitioned Data (Using Ward)")
rect.hclust(RD_ward, k = 5, border = 2:7)
```

## Dendrogram of Agnes with Partitioned Data (Using Ward)



```
scale(Partition_1)
agnes (*, "ward")
```

```
cut_2 <- cutree(RD_ward, k = 5)
```

```
#Calculating the centroids.
```

```
result <- as.data.frame(cbind(Partition_1, cut_2))
```

```
result[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
```

```
centroid_1 <- colMeans(result[result$cut_2==1,])
```

```
result[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
##      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
```

```
centroid_2 <- colMeans(result[result$cut_2==2,])
result[result$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
```

```
centroid_3 <- colMeans(result[result$cut_2==3,])
result[result$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
```

```
centroid_4 <- colMeans(result[result$cut_2==4,])
centroids <- rbind(centroid_1, centroid_2, centroid_3, centroid_4)
x2 <- as.data.frame(rbind(centroids[, -14], Partition_2))
```

*#Calculating the Distance.*

```
Distance_1 <- get_dist(x2)
Matrix_1 <- as.matrix(Distance_1)
dataframe1 <- data.frame(data=seq(1,nrow(Partition_2),1), Clusters = rep(0,nrow(Partition_2)))
for(i in 1:nrow(Partition_2))
  {dataframe1[i,2] <- which.min(Matrix_1[i+4, 1:4])}
dataframe1
```

```
##      data Clusters
## 1       1         1
## 2       2         4
## 3       3         3
## 4       4         2
## 5       5         2
```

```
## 6      6      1
## 7      7      2
## 8      8      2
## 9      9      3
## 10    10     3
## 11    11     2
## 12    12     2
## 13    13     2
## 14    14     3
## 15    15     4
## 16    16     2
## 17    17     3
## 18    18     2
## 19    19     4
## 20    20     4
## 21    21     3
## 22    22     4
## 23    23     4
## 24    24     3
```

```
cbind(dataframe2$Group[51:74], dataframe1$Clusters)
```

```
##      [,1] [,2]
## [1,]    2    1
## [2,]    4    4
## [3,]    5    3
## [4,]    5    2
## [5,]    2    2
## [6,]    2    1
## [7,]    2    2
## [8,]    5    2
## [9,]    4    3
## [10,]   4    3
## [11,]   5    2
## [12,]   5    2
## [13,]   5    2
## [14,]   3    3
## [15,]   4    4
## [16,]   5    2
## [17,]   4    3
## [18,]   2    2
## [19,]   4    4
## [20,]   4    4
## [21,]   3    3
## [22,]   4    4
## [23,]   4    4
## [24,]   3    3
```

```
table(dataframe2$Group[51:74] == dataframe1$Clusters)
```

```
##
## FALSE  TRUE
##     12    12
```

#Our results from the observation above are 12 False and 12 True. As a result, we may say that the model is just partly unstable.



*#TASK 3- The elementary public schools would like to choose a set of cereals to include in their daily cafeterias. Every day a different cereal is offered, but all cereals should support a healthy diet. For this goal, you are requested to find a cluster of “healthy cereals.”*

*#Clustering Healthy Cereals.*

```
Healthy_Cereals <- Cereals
Healthy_Cereals_RD <- na.omit(Healthy_Cereals)
clust <- cbind(Healthy_Cereals_RD, Group)
clust[clust$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 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$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 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$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 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$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 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
```

```
#The best cluster is chosen using mean ratings.
```

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

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

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

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

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

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

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

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

#According to the results above, cluster 1 may be selected as it is the highest. #Hence, Group 1 can be considered as the healthy diet cluster.