

# Assignment 5

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```
#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 collection that solely includes numbers by importing a dataset
```

```
library(readr)
```

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

```
View(RM_Cereals)
```

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

```
#Missing values should be omitted
```

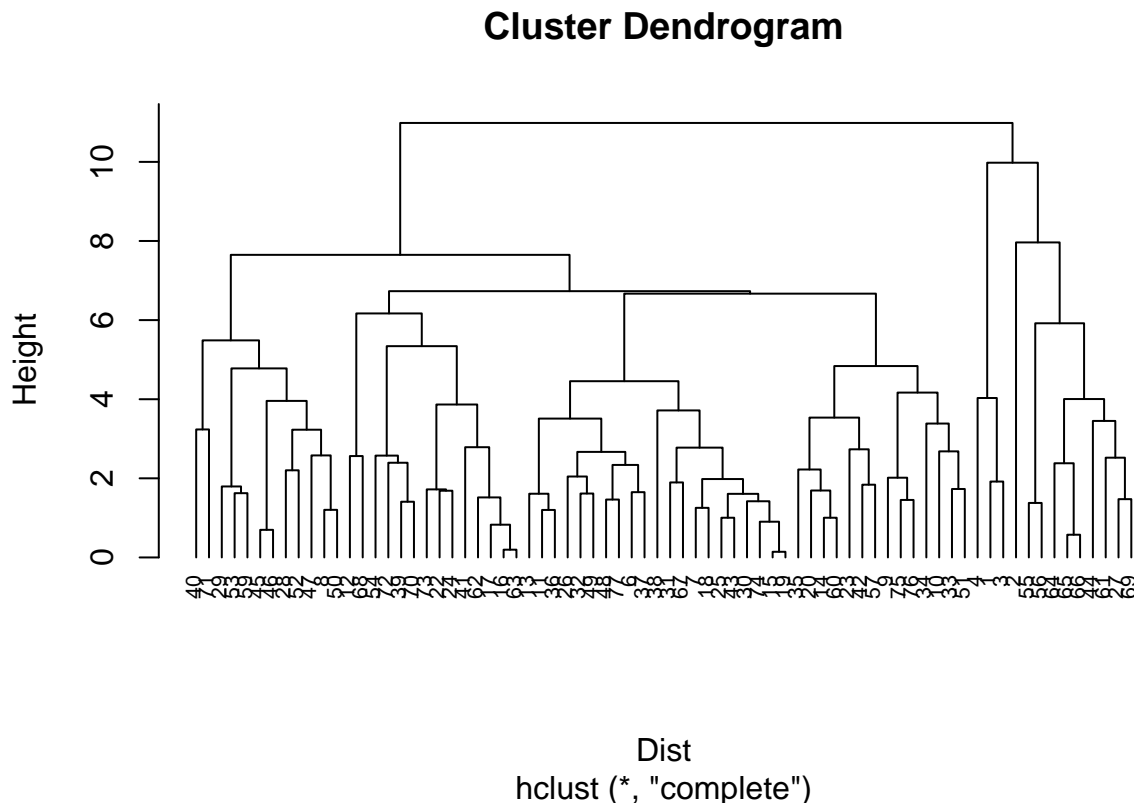
```
Num_data <- na.omit(Num_data)
```

```
#Normalizing data
RM_Cereals_normalize <- scale(Num_data) #Data is normalized using scale function
```

### TASK 1

```
#Use the normalized data to do hierarchical clustering using the Euclidean Distance technique.
Dist <- dist(RM_Cereals_normalize, method = "euclidean")
# Hierarchical clustering using Complete Linkage
H_clust <- hclust(Dist, method = "complete")

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



*#The dendrogram helps us in determining the number of clusters required to classify this dataset.*

```
#Compute with AGNES and with different linkage methods
single_Hclust <- agnes(RM_Cereals_normalize, method = "single")
complete_Hclust <- agnes(RM_Cereals_normalize, method = "complete")
average_Hclust <- agnes(RM_Cereals_normalize, method = "average")
ward_Hclust <- agnes(RM_Cereals_normalize, method = "ward")
```

```
#Choosing the most efficient course of action
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.

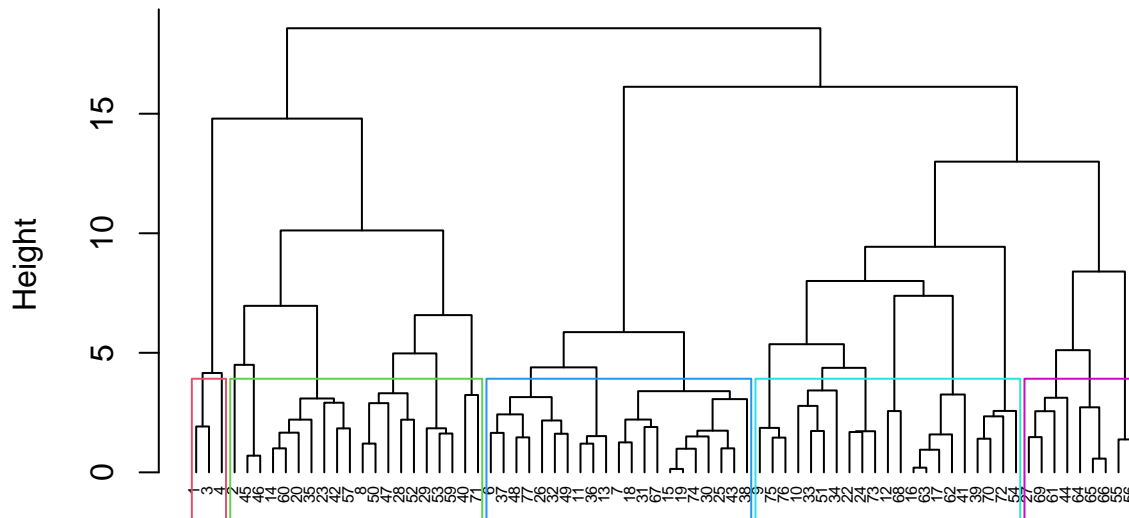
**TASK 2-** How many clusters would you choose?

*#Using the Ward linkage, 5 clusters seem to be enough for grouping the data.*

```
pltree(ward_Hclust, cex = 0.5, hang = -1, main = "Dendrogram of agnes (Using Ward)")
```

```
rect.hclust(ward_Hclust, k = 5, border = 2:7)
```

### Dendrogram of agnes (Using Ward)

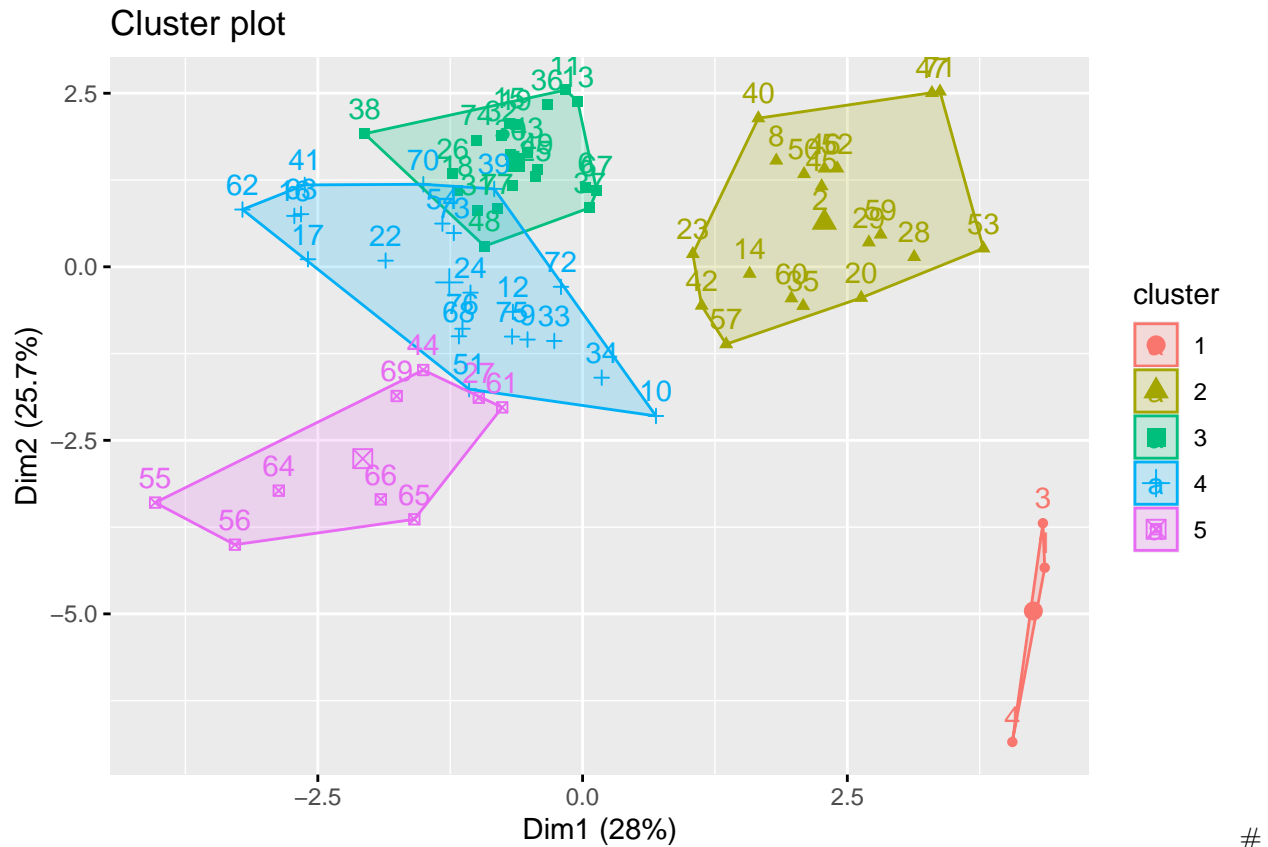


RM\_Cereals\_normalize  
agnes (\*, "ward")

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

```
D_frame_2 <- as.data.frame(cbind(RM_Cereals_normalize,R_Group))
```

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



From the observation mentioned above, 5 clusters can be selected.

**TASK 3** - Determining the stability and structure of the clusters.

*#Building Partitions: partition\_one and partition\_two*

```
set.seed(123)
```

```
partition_one <- Num_data[1:55,]
```

```
partition_two <- Num_data[56:74,]
```

*#Performing Hierarchical Clustering while considering k = 5. Compute with AGNES and with different link*

```
single_rm <- agnes(scale(partition_one), method = "single")
```

```
complete_rm <- agnes(scale(partition_one), method = "complete")
```

```
average_rm <- agnes(scale(partition_one), method = "average")
```

```
ward_rm <- agnes(scale(partition_one), method = "ward")
```

```
cbind(single=single_rm$ac , complete=complete_rm$ac , average= average_rm $ac , ward= ward_rm$ac)
```

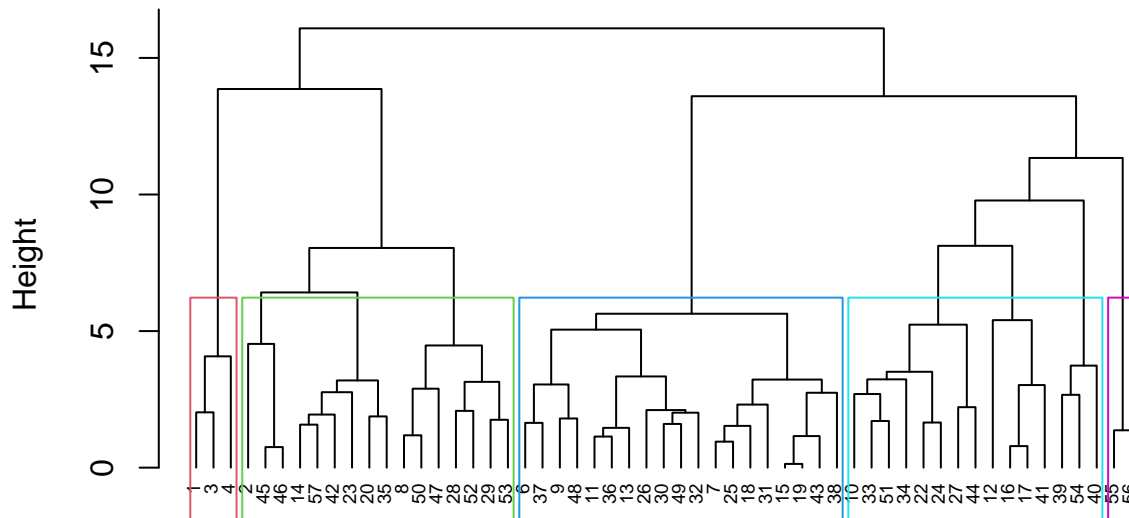
```
##          single complete average      ward
```

```
## [1,] 0.6564842 0.8120228 0.7449303 0.8808195
```

```
pltree(ward_rm, cex = 0.6, hang = -1, main = "Dendrogram of Agnes with Partitioned Data (Using Ward)")
```

```
rect.hclust(ward_rm, k = 5, border = 2:7)
```

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



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

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

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

```
RM_result <- as.data.frame(cbind(partition_one, cut_2))
```

```
RM_result[RM_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(RM_result[RM_result$cut_2==1,])
```

```
RM_result[RM_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
```

```
centroid_2 <- colMeans(RM_result[RM_result$cut_2==2,])
RM_result[RM_result$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(RM_result[RM_result$cut_2==3,])
RM_result[RM_result$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    1.0
## 12         110      6  2   290     2   17      1    105        25      1    1.0
## 16         110      2  0   280     0   22      3     25        25      1    1.0
## 17         100      2  0   290     1   21      2     35        25      1    1.0
## 22         110      2  0   220     1   21      3     30        25      3    1.0
## 24         100      2  0   190     1   18      5     80        25      3    1.0
## 27         100      3  0     0     3   14      7    100        25      2    1.0
## 33         100      3  1   140     3   15      5     85        25      3    1.0
## 34         110      3  0   170     3   17      3     90        25      3    1.0
## 39         110      2  1   170     1   17      6     60       100      3    1.0
## 40         140      3  1   170     2   20      9     95       100      3    1.3
## 41         110      2  1   260     0   21      3     40        25      2    1.0
## 44         100      4  1     0     0   16      3     95        25      2    1.0
## 51          90      3  0   170     3   18      2     90        25      3    1.0
## 54         100      3  0   320     1   20      3     45       100      3    1.0
```

```
##      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
## 39 1.00 36.52368      4
## 40 0.75 36.47151      4
## 41 1.50 39.24111      4
## 44 1.00 54.85092      4
## 51 1.00 59.64284      4
## 54 1.00 41.50354      4
```

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

```
#figuring out the Distance.
Dist_1 <- get_dist(x2)
Matrix_1 <- as.matrix(Dist_1)
```

```
dataframe1 <- data.frame(data=seq(1,nrow(partition_two),1), Clusters = rep(0,nrow(partition_two)))
for(i in 1:nrow(partition_two))
{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         4
## 6      6         2
## 7      7         2
## 8      8         2
## 9      9         3
## 10     10        4
## 11     11        2
## 12     12        4
## 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$R_Group[56:74], dataframe1$Clusters)
```

```
##      [,1] [,2]
## [1,]    2    1
## [2,]    2    2
## [3,]    5    2
## [4,]    4    3
## [5,]    4    4
## [6,]    5    2
## [7,]    5    2
## [8,]    5    2
## [9,]    3    3
## [10,]   4    4
## [11,]   5    2
## [12,]   4    4
## [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$R_Group[56:74] == dataframe1$Clusters)
```

```
##
## FALSE  TRUE
##      7    12
```

#From the above observation, we are getting 7 False and 12 True. Hence, we can conclude that the model is



partially stable.

**TASK 4** - The elementary public schools would like to choose a set of SB\_Cereals to include in their daily cafeterias. Every day a different cereal is offered, but all SB\_Cereals should support a healthy diet. For this goal, you are requested to find a cluster of “healthy Cereals”

```
#Clustering Healthy_RM_Cereals.
```

```
Healthy_RM_Cereals <- RM_Cereals
```

```
Healthy_RM_Cereals_RD <- na.omit(Healthy_RM_Cereals)
```

```
clust <- cbind(Healthy_RM_Cereals_RD, R_Group)
```

```
clust[clust$R_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 R_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$R_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 R_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$R_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 R_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$R_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 R_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$R_Group==1,"rating"])
```

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

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

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

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

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

```
mean(clust[clust$R_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.