

Assignment5

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First load the required packages using library function.

Data Pre-processing

Determine the amount of missing values and either eliminate or omit them.

```
Cereals_Data1 <- read.csv("C:/Users/lavak/Documents/R/Assignment5/Cereals.csv")
Cereals1<-read.csv("C:/Users/lavak/Documents/R/Assignment5/Cereals.csv")
str(Cereals_Data1)
```

```
## 'data.frame': 77 obs. of 16 variables:
## $ name : chr "100%_Bran" "100%_Natural_Bran" "All-Bran" "All-Bran_with_Extra_Fiber" ...
## $ mfr : chr "N" "Q" "K" "K" ...
## $ type : chr "C" "C" "C" "C" ...
## $ calories: int 70 120 70 50 110 110 110 130 90 90 ...
## $ protein : int 4 3 4 4 2 2 2 3 2 3 ...
## $ fat : int 1 5 1 0 2 2 0 2 1 0 ...
## $ sodium : int 130 15 260 140 200 180 125 210 200 210 ...
## $ fiber : num 10 2 9 14 1 1.5 1 2 4 5 ...
## $ carbo : num 5 8 7 8 14 10.5 11 18 15 13 ...
## $ sugars : int 6 8 5 0 8 10 14 8 6 5 ...
## $ potass : int 280 135 320 330 NA 70 30 100 125 190 ...
## $ vitamins: int 25 0 25 25 25 25 25 25 25 ...
## $ shelf : int 3 3 3 3 3 1 2 3 1 3 ...
## $ weight : num 1 1 1 1 1 1 1 1.33 1 1 ...
## $ cups : num 0.33 1 0.33 0.5 0.75 0.75 1 0.75 0.67 0.67 ...
## $ rating : num 68.4 34 59.4 93.7 34.4 ...
```

```
sum(is.na(Cereals_Data1))
```

```
## [1] 4
```

To eliminate any missing values from the data, enter the following:

```
Cereals_Data1 <- na.omit(Cereals_Data1)
Cereals1<-na.omit(Cereals1)
sum(is.na(Cereals_Data1))
```

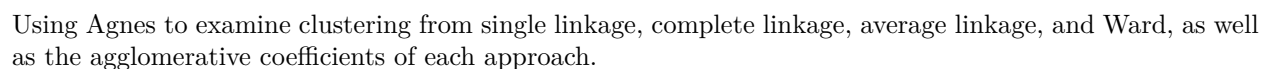
```
## [1] 0
```

```
rownames(Cereals_Data1) <- Cereals_Data1$name
rownames(Cereals1) <- Cereals1$name
```

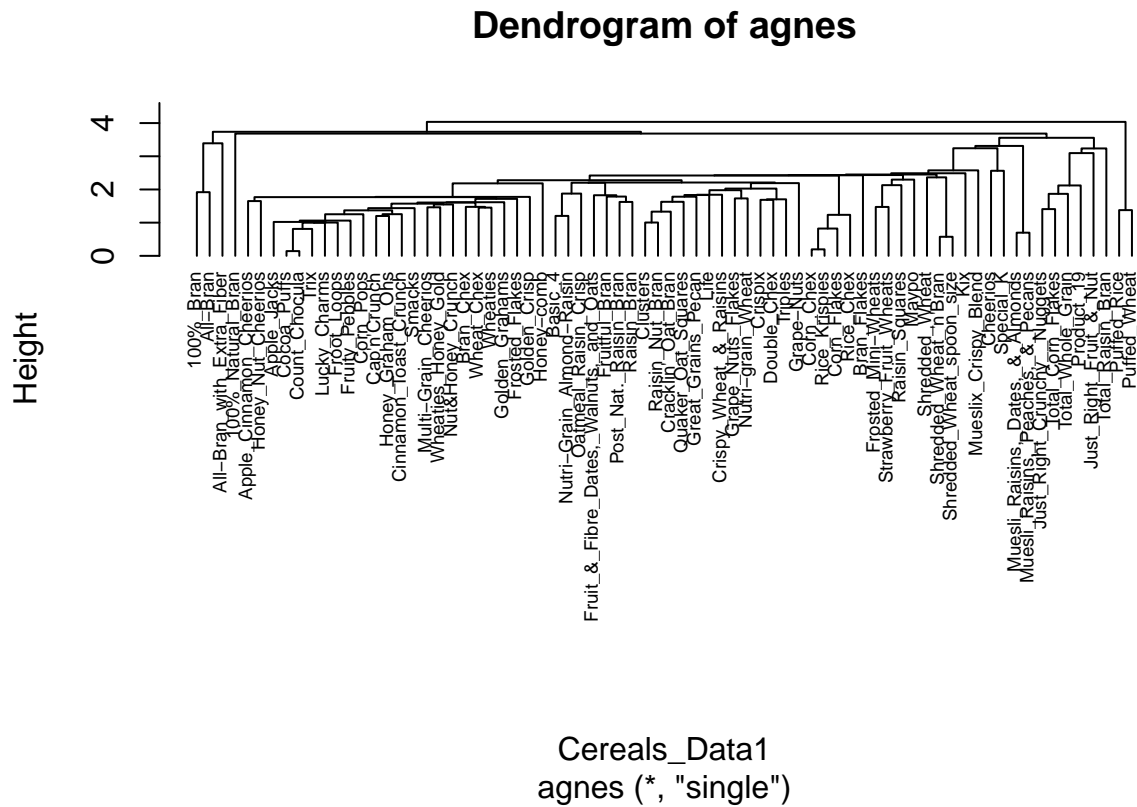
```
Cereals_Data1$name = NULL
Cereals1$name = NULL
```

```
Cereals_Data1 <- scale(Cereals_Data1[,3:15])
```

```
# Dissimilarity matrix
d <- dist(Cereals_Data1, method = "euclidean")
# Hierarchical clustering using Complete Linkage
HC_comp <- hclust(d, method = "complete" )
# Plot the obtained dendrogram
plot(HC_comp, cex = 0.6, hang = -1)
```

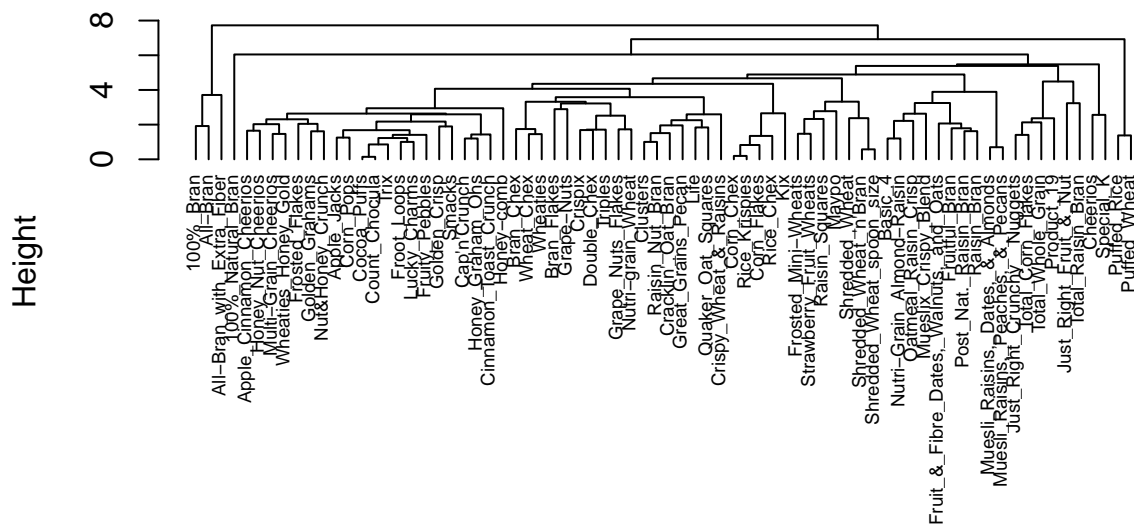


```
library(cluster)
HC_single1 <- agnes(Cereals_Data1, method = "single")
pltree(HC_single1, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



```
HC_avg <- agnes(Cereals_Data1, method = "average")
pltree(HC_avg, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```

Dendrogram of agnes



```
Cereals_Data1
agnes (*, "average")
```

We shall calculate the agnes coefficient for each approach.

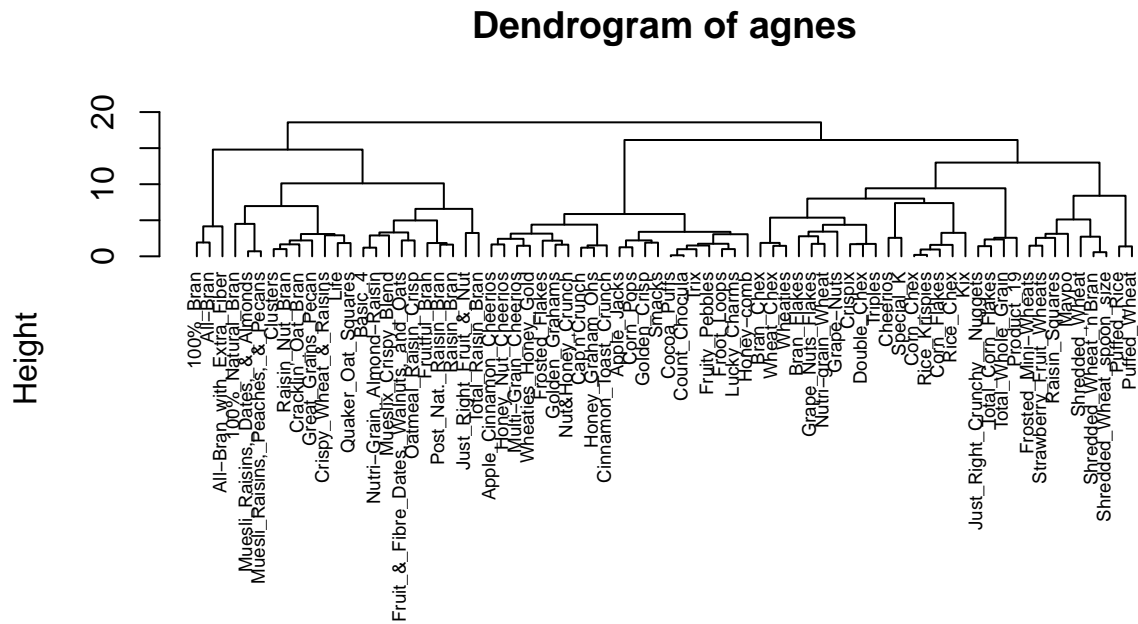
```
# methods to assess
m <- c( "average", "single", "complete", "ward")
names(m) <- c( "average", "single", "complete", "ward")
# function to compute coefficient
ac <- function(x) {
  agnes(Cereals_Data1, method = x)$ac
}
map_dbl(m, ac)
```

```
##      average      single  complete      ward
## 0.7766075 0.6067859 0.8353712 0.9046042
```

Ward is the best linking method, with an agglomerative coefficient of 0.9046042.

Using the wards approach to visualize the dendrogram:

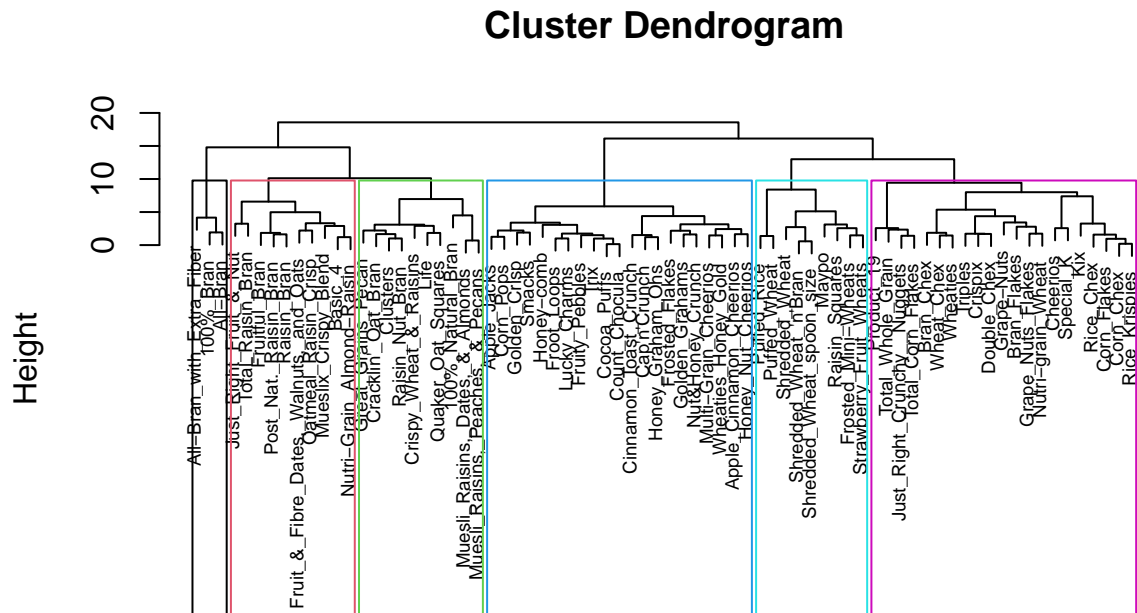
```
HC_Wards <- agnes(Cereals_Data1, method = "ward")
pltree(HC_Wards, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



Cereals_Data1
agnes (*, "ward")

Cut the dendrogram with `cutree()` to find sub-groups (i.e. clusters):

```
#Create the distance matrix
d <- dist(Cereals_Data1, method = "euclidean")
# Ward's method for Hierarchical clustering
HC_Ward_clust <- hclust(d, method = "ward.D2" )
plot(HC_Ward_clust, cex=0.6 )
rect.hclust(HC_Ward_clust,k=6,border = 1:6)
```



d
hclust (*, "ward.D2")

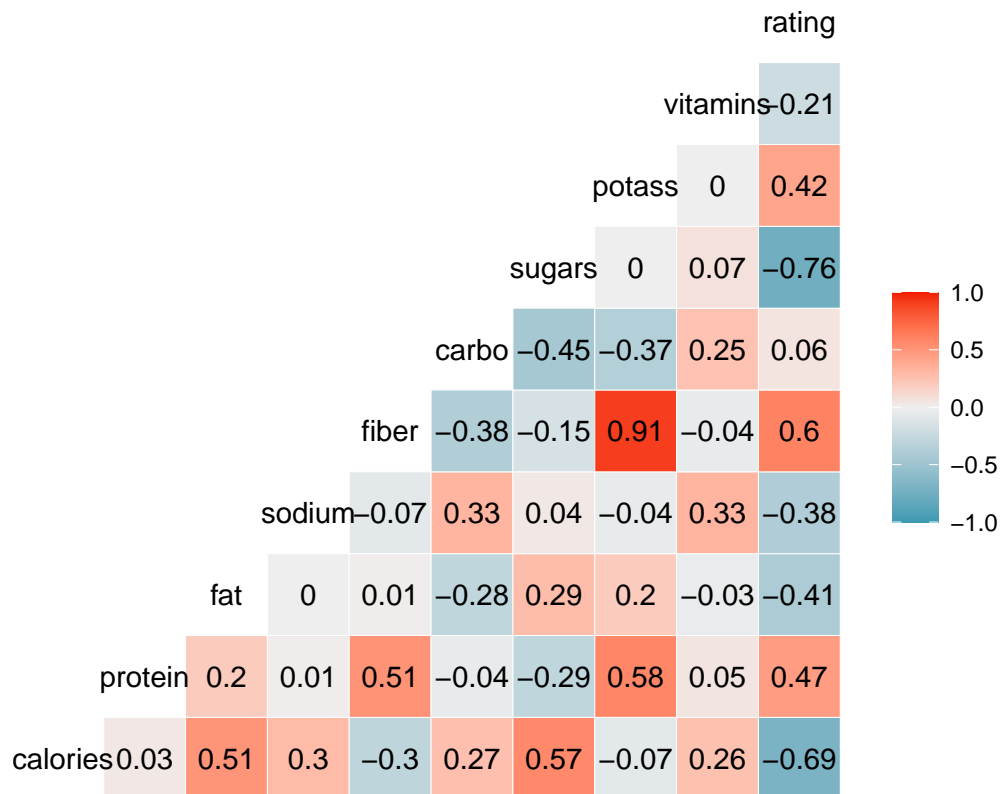
Let's examine how many data records have been categorized and allocated to clusters:

```
# Cut tree into 6 groups
sub_grup <- cutree(HC_Ward_clust, k = 6)
# Number of members in each cluster
table(sub_grup)
```

```
## sub_grup
##  1  2  3  4  5  6
##  3 10 21 10 21  9
```

Correlation matrix:

```
#install.packages("GGally")
Cereals1 %>%
  select(calories, protein, fat, sodium, fiber, carbo, sugars, potass, vitamins, rating) %>%
  ggcorr(palette = "RdBu", label = TRUE, label_round = 2)
```



The correlation matrix assists us in determining if there is a strong or weak relationship between the variables. This will provide us with a better perspective for calculating descriptive statistics between variables.

The `pvclust()` method from the `pvclust` package returns p-values for hierarchical clustering using multiscale bootstrap resampling. Large p values will be assigned to clusters that are strongly supported by the data. Suzuki provides interpretation information. Keep in mind that `pvclust` clusters columns rather of rows. Before you use your data, make sure you transpose it.

```
# Ward Hierarchical Clustering with Bootstrapped p values
#install.packages("pvclust")
library(pvclust)
```

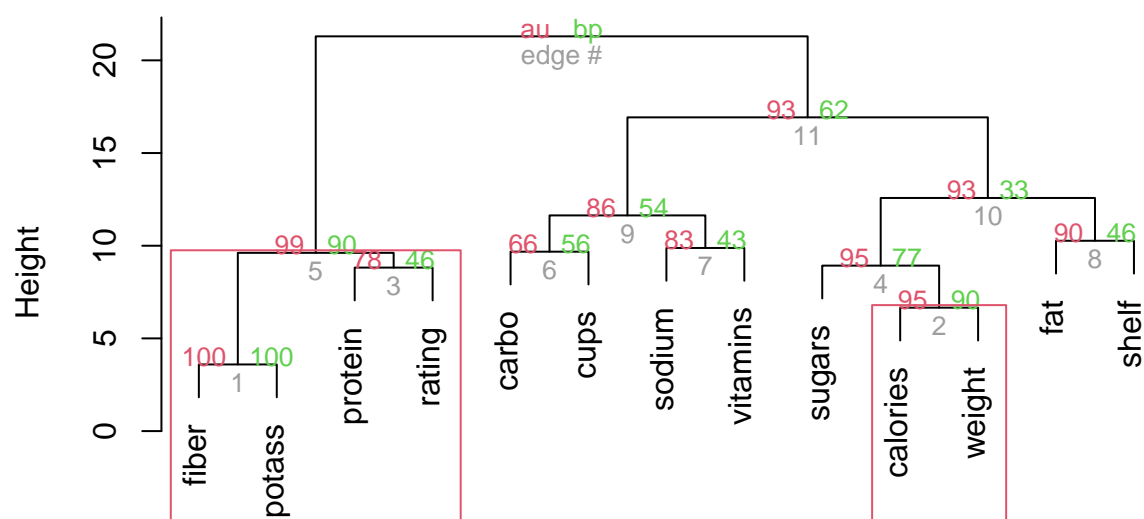
```
## Warning: package 'pvclust' was built under R version 4.1.3
```

```
## Registered S3 method overwritten by 'pvclust':
##   method      from
##   text.pvclust dendextend
```

```
fit.pv <- pvclust(Cereals_Data1, method.hclust="ward.D2",
                  method.dist="euclidean")
```

```
plot(fit.pv) # dendrogram with p values
# add rectangles around groups highly supported by the data
pvrect(fit.pv, alpha=.95)
```

Cluster dendrogram with p-values (%)



Distance: euclidean
Cluster method: ward.D2

In the initial clustering, the cluster stability of each cluster is the mean value of its Jaccard coefficient over all bootstrap iterations. Clusters having a stability rating of less than 0.6 should be deemed unstable. Values between 0.6 and 0.75 show that the cluster is detecting a pattern in the data, but there isn't a lot of conviction regarding which points should be grouped together. Clusters with stability ratings greater than 0.85 are regarded extremely stable.

1. The mean of the clusterwise Jaccard bootstrap should be maximized.
2. The number of dissolved clusters should be kept to a minimum.
3. The number of recovered clusters should be maximized while remaining as close to the number of pre-specified bootstraps as feasible.

```
#Running clusterboot()
```

```
library(fpc)
```

```
## Warning: package 'fpc' was built under R version 4.1.3
```

```
library(cluster)
```

```
Kbest_p<-6
```

```
cboot_hclust <- clusterboot(Cereals_Data1,clustermethod=hclustCBI,method="ward.D2", k=Kbest_p)
```

```
summary(cboot_hclust$result)
```

```
##           Length Class  Mode
```



```
## result      7      hclust list
## noise       1      -none- logical
## nc          1      -none- numeric
## clusterlist 6      -none- list
## partition   74     -none- numeric
## clustermethod 1    -none- character
## nccl        1      -none- numeric
```

```
groups<-cboot_hclust$result$partition
head(data.frame(groups))
```

```
##                groups
## 100%_Bran          1
## 100%_Natural_Bran  2
## All-Bran           1
## All-Bran_with_Extra_Fiber 1
## Apple_Cinnamon_Cheerios 3
## Apple_Jacks        3
```

```
#The vector of cluster stabilities
cboot_hclust$bootmean
```

```
## [1] 0.8861899 0.4725252 0.8923053 0.6919398 0.5837242 0.6673889
```

```
#The count of how many times each cluster was dissolved. By default clusterboot() runs 100 bootstrap iterations.
cboot_hclust$bootbrd
```

```
## [1] 12 68 0 28 36 39
```

Based on the output, we may conclude that clusters 1 and 3 are highly stable. Clusters 4 and 5 are measuring a pattern, but there isn't a lot of agreement on which points should be grouped together. Clusters 2 and 5 are in a state of instability.

Extracting the clusters found by hclust()

```
groups <- cutree(HC_Ward_clust, k = 6)
print_clusters <- function(labels, k) {
  for(i in 1:k) {
    print(paste("cluster", i))
    print(Cereals1[labels==i,c("mfr","calories","protein","fat","sodium","fiber","carbo","sugars","potass",
                              "vitamins","rating")])
  }
}
print_clusters(groups, 6)
```

```
## [1] "cluster 1"
##                mfr calories protein fat sodium fiber carbo sugars
## 100%_Bran       N      70        4  1    130    10     5      6
## All-Bran        K      70        4  1    260     9     7      5
## All-Bran_with_Extra_Fiber K    50        4  0    140    14     8      0
```

```

##                potass vitamins    rating
## 100%_Bran        280        25 68.40297
## All-Bran        320        25 59.42551
## All-Bran_with_Extra_Fiber 330        25 93.70491
## [1] "cluster 2"
##                mfr calories protein fat sodium fiber carbo
## 100%_Natural_Bran    Q      120        3  5      15  2.0  8.0
## Clusters            G      110        3  2     140  2.0 13.0
## Cracklin'_Oat_Bran   K      110        3  3     140  4.0 10.0
## Crispy_Wheat_&_Raisins G      100        2  1     140  2.0 11.0
## Great_Grains_Pecan   P      120        3  3      75  3.0 13.0
## Life                Q      100        4  2     150  2.0 12.0
## Muesli_Raisins,_Dates,_&_Almonds R      150        4  3      95  3.0 16.0
## Muesli_Raisins,_Peaches,_&_Pecans R      150        4  3     150  3.0 16.0
## Quaker_Oat_Squares   Q      100        4  1     135  2.0 14.0
## Raisin_Nut_Bran      G      100        3  2     140  2.5 10.5
##                sugars potass vitamins    rating
## 100%_Natural_Bran      8     135        0 33.98368
## Clusters              7     105        25 40.40021
## Cracklin'_Oat_Bran      7     160        25 40.44877
## Crispy_Wheat_&_Raisins 10     120        25 36.17620
## Great_Grains_Pecan      4     100        25 45.81172
## Life                  6      95        25 45.32807
## Muesli_Raisins,_Dates,_&_Almonds 11     170        25 37.13686
## Muesli_Raisins,_Peaches,_&_Pecans 11     170        25 34.13976
## Quaker_Oat_Squares      6     110        25 49.51187
## Raisin_Nut_Bran        8     140        25 39.70340
## [1] "cluster 3"
##                mfr calories protein fat sodium fiber carbo sugars
## Apple_Cinnamon_Cheerios G      110        2  2     180  1.5 10.5    10
## Apple_Jacks            K      110        2  0     125  1.0 11.0    14
## Cap'n'_Crunch          Q      120        1  2     220  0.0 12.0    12
## Cinnamon_Toast_Crunch   G      120        1  3     210  0.0 13.0     9
## Cocoa_Puffs            G      110        1  1     180  0.0 12.0    13
## Corn_Pops              K      110        1  0      90  1.0 13.0    12
## Count_Chocula          G      110        1  1     180  0.0 12.0    13
## Froot_Loops            K      110        2  1     125  1.0 11.0    13
## Frosted_Flakes         K      110        1  0     200  1.0 14.0    11
## Fruity_Pebbles         P      110        1  1     135  0.0 13.0    12
## Golden_Crisp           P      100        2  0      45  0.0 11.0    15
## Golden_Grahams         G      110        1  1     280  0.0 15.0     9
## Honey_Graham_Ohs       Q      120        1  2     220  1.0 12.0    11
## Honey_Nut_Cheerios     G      110        3  1     250  1.5 11.5    10
## Honey-comb             P      110        1  0     180  0.0 14.0    11
## Lucky_Charms           G      110        2  1     180  0.0 12.0    12
## Multi-Grain_Cheerios   G      100        2  1     220  2.0 15.0     6
## Nut&Honey_Crunch       K      120        2  1     190  0.0 15.0     9
## Smacks                 K      110        2  1      70  1.0  9.0    15
## Trix                   G      110        1  1     140  0.0 13.0    12
## Wheaties_Honey_Gold    G      110        2  1     200  1.0 16.0     8
##                potass vitamins    rating
## Apple_Cinnamon_Cheerios 70        25 29.50954
## Apple_Jacks             30        25 33.17409
## Cap'n'_Crunch           35        25 18.04285

```

```

## Cinnamon_Toast_Crunch      45      25 19.82357
## Cocoa_Puffs                 55      25 22.73645
## Corn_Pops                   20      25 35.78279
## Count_Chocula               65      25 22.39651
## Froot_Loops                 30      25 32.20758
## Frosted_Flakes              25      25 31.43597
## Fruity_Pebbles              25      25 28.02576
## Golden_Crisp                40      25 35.25244
## Golden_Grahams              45      25 23.80404
## Honey_Graham_Ohs            45      25 21.87129
## Honey_Nut_Cheerios          90      25 31.07222
## Honey-comb                  35      25 28.74241
## Lucky_Charms                55      25 26.73451
## Multi-Grain_Cheerios        90      25 40.10596
## Nut&Honey_Crunch            40      25 29.92429
## Smacks                      40      25 31.23005
## Trix                        25      25 27.75330
## Wheaties_Honey_Gold         60      25 36.18756
## [1] "cluster 4"
##
##                               mfr calories protein fat sodium fiber
## Basic_4                      G      130      3   2    210    2.0
## Fruit_&Fibre_Dates,_Walnuts,_and_Oats P      120      3   2    160    5.0
## Fruitful_Bran                 K      120      3   0    240    5.0
## Just_Right_Fruit_&Nut         K      140      3   1    170    2.0
## Mueslix_Crispy_Blend          K      160      3   2    150    3.0
## Nutri-Grain_Almond-Raisin     K      140      3   2    220    3.0
## Oatmeal_Raisin_Crisp          G      130      3   2    170    1.5
## Post_Nat._Raisin_Bran         P      120      3   1    200    6.0
## Raisin_Bran                  K      120      3   1    210    5.0
## Total_Raisin_Bran            G      140      3   1    190    4.0
##
##                               carbo sugars potass vitamins  rating
## Basic_4                      18.0      8    100      25 37.03856
## Fruit_&Fibre_Dates,_Walnuts,_and_Oats 12.0     10    200      25 40.91705
## Fruitful_Bran                 14.0     12    190      25 41.01549
## Just_Right_Fruit_&Nut         20.0      9     95     100 36.47151
## Mueslix_Crispy_Blend          17.0     13    160      25 30.31335
## Nutri-Grain_Almond-Raisin     21.0      7    130      25 40.69232
## Oatmeal_Raisin_Crisp          13.5     10    120      25 30.45084
## Post_Nat._Raisin_Bran         11.0     14    260      25 37.84059
## Raisin_Bran                  14.0     12    240      25 39.25920
## Total_Raisin_Bran            15.0     14    230     100 28.59278
## [1] "cluster 5"
##
##                               mfr calories protein fat sodium fiber carbo sugars
## Bran_Chex                     R      90      2   1    200      4    15      6
## Bran_Flakes                   P      90      3   0    210      5    13      5
## Cheerios                      G     110      6   2    290      2    17      1
## Corn_Chex                     R     110      2   0    280      0    22      3
## Corn_Flakes                   K     100      2   0    290      1    21      2
## Crispix                       K     110      2   0    220      1    21      3
## Double_Chex                   R     100      2   0    190      1    18      5
## Grape_Nuts_Flakes              P     100      3   1    140      3    15      5
## Grape-Nuts                    P     110      3   0    170      3    17      3
## Just_Right_Crunchy__Nuggets    K     110      2   1    170      1    17      6
## Kix                           G     110      2   1    260      0    21      3

```

## Nutri-grain_Wheat	K	90	3	0	170	3	18	2
## Product_19	K	100	3	0	320	1	20	3
## Rice_Chex	R	110	1	0	240	0	23	2
## Rice_Krispies	K	110	2	0	290	0	22	3
## Special_K	K	110	6	0	230	1	16	3
## Total_Corn_Flakes	G	110	2	1	200	0	21	3
## Total_Whole_Grain	G	100	3	1	200	3	16	3
## Triples	G	110	2	1	250	0	21	3
## Wheat_Chex	R	100	3	1	230	3	17	3
## Wheaties	G	100	3	1	200	3	17	3
##	potass	vitamins	rating					
## Bran_Chex	125	25	49.12025					
## Bran_Flakes	190	25	53.31381					
## Cheerios	105	25	50.76500					
## Corn_Chex	25	25	41.44502					
## Corn_Flakes	35	25	45.86332					
## Crispix	30	25	46.89564					
## Double_Chex	80	25	44.33086					
## Grape_Nuts_Flakes	85	25	52.07690					
## Grape-Nuts	90	25	53.37101					
## Just_Right_Crunchy__Nuggets	60	100	36.52368					
## Kix	40	25	39.24111					
## Nutri-grain_Wheat	90	25	59.64284					
## Product_19	45	100	41.50354					
## Rice_Chex	30	25	41.99893					
## Rice_Krispies	35	25	40.56016					
## Special_K	55	25	53.13132					
## Total_Corn_Flakes	35	100	38.83975					
## Total_Whole_Grain	110	100	46.65884					
## Triples	60	25	39.10617					
## Wheat_Chex	115	25	49.78744					
## Wheaties	110	25	51.59219					
## [1] "cluster 6"								
##	mfr	calories	protein	fat	sodium	fiber	carbo	sugars
## Frosted_Mini-Wheats	K	100	3	0	0	3	14	7
## Maypo	A	100	4	1	0	0	16	3
## Puffed_Rice	Q	50	1	0	0	0	13	0
## Puffed_Wheat	Q	50	2	0	0	1	10	0
## Raisin_Squares	K	90	2	0	0	2	15	6
## Shredded_Wheat	N	80	2	0	0	3	16	0
## Shredded_Wheat_'n'Bran	N	90	3	0	0	4	19	0
## Shredded_Wheat_spoon_size	N	90	3	0	0	3	20	0
## Strawberry_Fruit_Wheats	N	90	2	0	15	3	15	5
##	potass	vitamins	rating					
## Frosted_Mini-Wheats	100	25	58.34514					
## Maypo	95	25	54.85092					
## Puffed_Rice	15	0	60.75611					
## Puffed_Wheat	50	0	63.00565					
## Raisin_Squares	110	25	55.33314					
## Shredded_Wheat	95	0	68.23588					
## Shredded_Wheat_'n'Bran	140	0	74.47295					
## Shredded_Wheat_spoon_size	120	0	72.80179					
## Strawberry_Fruit_Wheats	90	25	59.36399					

Note***

Because there is no mention of an appropriate measure/scale to construct a healthy diet, I chose to pick clusters based on statistical values and rich in nutritional values to form a healthy diet, which is totally subjective.

To determine whether or not normalization was required. No, I would say. When we normalize the data, the magnitude of the data is gone, making it exceedingly difficult to interpret and decide.

The cereal diet levels in the clusters are nutritionally rich, sufficient, and poor. We divided all of the data into six groups, and we will analyze these clusters based on all of the variables/factors.

Despite the fact that Cluster 1 contains nutritionally consistent parameters for forming a balanced diet, the possibilities are quite restricted. Clusters 2 and 3 have low ratings and high fat and sugar content, which is not ideal for a healthy lunch.

Clusters 4 and 5 offer well-balanced nutritional values and high customer satisfaction scores. Hence Clusters 4 and 5 should be ideal choices for primary public schools to include this into their cafeterias.