

# FML ASSIGNMENT 5

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```
Cereals <- read.csv("C:/Users/srich/OneDrive/Desktop/Cereals.csv")
Cereals_1<-read.csv("C:/Users/srich/OneDrive/Desktop/Cereals.csv")
# Displays the structure of the dataset
str(Cereals)
```

```
## '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 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 ...
```

```
# Displays the first 6 rows of the "Cereals" dataset
head(Cereals)
```

```
##           name mfr type calories protein fat sodium fiber carbo
## 1      100%_Bran   N    C        70         4  1   130   10.0   5.0
## 2    100%_Natural_Bran Q    C       120        3  5    15    2.0   8.0
## 3          All-Bran   K    C        70        4  1   260    9.0   7.0
## 4 All-Bran_with_Extra_Fiber K    C        50        4  0   140   14.0   8.0
## 5      Almond_Delight   R    C       110        2  2   200    1.0  14.0
## 6 Apple_Cinnamon_Cheerios G    C       110        2  2   180    1.5  10.5
##  sugars potass vitamins shelf weight cups  rating
## 1      6     280       25     3      1 0.33 68.40297
## 2      8     135        0     3      1 1.00 33.98368
## 3      5     320       25     3      1 0.33 59.42551
## 4      0     330       25     3      1 0.50 93.70491
## 5      8      NA       25     3      1 0.75 34.38484
## 6     10      70       25     1      1 0.75 29.50954
```

```
# Calculates and returns the total count of missing values in the "Cereals" dataset  
sum(is.na(Cereals))
```

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

```
# Removes rows with missing values from the "Cereals" dataset  
Cereals <- na.omit(Cereals)  
# Removes rows with missing values from the "Cereals_1" dataset  
Cereals_1 <-na.omit(Cereals_1)  
# Calculates the sum of missing values in the "Cereals" data and print the result  
sum(is.na(Cereals))
```

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

```
# Convert the names of the cereals to row names  
rownames(Cereals) <- Cereals$name  
rownames(Cereals_1) <- Cereals_1$name
```

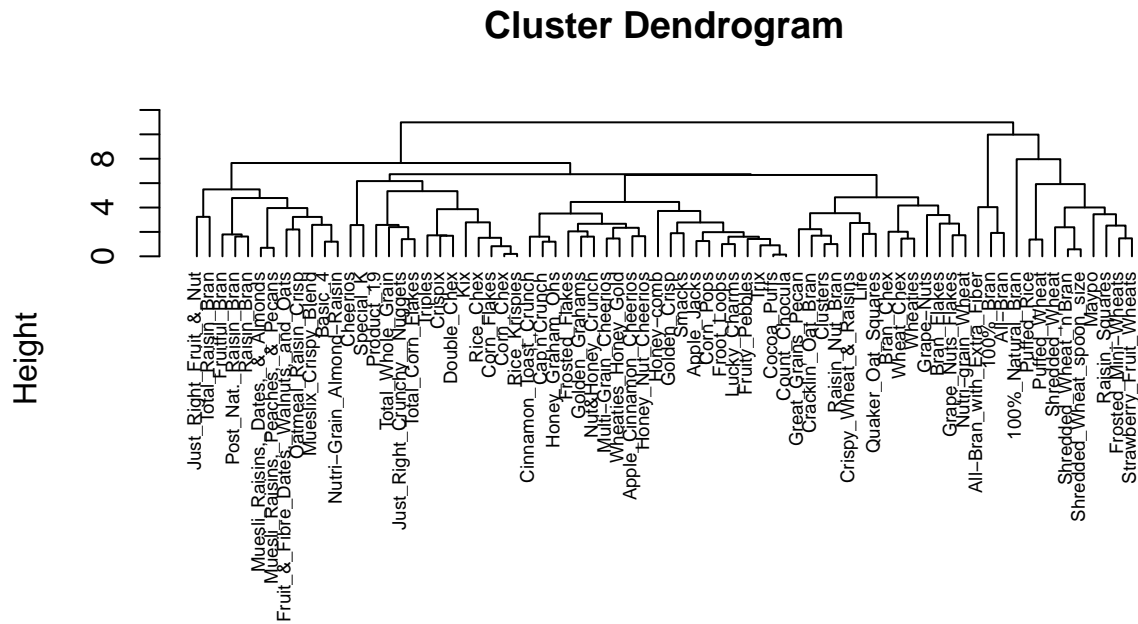
```
# Sets the "name" variable in the "Cereals" dataset to NULL  
# Sets the "name" variable in the "Cereals_1" dataset to NULL  
Cereals$name = NULL  
Cereals_1$name = NULL
```

```
# Scales the variables in columns 3 to 15 of the "Cereals" dataset  
Cereals <- scale(Cereals[,3:15])
```

```
# Calculates Euclidean distance matrix for the "Cereals" dataset  
dis <- dist(Cereals, method = "euclidean")
```

```
# Performs hierarchical clustering using complete linkage method  
hc_comp <- hclust(dis, method = "complete" )
```

```
# Plotting hierarchical clustering dendrogram  
plot(hc_comp, cex = 0.6, hang = -1)
```

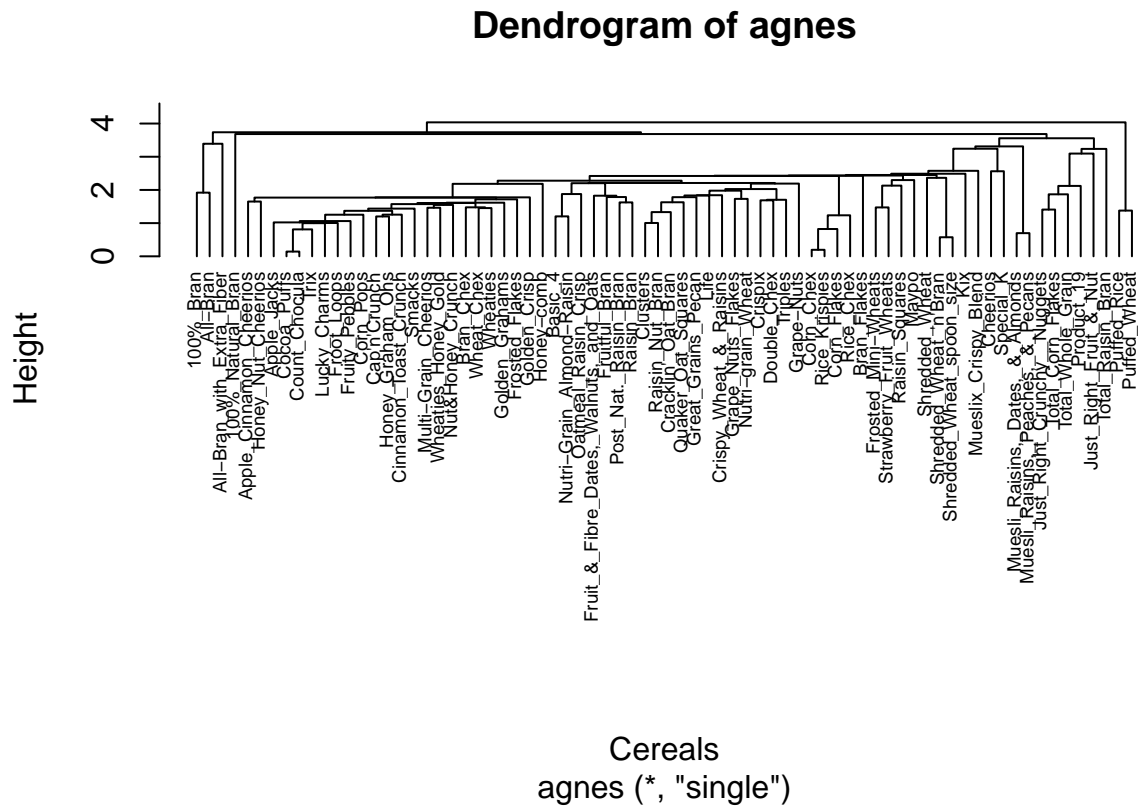


```
dis
hclust(*, "complete")
```

```
# Loads the specified library
library(cluster)
```

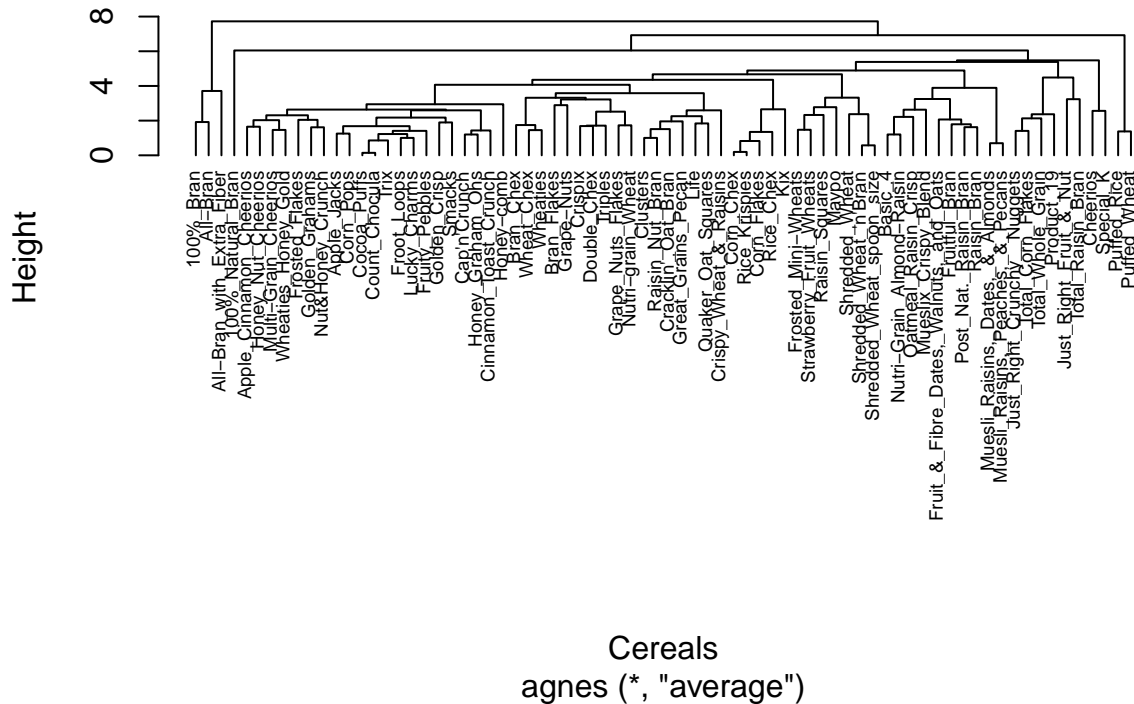
```
## Warning: package 'cluster' was built under R version 4.3.2
```

```
# Performs hierarchical clustering using the agnes function with the "single" method
hc_single1 <- agnes(Cereals, method = "single")
# Visualizes the dendrogram using the pltree function
pltree(hc_single1, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



```
# Performs hierarchical clustering
hc_avg <- agnes(Cereals, method = "average")
# Plots the dendrogram
pltree(hc_avg, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```

## Dendrogram of agnes



```
# Defines a vector "m" with clustering method names
m <- c( "average", "single", "complete", "ward")
# Assigns names to the vector elements
names(m) <- c( "average", "single", "complete", "ward")
# Defines a function "ac" that takes a clustering method as an argument and returns the agglomerative c
ac <- function(x) {
  agnes(Cereals, method = x)$ac
}
#install.packages("purrr")
library(purrr)
```

```
## Warning: package 'purrr' was built under R version 4.3.2
```

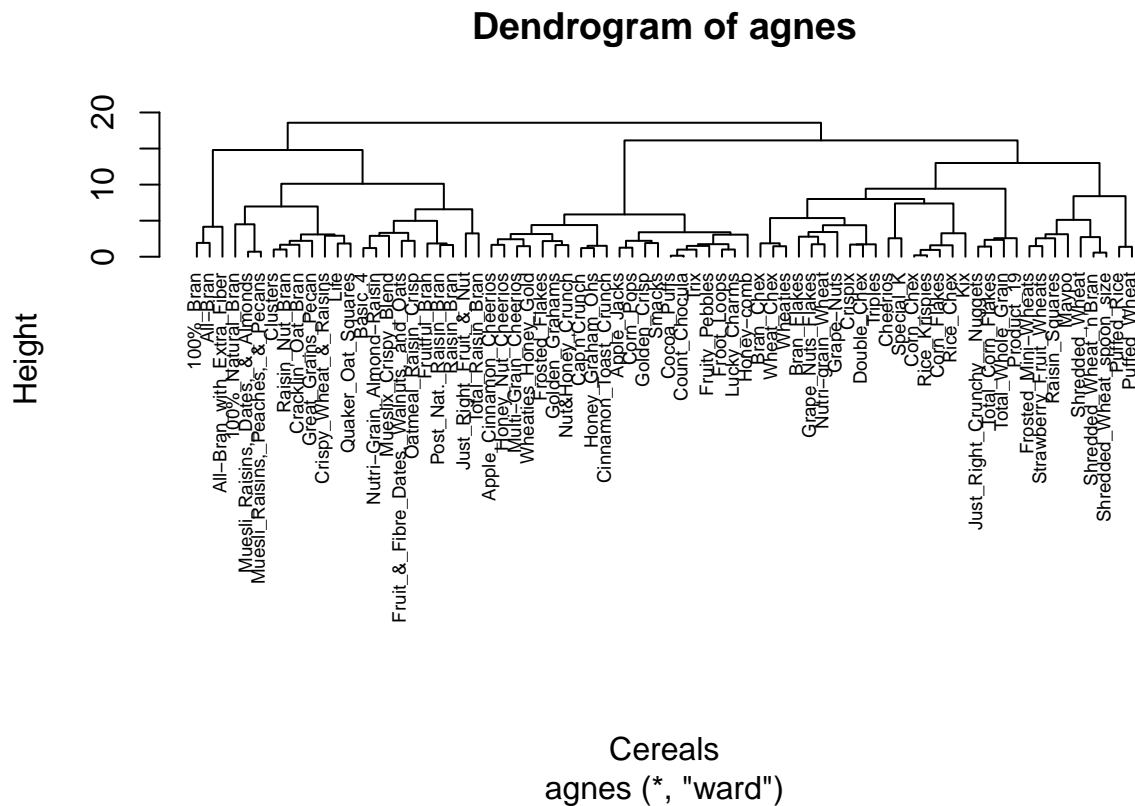
```
map_dbl(m, ac)
```

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

```
### Based on the obtained coefficients, we can see that the ward linkage method has the highest coefficient
### Ward linkage method is the best method for clustering the cereal data based on Euclidean distance
```

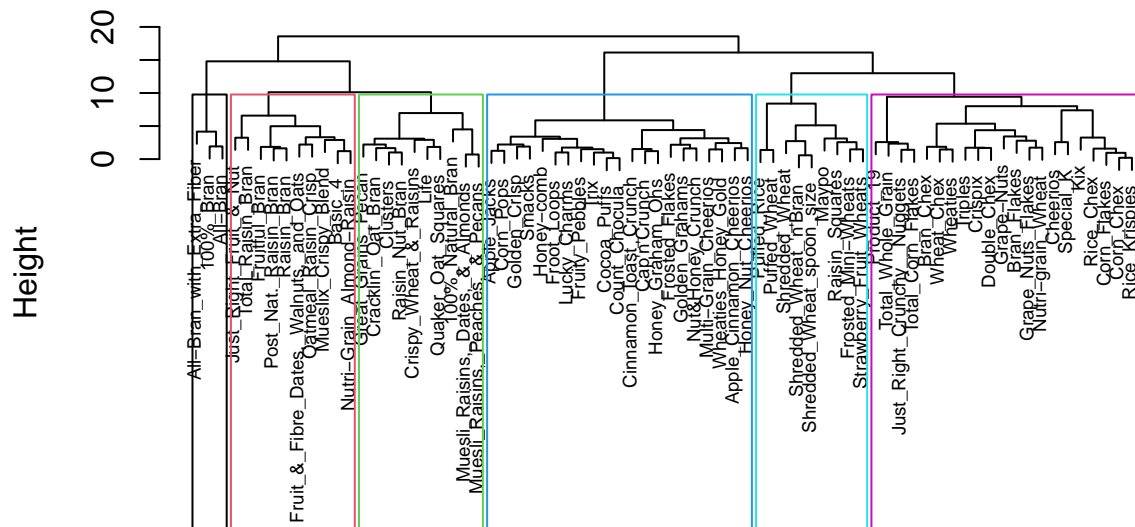
```
# Performs hierarchical clustering on the "Cereals" dataset using the agnes function with the "ward" me
```

```
hc_wards <- agnes(Cereals, method = "ward")
# Plots the dendrogram based on the hierarchical clustering results
# Adjusting the text size (cex), hang distance, and adding title to the plot.
pltree(hc_wards, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



```
# Calculates the Euclidean distance matrix for the "Cereals" dataset and store it in "d"
d <- dist(Cereals, method = "euclidean")
# Performs hierarchical clustering using the Ward's method on the distance matrix "d"
hc_ward_clust <- hclust(d, method = "ward.D2" )
# Plots the hierarchical clustering dendrogram with reduced text size
plot(hc_ward_clust, cex=0.6 )
# Highlights clusters by drawing rectangles around them where "k=6" specifies 6 clusters
rect.hclust(hc_ward_clust,k=6,border = 1:6)
```

## Cluster Dendrogram



d  
hclust(\*, "ward.D2")

### 6 Clusters appear to be a good number to group the data using the Ward linkage

# Using hierarchical clustering with the Ward's method (hc\_ward\_clust) to create clusters, and then ass  
# Displays the table of counts for each subgroup.

```
sub_group <- cutree(hc_ward_clust, k = 6)
table(sub_group)
```

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

```
#install.packages("GGally")
# Load the required library
library(GGally)
```

```
## Warning: package 'GGally' was built under R version 4.3.2
```

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

```
## Warning: package 'ggplot2' was built under R version 4.3.2
```

```
## Registered S3 method overwritten by 'GGally':
## method from
## +.gg ggplot2
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 4.3.2
```

```
##
```

```
## 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
```

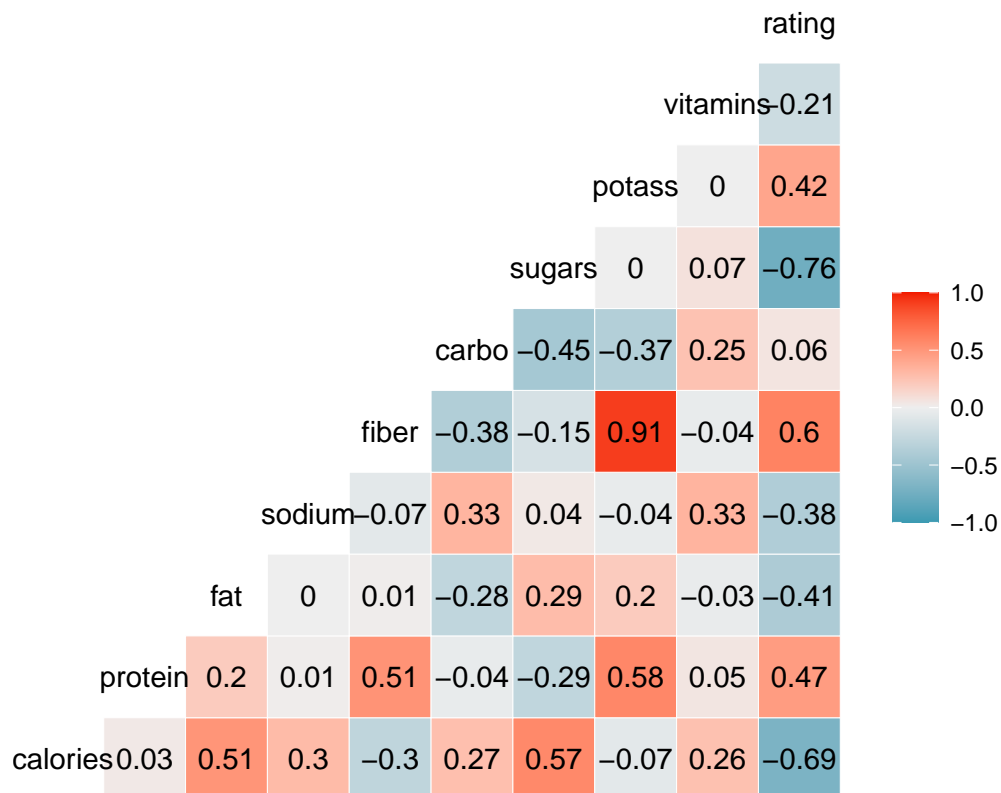
```
# Using the Cereals_1 data frame and selecting specific columns related to nutrition and rating  
# Adjusting the data frame name and column names based on actual data
```

```
Cereals_1 %>%
```

```
  select(calories, protein, fat, sodium, fiber, carbo, sugars, potass, vitamins, rating) %>%
```

```
# Create a correlation matrix plot using ggcorr function from GGally
```

```
ggcorr(palette = "RdBu", label = TRUE, label_round = 2)
```





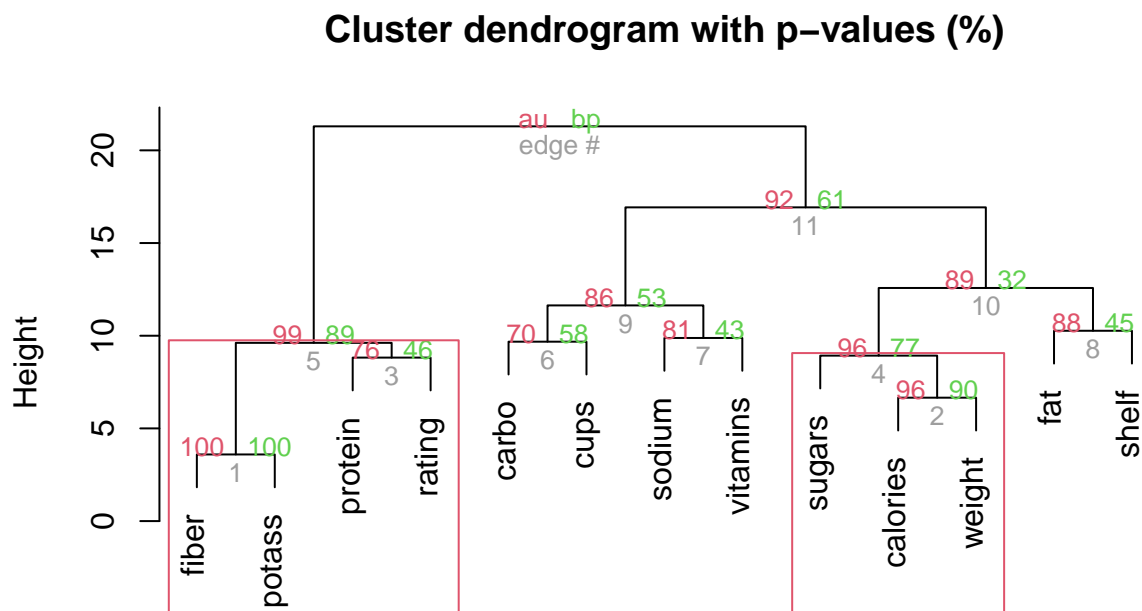
```
#install.packages("pvclust")
# Load the required library
library(pvclust)
```

```
## Warning: package 'pvclust' was built under R version 4.3.2
```

```
# Performs hierarchical clustering on the "Cereals" data using the Ward.D2 method with Euclidean distance
fit.pv <- pvclust(Cereals, method.hclust="ward.D2",
                  method.dist="euclidean")
```

```
## Bootstrap (r = 0.5)... Done.
## Bootstrap (r = 0.59)... Done.
## Bootstrap (r = 0.69)... Done.
## Bootstrap (r = 0.8)... Done.
## Bootstrap (r = 0.89)... Done.
## Bootstrap (r = 1.0)... Done.
## Bootstrap (r = 1.09)... Done.
## Bootstrap (r = 1.19)... Done.
## Bootstrap (r = 1.3)... Done.
## Bootstrap (r = 1.39)... Done.
```

```
# Plots the results of the fit.pv model using the plot function
plot(fit.pv)
pvrect(fit.pv, alpha=.95)
```



```
### The effectiveness of the initial clustering is evaluated through the cluster stability, represented  
### Stability rating between 0.6 and 0.75 suggests that a cluster identifies a pattern in the data, but  
### The goal is to maximize the Jaccard bootstrap for each cluster. Where efforts should be made to mini
```

```
#install.packages("fpc")  
# Load the required library  
library(fpc)
```

```
## Warning: package 'fpc' was built under R version 4.3.2
```

```
# Set the number of clusters to 6  
Kbest_p<-6  
# Perform hierarchical clustering using the Ward.D2 method and the hclustCBI function on the "Cereals"  
cboot_hclust <- clusterboot(Cereals,clustermethod=hclustCBI,method="ward.D2", k=Kbest_p)
```

```
## boot 1  
## boot 2  
## boot 3  
## boot 4  
## boot 5  
## boot 6  
## boot 7  
## boot 8  
## boot 9  
## boot 10  
## boot 11  
## boot 12  
## boot 13  
## boot 14  
## boot 15  
## boot 16  
## boot 17  
## boot 18  
## boot 19  
## boot 20  
## boot 21  
## boot 22  
## boot 23  
## boot 24  
## boot 25  
## boot 26  
## boot 27  
## boot 28  
## boot 29  
## boot 30  
## boot 31  
## boot 32  
## boot 33  
## boot 34  
## boot 35  
## boot 36  
## boot 37
```

## boot 38  
## boot 39  
## boot 40  
## boot 41  
## boot 42  
## boot 43  
## boot 44  
## boot 45  
## boot 46  
## boot 47  
## boot 48  
## boot 49  
## boot 50  
## boot 51  
## boot 52  
## boot 53  
## boot 54  
## boot 55  
## boot 56  
## boot 57  
## boot 58  
## boot 59  
## boot 60  
## boot 61  
## boot 62  
## boot 63  
## boot 64  
## boot 65  
## boot 66  
## boot 67  
## boot 68  
## boot 69  
## boot 70  
## boot 71  
## boot 72  
## boot 73  
## boot 74  
## boot 75  
## boot 76  
## boot 77  
## boot 78  
## boot 79  
## boot 80  
## boot 81  
## boot 82  
## boot 83  
## boot 84  
## boot 85  
## boot 86  
## boot 87  
## boot 88  
## boot 89  
## boot 90  
## boot 91

```
## boot 92
## boot 93
## boot 94
## boot 95
## boot 96
## boot 97
## boot 98
## boot 99
## boot 100
```

```
# Summarizing the results of hierarchical clustering using cboot_hclust
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
```

```
# Extracting the cluster assignments from the hierarchical clustering results stored in cboot_hclust$result
groups<-cboot_hclust$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
```

```
# Extracting the bootstrap means from the "cboot_hclust" object
cboot_hclust$bootmean
```

```
## [1] 0.8965202 0.5363575 0.9115696 0.6582853 0.6315921 0.7054986
```

```
# Count of how many times each cluster was dissolved and by default clusterboot() runs 100 bootstrap iterations
# Accessing the "bootbrd" column in the "cboot_hclust" data frame
cboot_hclust$bootbrd
```

```
## [1] 11 58 0 37 25 34
```

```
### The results suggest that clusters 1 and 3 exhibit a high degree of stability. While clusters 4 and 6 are unstable.
```

```
# Assign cluster labels using cutree() based on hierarchical clustering (hc_ward_clust) with k=6
groups <- cutree(hc_ward_clust, k = 6)
# Define a function to print details of each cluster
```

```

print_clusters <- function(labels, k) {
  for(i in 1:k) {
    print(paste("cluster", i))
    print(Cereals_1[labels==i,c("mfr","calories","protein","fat","sodium","fiber","carbo","sugars","potass"
                                "vitamins","rating")])
  }
}

# Call the function to print cluster details for the assigned labels
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				

### I opted to select clusters based on statistical values and nutritional richness with the goal of fo

### To determine whether normalization was necessary,I concluded that it was not. Normalizing the data

### The cereal diet levels within the clusters exhibit variations in richness, adequacy, and deficienci

### While Cluster 1 provides nutritionally consistent guidelines for crafting a balanced diet, the opti

### Clusters 4 and 5 showcase well-balanced nutritional values and receive high ratings for consumer sa