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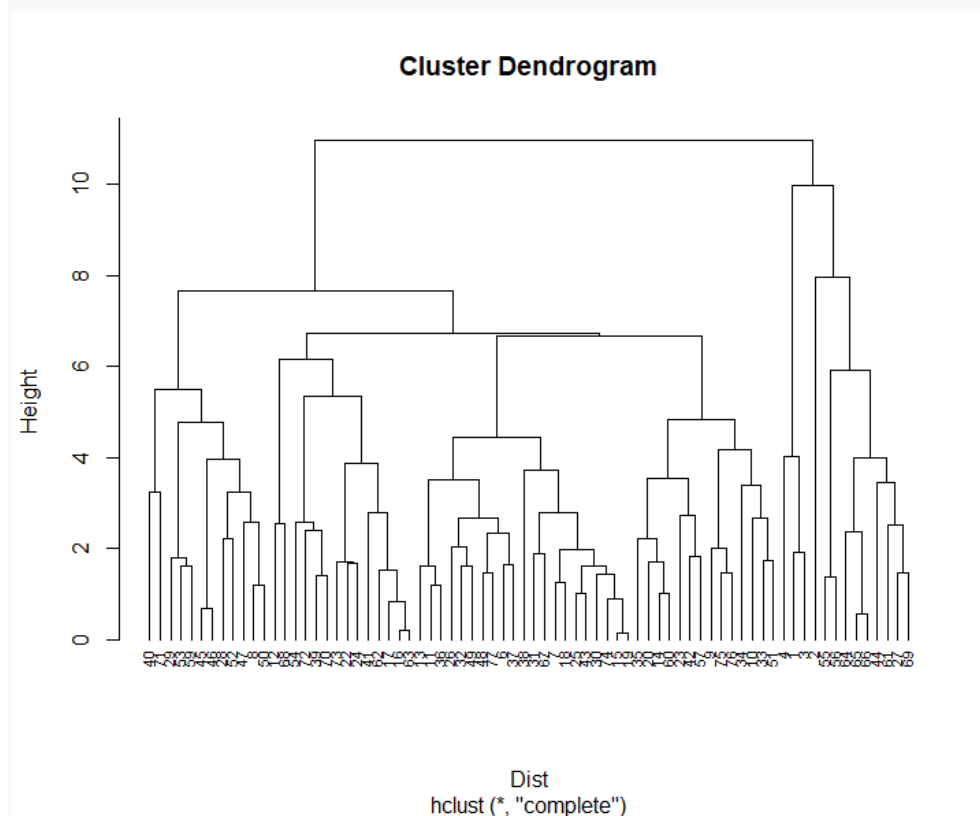
## Assignment 5

```
> library(cluster)
> library(caret)
> library(dendextend)
> library(knitr)
> library(factoextra)
> library(readr)

> Cereals <- read.csv("C:/Users/choll/Downloads/Cereals.csv")
> NumData <- data.frame(Cereals[,4:16])
> NumData <- na.omit(NumData)
> CerealsNormalize <- scale(NumData)
```

### Task 1

```
> Dist <- dist(CerealsNormalize, method = "euclidean")
> HClust <- hclust(Dist, method = "complete")
> plot(HClust, cex = 0.7, hang = -1)
```



```
> SingleHClust <- agnes(CerealsNormalize, method = "single")
> CompleteHClust <- agnes(CerealsNormalize, method = "complete")
> AverageHClust <- agnes(CerealsNormalize, method = "average")
> wardHClust <- agnes(CerealsNormalize, method = "ward")

> print(SingleHClust$ac)
[1] 0.6067859

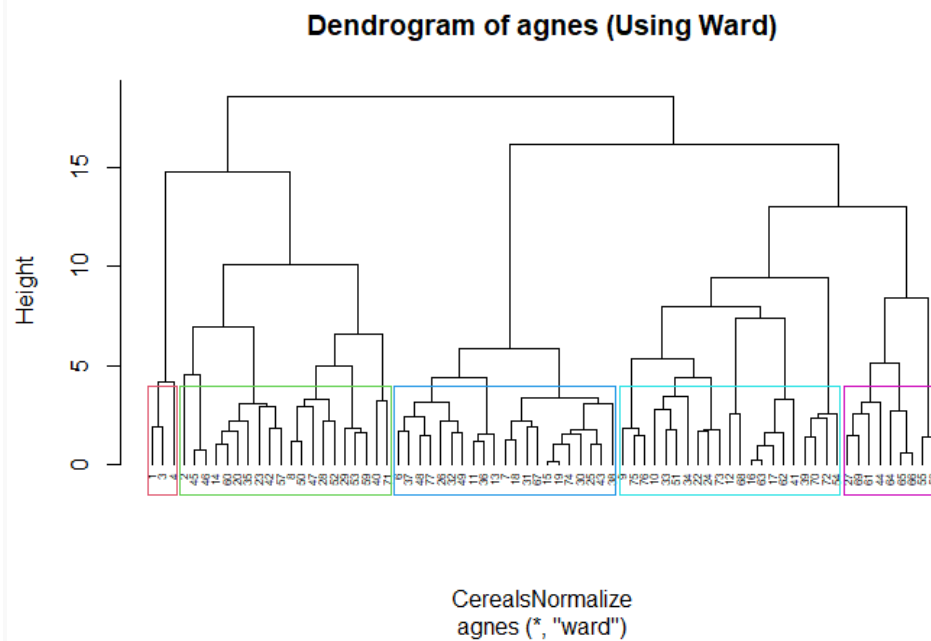
> print(CompleteHClust$ac)
[1] 0.8353712
```

```
> print(AverageHClust$ac)
[1] 0.7766075
> print(wardHClust$ac)
[1] 0.9046042
```

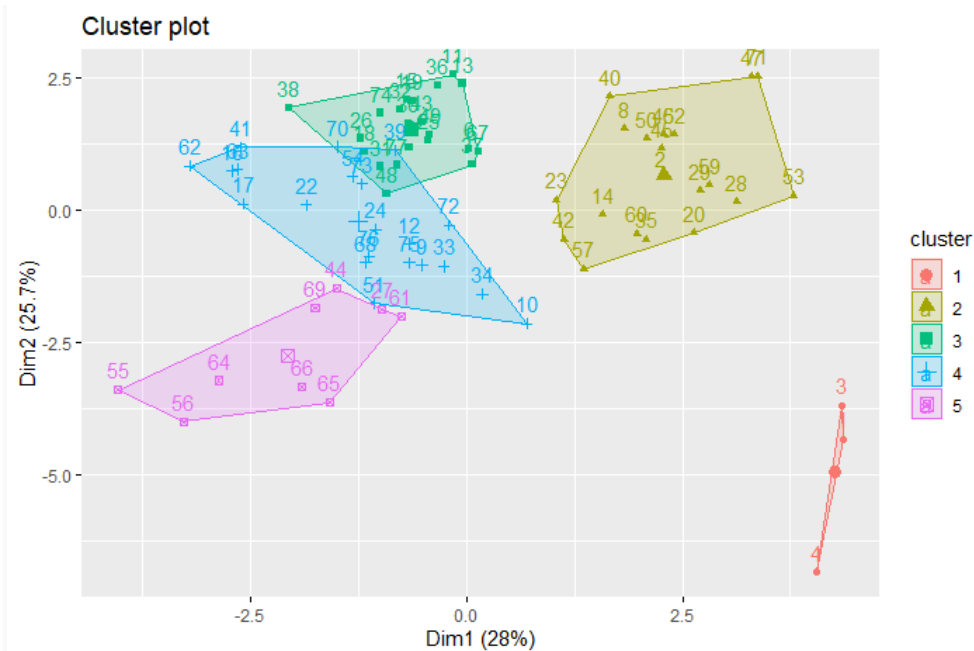
**##ward is the best option since its value is highest.**

### Task 2

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



```
> SGroup <- cutree(wardHClust, k=5)
> DFrame2 <- as.data.frame(cbind(CerealsNormalize,SGroup))
> fviz_cluster(list(data = DFrame2, cluster = SGroup))
```



##Five clusters total

### Task 3

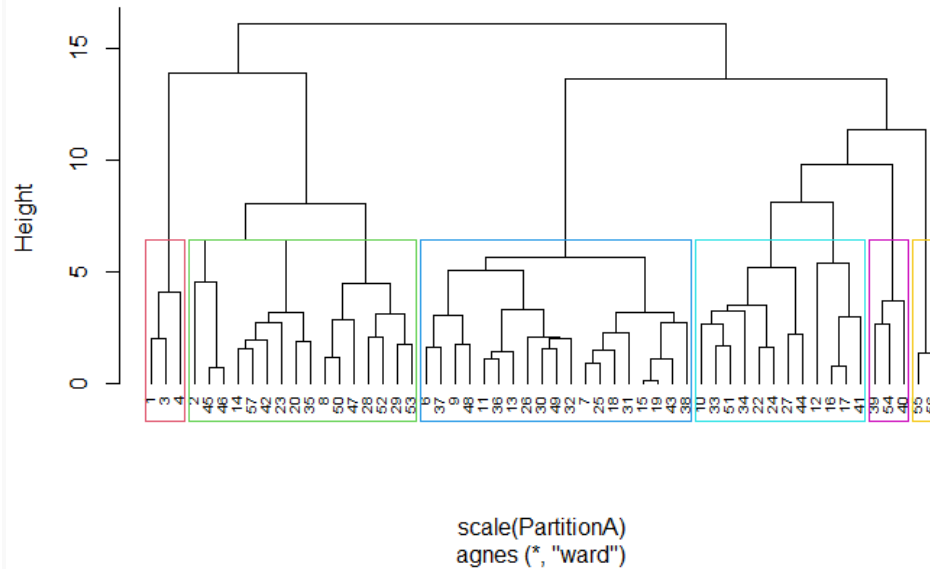
```
> set.seed(123)
> PartitionA <- NumData[1:55,]
> PartitionB <- NumData[56:74,]

> SingleSL <- agnes(scale(PartitionA), method = "single")
> CompleteSL <- agnes(scale(PartitionA), method = "complete")
> AverageSL <- agnes(scale(PartitionA), method = "average")
> WardSL <- agnes(scale(PartitionA), method = "ward")
> cbind(single=SingleSL$ac , complete=CompleteSL$ac , average= AverageSL$ac ,
ward= WardSL$ac)

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

> pltree(WardSL, cex = 0.6, hang = -1, main = "Dendrogram of Agnes with Partit
ioned Data (Using ward)")
> rect.hclust(WardSL, k = 6, border = 2:7)
```

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



```
> Cut2 <- cutree(WardSL, k = 6)
```

```
> ResultSL <- as.data.frame(cbind(PartitionA, Cut2))
```

```
> ResultSL[ResultSL$Cut2==1,]
```

	calories	protein	fat	sodium	fiber	carbo	sugars	potass	vitamins	shelf	weight	cups	rating	Cut2
1	70	4	1	130	10	5	6	280	25	3	1	0.33	68.40297	1
3	70	4	1	260	9	7	5	320	25	3	1	0.33	59.42551	1
4	50	4	0	140	14	8	0	330	25	3	1	0.50	93.70491	1

```
> CentroidOne <- colMeans(ResultSL[ResultSL$Cut2==1,])
```

```
> ResultSL[ResultSL$Cut2==2,]
```

	calories	protein	fat	sodium	fiber	carbo	sugars	potass	vitamins	shelf	weight	cups	rating	Cut2
2	120	3	5	15	2.0	8.0	8	135	0	3	1.00	1.00	33.98368	2
8	130	3	2	210	2.0	18.0	8	100	25	3	1.33	0.75	37.03856	2
14	110	3	2	140	2.0	13.0	7	105	25	3	1.00	0.50	40.40021	2
20	110	3	3	140	4.0	10.0	7	160	25	3	1.00	0.50	40.44877	2
23	100	2	1	140	2.0	11.0	10	120	25	3	1.00	0.75	36.17620	2
28	120	3	2	160	5.0	12.0	10	200	25	3	1.25	0.67	40.91705	2
29	120	3	0	240	5.0	14.0	12	190	25	3	1.33	0.67	41.01549	2
35	120	3	3	75	3.0	13.0	4	100	25	3	1.00	0.33	45.81172	2
42	100	4	2	150	2.0	12.0	6	95	25	2	1.00	0.67	45.32807	2
45	150	4	3	95	3.0	16.0	11	170	25	3	1.00	1.00	37.13686	2
46	150	4	3	150	3.0	16.0	11	170	25	3	1.00	1.00	34.13976	2
47	160	3	2	150	3.0	17.0	13	160	25	3	1.50	0.67	30.31335	2
50	140	3	2	220	3.0	21.0	7	130	25	3	1.33	0.67	40.69232	2
52	130	3	2	170	1.5	13.5	10	120	25	3	1.25	0.50	30.45084	2
53	120	3	1	200	6.0	11.0	14	260	25	3	1.33	0.67	37.84059	2
57	100	4	1	135	2.0	14.0	6	110	25	3	1.00	0.50	49.51187	2

```
> CentroidTwo <- colMeans(ResultSL[ResultSL$Cut2==2,])
```

```
> ResultSL[ResultSL$Cut2==3,]
```

	calories	protein	fat	sodium	fiber	carbo	sugars	potass	vitamins	shelf	weight	cups	rating	Cut2
6	110	2	2	180	1.5	10.5	10	70	25	1	1	0.75	29.50954	3

7	110	2	0	125	1.0	11.0	14	30	25	2	1	1.00	33.17409	3
9	90	2	1	200	4.0	15.0	6	125	25	1	1	0.67	49.12025	3
11	120	1	2	220	0.0	12.0	12	35	25	2	1	0.75	18.04285	3
13	120	1	3	210	0.0	13.0	9	45	25	2	1	0.75	19.82357	3
15	110	1	1	180	0.0	12.0	13	55	25	2	1	1.00	22.73645	3
18	110	1	0	90	1.0	13.0	12	20	25	2	1	1.00	35.78279	3
19	110	1	1	180	0.0	12.0	13	65	25	2	1	1.00	22.39651	3
25	110	2	1	125	1.0	11.0	13	30	25	2	1	1.00	32.20758	3
26	110	1	0	200	1.0	14.0	11	25	25	1	1	0.75	31.43597	3
30	110	1	1	135	0.0	13.0	12	25	25	2	1	0.75	28.02576	3
31	100	2	0	45	0.0	11.0	15	40	25	1	1	0.88	35.25244	3
32	110	1	1	280	0.0	15.0	9	45	25	2	1	0.75	23.80404	3
36	120	1	2	220	1.0	12.0	11	45	25	2	1	1.00	21.87129	3
37	110	3	1	250	1.5	11.5	10	90	25	1	1	0.75	31.07222	3
38	110	1	0	180	0.0	14.0	11	35	25	1	1	1.33	28.74241	3
43	110	2	1	180	0.0	12.0	12	55	25	2	1	1.00	26.73451	3
48	100	2	1	220	2.0	15.0	6	90	25	1	1	1.00	40.10596	3
49	120	2	1	190	0.0	15.0	9	40	25	2	1	0.67	29.92429	3

```
> CentroidThree <- colMeans(ResultSL[ResultSL$Cut2==3,])
> ResultSL[ResultSL$Cut2==4,]
  calories protein fat sodium fiber carbo sugars potass vitamins shelf weight cups rating Cut2
10      90       3  0   210     5    13      5    190     25      3      1  0.67  53.31381    4
12     110       6  2   290     2    17      1    105     25      1      1  1.25  50.76500    4
16     110       2  0   280     0    22      3     25     25      1      1  1.00  41.44502    4
17     100       2  0   290     1    21      2     35     25      1      1  1.00  45.86332    4
22     110       2  0   220     1    21      3     30     25      3      1  1.00  46.89564    4
24     100       2  0   190     1    18      5     80     25      3      1  0.75  44.33086    4
27     100       3  0     0     3    14      7    100     25      2      1  0.80  58.34514    4
33     100       3  1   140     3    15      5     85     25      3      1  0.88  52.07690    4
34     110       3  0   170     3    17      3     90     25      3      1  0.25  53.37101    4
41     110       2  1   260     0    21      3     40     25      2      1  1.50  39.24111    4
44     100       4  1     0     0    16      3     95     25      2      1  1.00  54.85092    4
51      90       3  0   170     3    18      2     90     25      3      1  1.00  59.64284    4
```

```
> CentroidFour <- colMeans(ResultSL[ResultSL$Cut2==4,])
> centroids <- rbind(CentroidOne, CentroidTwo, CentroidThree, CentroidFour)
> x2 <- as.data.frame(rbind(centroids[,-14], PartitionB))
```

```
> Dist1 <- get_dist(x2)
> Matrix1 <- as.matrix(Dist1)
> dataframe1 <- data.frame(data=seq(1,nrow(PartitionB),1), Clusters =rep(0,nr
ow(PartitionB)))
> for(i in 1:nrow(PartitionB))
+ {dataframe1[i,2] <- which.min(Matrix1[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(DFrame2$SGroup[56:74], dataframe1$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(DFrame2$SGroup[56:74] == dataframe1$Clusters)
```

```
FALSE TRUE
9      10
```

**##The model would be considered partially stable given that there are 9 False values and 10 true values.**

#### **Task 4**

```
> HealthyCereals <- Cereals
> HealthyCerealsRD <- na.omit(HealthyCereals)
> clust <- cbind(HealthyCerealsRD, SGroup)
> clust[clust$SGroup==1,]
```

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

```
> clust[clust$SGroup==2,]
```

	m	fiber	carbo	sugars	potass	vitamins	shelf	weight	cups	rating	SGroup	name	mfr	type	calories	protein	fat	sodium
2							100%_Natural_Bran	Q	C	120	3	5	1					
5	2.0	8.0		8	135	0	3	1.00	1.00	33.98368	3	2	21					
8							Basic_4	G	C	130	3	2	21					
0	2.0	18.0		8	100	25	3	1.33	0.75	37.03856	3	2	14					
14							Clusters	G	C	110	3	2	14					
0	2.0	13.0		7	105	25	3	1.00	0.50	40.40021	3	2	14					
20							Cracklin'_Oat_Bran	K	C	110	3	3	14					
0	4.0	10.0		7	160	25	3	1.00	0.50	40.44877	2	2	14					
23							Crispy_Wheat_&_Raisins	G	C	100	2	1	14					
0	2.0	11.0		10	120	25	3	1.00	0.75	36.17620	2	2	16					
28	Fruit_&_Fibre_Dates,_Walnuts,_and_Oats							P	C	120	3	2	16					
0	5.0	12.0		10	200	25	3	1.25	0.67	40.91705	2	2	16					
29							Fruitful_Bran	K	C	120	3	0	24					
0	5.0	14.0		12	190	25	3	1.33	0.67	41.01549	2	2	15					
35							Great_Grains_Pecan	P	C	120	3	3	7					
5	3.0	13.0		4	100	25	3	1.00	0.33	45.81172	2	2	17					
40							Just_Right_Fruit_&_Nut	K	C	140	3	1	17					
0	2.0	20.0		9	95	100	3	1.30	0.75	36.47151	2	2	15					
42							Life	Q	C	100	4	2	15					
0	2.0	12.0		6	95	25	2	1.00	0.67	45.32807	2	2	15					

45				Muesli_Raisins,_Dates,_&Almonds	R	C	150		4	3	9
5	3.0	16.0	11	170	25	3	1.00	1.00	37.13686	2	
46				Muesli_Raisins,_Peaches,_&Pecans	R	C	150		4	3	15
0	3.0	16.0	11	170	25	3	1.00	1.00	34.13976	2	
47				Mueslix_Crispy_Blend	K	C	160		3	2	15
0	3.0	17.0	13	160	25	3	1.50	0.67	30.31335	2	
50				Nutri-Grain_Almond-Raisin	K	C	140		3	2	22
0	3.0	21.0	7	130	25	3	1.33	0.67	40.69232	2	
52				Oatmeal_Raisin_Crisp	G	C	130		3	2	17
0	1.5	13.5	10	120	25	3	1.25	0.50	30.45084	2	
53				Post_Nat._Raisin_Bran	P	C	120		3	1	20
0	6.0	11.0	14	260	25	3	1.33	0.67	37.84059	2	
57				Quaker_Oat_Squares	Q	C	100		4	1	13
5	2.0	14.0	6	110	25	3	1.00	0.50	49.51187	2	
59				Raisin_Bran	K	C	120		3	1	21
0	5.0	14.0	12	240	25	2	1.33	0.75	39.25920	2	
60				Raisin_Nut_Bran	G	C	100		3	2	14
0	2.5	10.5	8	140	25	3	1.00	0.50	39.70340	2	
71				Total_Raisin_Bran	G	C	140		3	1	19
0	4.0	15.0	14	230	100	3	1.50	1.00	28.59278	2	

```
> clust[clust$SGroup==3,]
```

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

```
77    wheaties_Honey_Gold    G    C    110    2    1    200    1.0    16.0
8      60      25      1      1 0.75 36.18756    3
```

```
> clust[clust$SGroup==4,]
```

bo	sugars	potass	vitamins	shelf	name	mfr	type	calories	protein	fat	sodium	fiber	car
9					Bran_Chex	R	C	90	2	1	200	4	
15	6	125		25	1	1	0.67	49.12025		4			
10					Bran_Flakes	P	C	90	3	0	210	5	
13	5	190		25	3	1	0.67	53.31381		4			
12					Cheerios	G	C	110	6	2	290	2	
17	1	105		25	1	1	1.25	50.76500		4			
16					Corn_Chex	R	C	110	2	0	280	0	
22	3	25		25	1	1	1.00	41.44502		4			
17					Corn_Flakes	K	C	100	2	0	290	1	
21	2	35		25	1	1	1.00	45.86332		4			
22					Crispix	K	C	110	2	0	220	1	
21	3	30		25	3	1	1.00	46.89564		4			
24					Double_Chex	R	C	100	2	0	190	1	
18	5	80		25	3	1	0.75	44.33086		4			
33					Grape_Nuts_Flakes	P	C	100	3	1	140	3	
15	5	85		25	3	1	0.88	52.07690		4			
34					Grape-Nuts	P	C	110	3	0	170	3	
17	3	90		25	3	1	0.25	53.37101		4			
39	Just_Right_Crunchy_Nuggets					K	C	110	2	1	170	1	
17	6	60		100	3	1	1.00	36.52368		4			
41					Kix	G	C	110	2	1	260	0	
21	3	40		25	2	1	1.50	39.24111		4			
51					Nutri-grain_Wheat	K	C	90	3	0	170	3	
18	2	90		25	3	1	1.00	59.64284		4			
54					Product_19	K	C	100	3	0	320	1	
20	3	45		100	3	1	1.00	41.50354		4			
62					Rice_Chex	R	C	110	1	0	240	0	
23	2	30		25	1	1	1.13	41.99893		4			
63					Rice_Krispies	K	C	110	2	0	290	0	
22	3	35		25	1	1	1.00	40.56016		4			
68					Special_K	K	C	110	6	0	230	1	
16	3	55		25	1	1	1.00	53.13132		4			
70					Total_Corn_Flakes	G	C	110	2	1	200	0	
21	3	35		100	3	1	1.00	38.83975		4			
72					Total_whole_Grain	G	C	100	3	1	200	3	
16	3	110		100	3	1	1.00	46.65884		4			
73					Triples	G	C	110	2	1	250	0	
21	3	60		25	3	1	0.75	39.10617		4			
75					wheat_Chex	R	C	100	3	1	230	3	
17	3	115		25	1	1	0.67	49.78744		4			
76					wheaties	G	C	100	3	1	200	3	
17	3	110		25	1	1	1.00	51.59219		4			

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

```
[1] 73.84446
```

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

```
[1] 38.26161
```

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

```
[1] 28.84825
```

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

```
[1] 46.46513
```

```
##The cluster with the highest rating is Cluster 1, so it should be chosen
```