Data Mining HW9

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1. R code: Use attached dataset and hierarchical clustering algorithm (single and complete) to find 5 or 6 clusters

```
library(cluster)
library(dendextend)
#Reading the file
a2<-read.csv(file.choose(),head = TRUE)
#Converting to a dataframe
a2_df<- as.data.frame(a2)
head(a2_df)</pre>
```

```
> a2_df<- as.data.frame(a2)</pre>
> head(a2_df)
      Company Fixed_charge RoR Cost Load_factor Demand_growth Sales Nuclear Fuel_Cost
                      1.06 9.2 151
1
      Arizona
                                            54.4
                                                          1.6 9077
                                                                        0.0
                                                                                0.628
2
      Boston
                      0.89 10.3
                                            57.9
                                                               5088
                                                                       25.3
                                                                                1.555
                                 202
                                                          2.2
                      1.43 15.4
3
     Central
                                 113
                                            53.0
                                                          3.4
                                                               9212
                                                                        0.0
                                                                                1.058
4 Commonwealth
                      1.02 11.2
                                 168
                                            56.0
                                                          0.3 6423
                                                                       34.3
                                                                                0.700
5
           NY
                      1.49 8.8
                                 192
                                            51.2
                                                          1.0 3300
                                                                       15.6
                                                                                2.044
      Florida
                      1.32 13.5 111
                                            60.0
                                                         -2.2 11127
                                                                                1.241
```

Figure 0.1: Dimensions of the data set

```
#Splitting the categorical column and numerical column
a2_col1<-a2_df[1]
head(a2_col1)
a2_df_num<-a2_df[2:9]
head(a2_df_num)</pre>
```

```
a2_col1<-a2_df[1]
  a2_df_num<-a2_df[2:9]
  head(a2_col1)
       Company
1
      Arizona
2
       Boston
      Central
4
  Commonwealth
5
      Florida
6
  head(a2_df_num)
  Fixed_charge
                 ROR Cost Load_factor Demand_growth Sales Nuclear Fuel_Cost
1
                                                                   0.0
           1.06
                 9.2
                       151
                                   54.4
                                                   1.6
                                                         9077
                                                                            0.628
2
           0.89 10.3
                       202
                                   57.9
                                                    2.2
                                                         5088
                                                                  25.3
                                                                            1.555
3
           1.43
                15.4
                       113
                                   53.0
                                                         9212
                                                                   0.0
                                                                            1.058
4
                                                                            0.700
           1.02
                11.2
                       168
                                   56.0
                                                         6423
                                                                  34.3
                                                    0.3
5
           1.49
                 8.8
                       192
                                                    1.0
                                                         3300
                                                                  15.6
                                                                            2.044
                                   51.2
6
           1.32 13.5
                       111
                                   60.0
                                                                  22.5
                                                                            1.241
```

Figure 0.2: Splitting the columns

We are using the 'hclust' function to do the hierarchical clustering, we require the distance values which can be computed in R by using the 'dist' function. Default measure for dist function is 'Euclidean'

1 SINGLE LINKAGE

```
hc_s <- hclust(dist(a2_df_num), method = "single")
plot(hc_s,labels = a2_col1[hc_s$order,1])</pre>
```

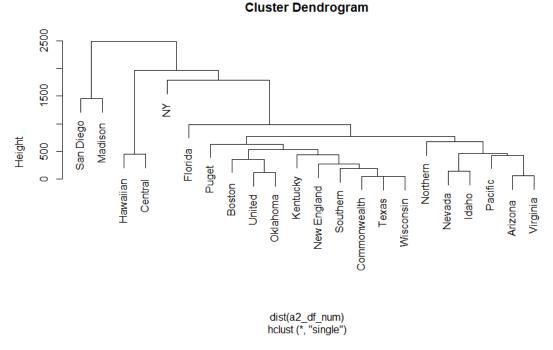
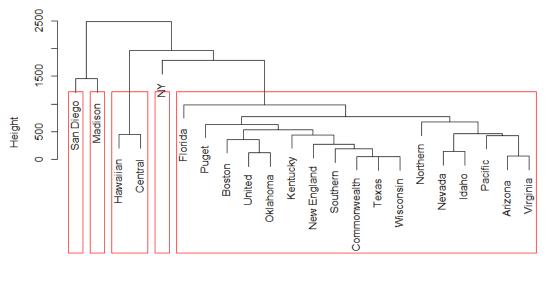


Figure 1.1: Single linkage with hclust

Finding 5 clusters from the single linkage dendrogram

Cluster Dendrogram



dist(a2_df_num) hclust (*, "single")

Figure 1.2: Highlighted clusters with red boxes

#selective coloring of branches AND labels
dend_s <- as.dendrogram(hc_s)
dend_s=color_labels(dend_s,k=5)
dend_s=color_branches(dend_s,k=5,groupLabels=TRUE)
plot(dend_s)</pre>

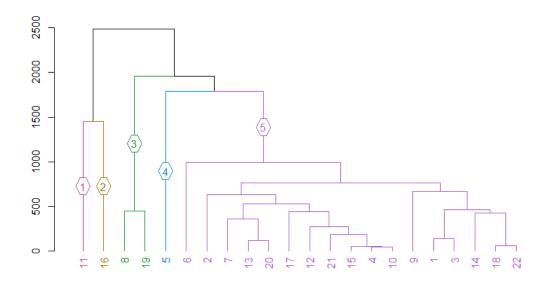


Figure 1.3: selective coloring of branches for single linkage clusters

2 COMPLETE LINKAGE

hc_c <- hclust(dist(a2_df_num), method = "complete")
plot(hc_c,labels = a2_col1[hc_c\$order,1])</pre>

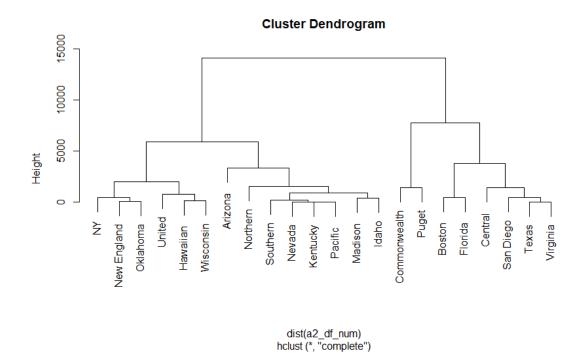


Figure 2.1: Complete linkage with hclust

```
rect.hclust(hc_c, k = 5, border = "red")
```

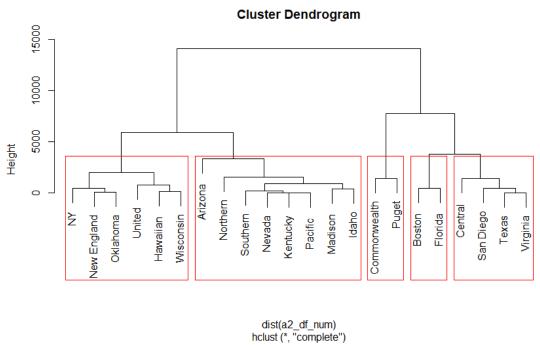


Figure 2.2: Highlighted clusters with red boxes

```
#selective coloring of branches AND labels
dend <- as.dendrogram(hc_c)</pre>
dend=color_labels(dend, k=5)
dend=color_branches(dend,k=5,groupLabels=TRUE)
plot(dend)
```

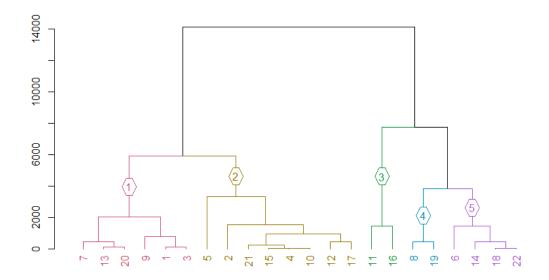


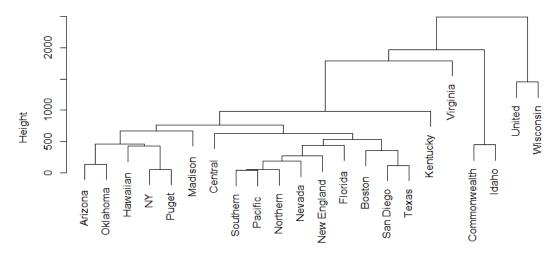
Figure 2.3: selective coloring of branches for Complete linkage clusters

Another additional method is to use the AGNES (acronym for Agglomerative Nesting) function. In this method, each observation is assigned to its own cluster. Then, the similarity (or distance) between each of the clusters is computed and the two most similar clusters are merged into one. Finally, steps 2 and 3 are repeated until there is only one cluster left

3 SINGLE LINKAGE

```
a2_agnes_single<-agnes(a2_df_num,diss=FALSE, stand=FALSE, method="single")
plot(a2_agnes_single,labels = a2_col1[a2_agnes_single$order,1])</pre>
```

Dendrogram of agnes(x = a2_df_num, diss = FALSE, stand = FALSE, method = "single")

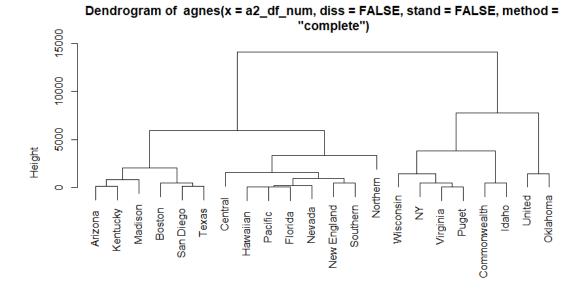


a2_df_num Agglomerative Coefficient = 0.81

Figure 3.1: Single linkage with AGNES

4 COMPLETE LINKAGE

a2_agnes_complete<-agnes(a2_df_num,diss=FALSE, stand=FALSE, method="complete") plot(a2_agnes_complete,labels = a2_col1[a2_agnes_complete\$order,1])



a2_df_num Agglomerative Coefficient = 0.96

Figure 4.1: Complete linkage with AGNES