

Assignment4_Clustering

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```
library(factoextra) # clustering algorithms & visualization
```

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

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

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(ISLR)  
library(caret)
```

```
## Loading required package: lattice
```

#Importing the dataset

```
Pharmaceuticals <- read.csv("C:\\Users\\HP\\Desktop\\Pharmaceuticals.csv")  
summary(Pharmaceuticals)
```

```
##      Symbol      Name      Market_Cap      Beta
## Length:21      Length:21      Min.   : 0.41      Min.   :0.1800
## Class :character Class :character 1st Qu.: 6.30      1st Qu.:0.3500
## Mode  :character Mode  :character Median : 48.19      Median :0.4600
##                                     Mean  : 57.65      Mean   :0.5257
##                                     3rd Qu.: 73.84      3rd Qu.:0.6500
##                                     Max.   :199.47      Max.   :1.1100
##      PE_Ratio      ROE      ROA      Asset_Turnover      Leverage
## Min.   : 3.60      Min.   : 3.9      Min.   : 1.40      Min.   :0.3      Min.   :0.0000
## 1st Qu.:18.90      1st Qu.:14.9      1st Qu.: 5.70      1st Qu.:0.6      1st Qu.:0.1600
## Median :21.50      Median :22.6      Median :11.20      Median :0.6      Median :0.3400
## Mean   :25.46      Mean   :25.8      Mean   :10.51      Mean   :0.7      Mean   :0.5857
## 3rd Qu.:27.90      3rd Qu.:31.0      3rd Qu.:15.00      3rd Qu.:0.9      3rd Qu.:0.6000
## Max.   :82.50      Max.   :62.9      Max.   :20.30      Max.   :1.1      Max.   :3.5100
##      Rev_Growth      Net_Profit_Margin      Median_Recommendation      Location
## Min.   : -3.17      Min.   : 2.6      Length:21      Length:21
## 1st Qu.: 6.38      1st Qu.:11.2      Class :character      Class :character
## Median : 9.37      Median :16.1      Mode  :character      Mode  :character
## Mean   :13.37      Mean   :15.7
## 3rd Qu.:21.87      3rd Qu.:21.1
## Max.   :34.21      Max.   :25.5
##      Exchange
## Length:21
## Class :character
## Mode  :character
##
##
##
```

a. Cluster the 21 firms using only the numerical variables (1–9). Justify the various decisions made during the cluster analysis, such as variable weights, the specific clustering algorithm(s) used, the number of clusters formed, and so on.

#Before clustering data, remove missing data and rescale variables for comparability.

```
Pharma<- na.omit(Pharmaceuticals) #gives the data after removing the missing values.
Pharma
```

##	Symbol	Name	Market_Cap	Beta	PE_Ratio	ROE	ROA
## 1	ABT	Abbott Laboratories	68.44	0.32	24.7	26.4	11.8
## 2	AGN	Allergan, Inc.	7.58	0.41	82.5	12.9	5.5
## 3	AHM	Amersham plc	6.30	0.46	20.7	14.9	7.8
## 4	AZN	AstraZeneca PLC	67.63	0.52	21.5	27.4	15.4
## 5	AVE	Aventis	47.16	0.32	20.1	21.8	7.5
## 6	BAY	Bayer AG	16.90	1.11	27.9	3.9	1.4
## 7	BMJ	Bristol-Myers Squibb Company	51.33	0.50	13.9	34.8	15.1
## 8	CHTT	Chattem, Inc	0.41	0.85	26.0	24.1	4.3
## 9	ELN	Elan Corporation, plc	0.78	1.08	3.6	15.1	5.1
## 10	LLY	Eli Lilly and Company	73.84	0.18	27.9	31.0	13.5
## 11	GSK	GlaxoSmithKline plc	122.11	0.35	18.0	62.9	20.3
## 12	IVX	IVAX Corporation	2.60	0.65	19.9	21.4	6.8
## 13	JNJ	Johnson & Johnson	173.93	0.46	28.4	28.6	16.3
## 14	MRX	Medicis Pharmaceutical Corporation	1.20	0.75	28.6	11.2	5.4
## 15	MRK	Merck & Co., Inc.	132.56	0.46	18.9	40.6	15.0
## 16	NVS	Novartis AG	96.65	0.19	21.6	17.9	11.2
## 17	PFE	Pfizer Inc	199.47	0.65	23.6	45.6	19.2
## 18	PHA	Pharmacia Corporation	56.24	0.40	56.5	13.5	5.7
## 19	SGP	Schering-Plough Corporation	34.10	0.51	18.9	22.6	13.3
## 20	WPI	Watson Pharmaceuticals, Inc.	3.26	0.24	18.4	10.2	6.8
## 21	WYE	Wyeth	48.19	0.63	13.1	54.9	13.4

##	Asset_Turnover	Leverage	Rev_Growth	Net_Profit_Margin	Median_Recommendation
## 1	0.7	0.42	7.54	16.1	Moderate Buy
## 2	0.9	0.60	9.16	5.5	Moderate Buy
## 3	0.9	0.27	7.05	11.2	Strong Buy
## 4	0.9	0.00	15.00	18.0	Moderate Sell
## 5	0.6	0.34	26.81	12.9	Moderate Buy
## 6	0.6	0.00	-3.17	2.6	Hold
## 7	0.9	0.57	2.70	20.6	Moderate Sell
## 8	0.6	3.51	6.38	7.5	Moderate Buy
## 9	0.3	1.07	34.21	13.3	Moderate Sell
## 10	0.6	0.53	6.21	23.4	Hold
## 11	1.0	0.34	21.87	21.1	Hold
## 12	0.6	1.45	13.99	11.0	Hold
## 13	0.9	0.10	9.37	17.9	Moderate Buy
## 14	0.3	0.93	30.37	21.3	Moderate Buy
## 15	1.1	0.28	17.35	14.1	Hold
## 16	0.5	0.06	-2.69	22.4	Hold
## 17	0.8	0.16	25.54	25.2	Moderate Buy
## 18	0.6	0.35	15.00	7.3	Hold
## 19	0.8	0.00	8.56	17.6	Hold
## 20	0.5	0.20	29.18	15.1	Moderate Sell
## 21	0.6	1.12	0.36	25.5	Hold

##	Location	Exchange
## 1	US	NYSE
## 2	CANADA	NYSE
## 3	UK	NYSE
## 4	UK	NYSE
## 5	FRANCE	NYSE
## 6	GERMANY	NYSE
## 7	US	NYSE
## 8	US	NASDAQ
## 9	IRELAND	NYSE
## 10	US	NYSE

```
## 11      UK      NYSE
## 12      US      AMEX
## 13      US      NYSE
## 14      US      NYSE
## 15      US      NYSE
## 16 SWITZERLAND NYSE
## 17      US      NYSE
## 18      US      NYSE
## 19      US      NYSE
## 20      US      NYSE
## 21      US      NYSE
```

#Only the quantitative variables (1-9) are required to cluster the 21 firms.

```
row.names(Pharma)<- Pharma[,1]
Pharma_1<- Pharma[,3:11]
head(Pharma_1)
```

```
##      Market_Cap Beta PE_Ratio ROE ROA Asset_Turnover Leverage Rev_Growth
## ABT      68.44 0.32    24.7 26.4 11.8          0.7      0.42      7.54
## AGN      7.58 0.41    82.5 12.9 5.5          0.9      0.60      9.16
## AHM      6.30 0.46    20.7 14.9 7.8          0.9      0.27      7.05
## AZN     67.63 0.52    21.5 27.4 15.4          0.9      0.00     15.00
## AVE     47.16 0.32    20.1 21.8 7.5          0.6      0.34     26.81
## BAY     16.90 1.11    27.9 3.9 1.4          0.6      0.00     -3.17
##      Net_Profit_Margin
## ABT              16.1
## AGN              5.5
## AHM             11.2
## AZN             18.0
## AVE             12.9
## BAY              2.6
```

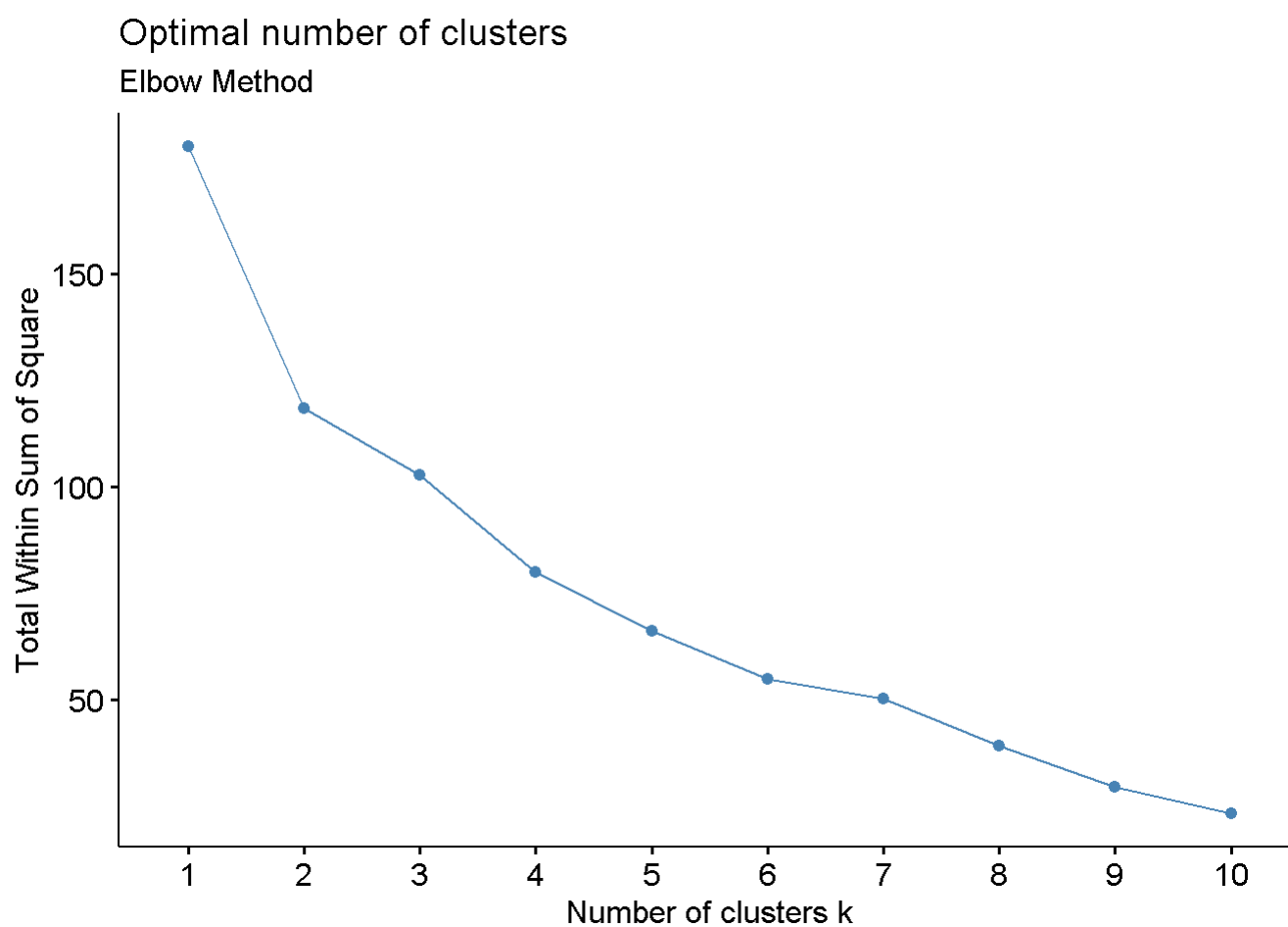
#Scale all the dataframe's quantitative variables

```
Pharma_2<-scale(Pharma_1)
head(Pharma_2)
```

```
##      Market_Cap      Beta      PE_Ratio      ROE      ROA Asset_Turnover
## ABT  0.1840960 -0.80125356 -0.04671323  0.04009035  0.2416121  0.0000000
## AGN -0.8544181 -0.45070513  3.49706911 -0.85483986 -0.9422871  0.9225312
## AHM -0.8762600 -0.25595600 -0.29195768 -0.72225761 -0.5100700  0.9225312
## AZN  0.1702742 -0.02225704 -0.24290879  0.10638147  0.9181259  0.9225312
## AVE -0.1790256 -0.80125356 -0.32874435 -0.26484883 -0.5664461 -0.4612656
## BAY -0.6953818  2.27578267  0.14948233 -1.45146000 -1.7127612 -0.4612656
##      Leverage Rev_Growth Net_Profit_Margin
## ABT -0.2120979 -0.5277675      0.06168225
## AGN  0.0182843 -0.3811391     -1.55366706
## AHM -0.4040831 -0.5721181     -0.68503583
## AZN -0.7496565  0.1474473      0.35122600
## AVE -0.3144900  1.2163867     -0.42597037
## BAY -0.7496565 -1.4971443     -1.99560225
```

##Using the Elbow Method, determine the number of clusters to be used in the cluster analysis.

```
fviz_nbclust(Pharma_2, kmeans, method = "wss") + labs(subtitle = "Elbow Method")
```

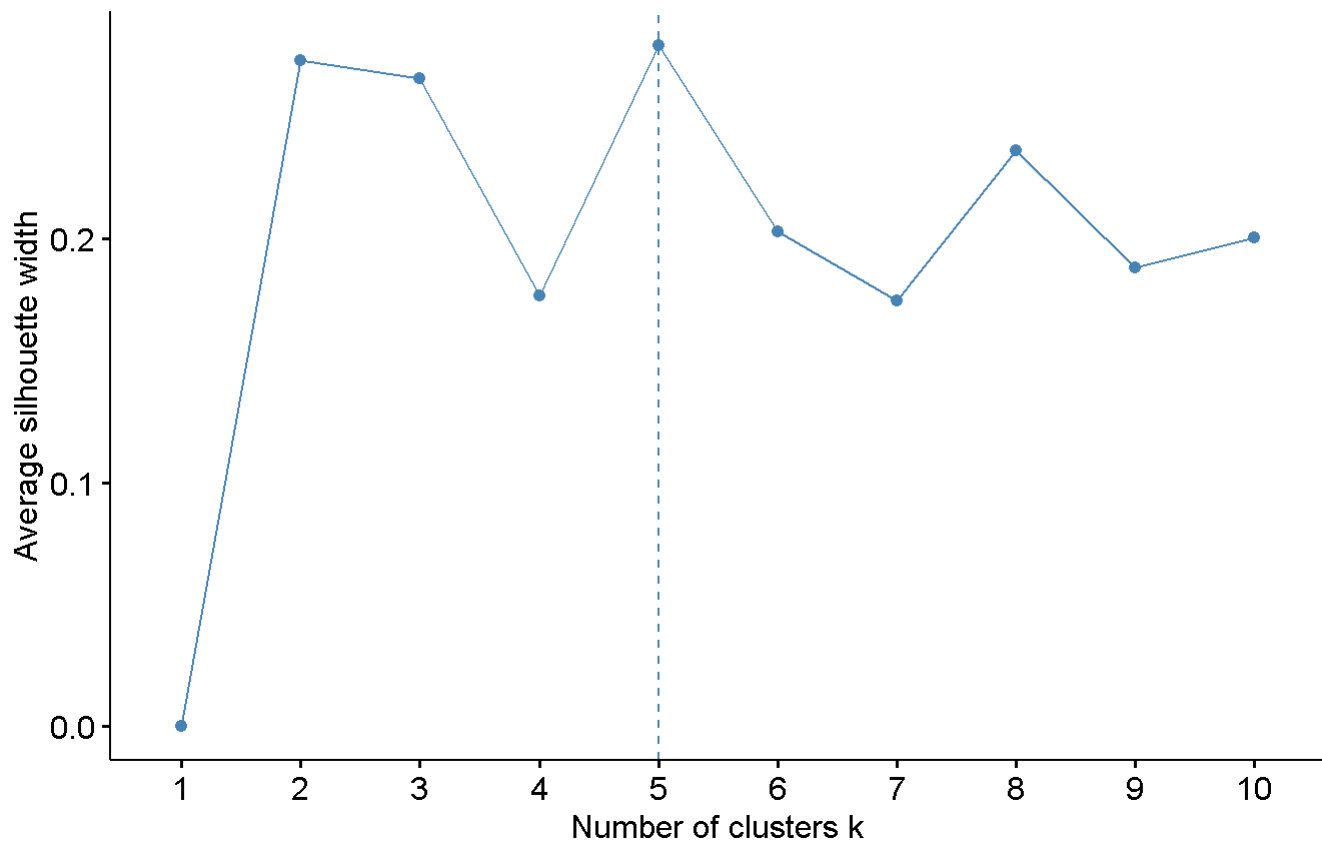


##Using the Silhouette method to determine the number of clusters

```
fviz_nbclust(Pharma_2, kmeans, method = "silhouette")+ labs(subtitle = "Silhouette Method")
```

Optimal number of clusters

Silhouette Method



The number of clusters is 5 in the above plots, which is sufficient to display the data variations.

```
set.seed(64060)
k5<- kmeans(Pharma_2,centers=5,nstart = 25)
```

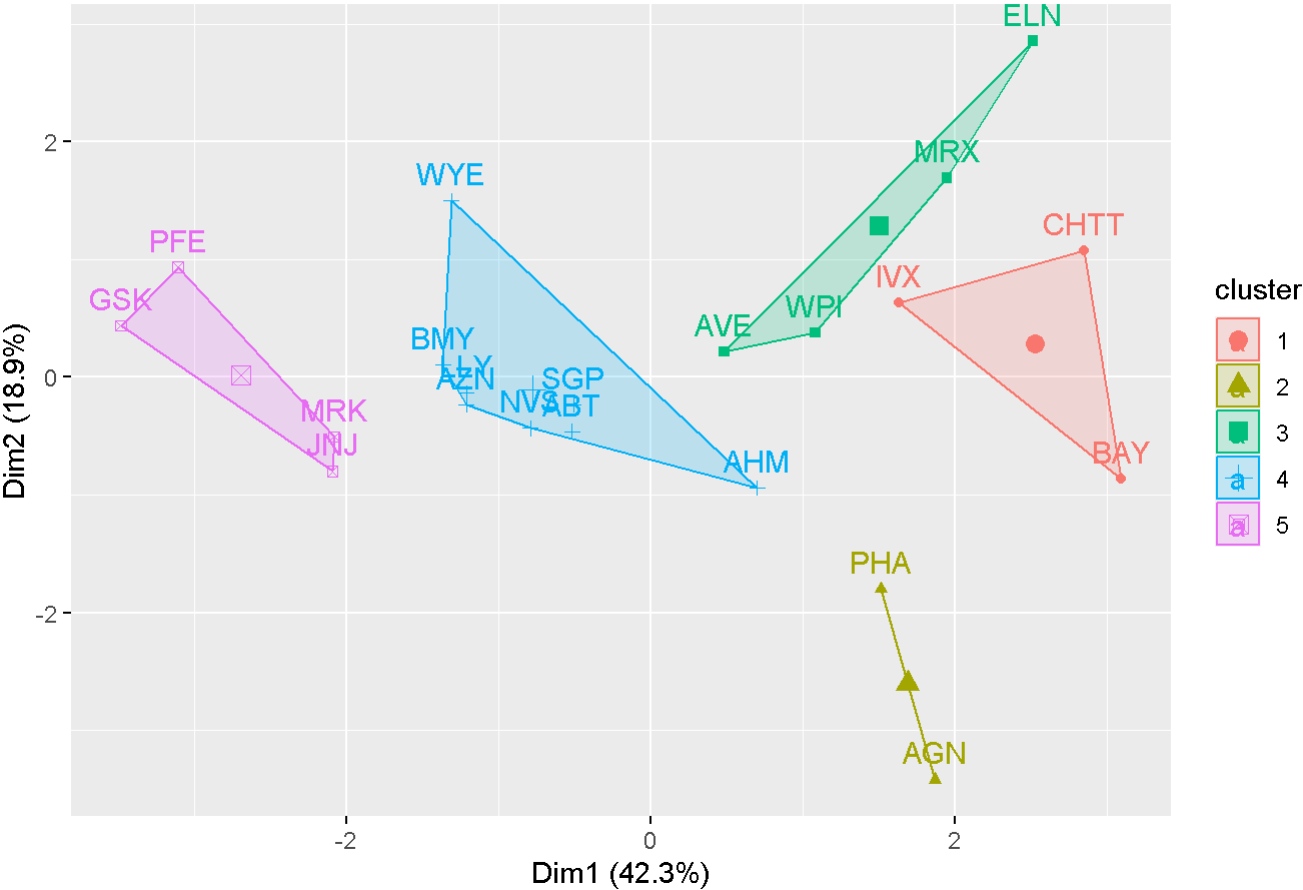
#Visualizing the output

```
k5$centers #for centroids
```

```
##      Market_Cap      Beta      PE_Ratio      ROE      ROA Asset_Turnover
## 1 -0.87051511  1.3409869 -0.05284434 -0.6184015 -1.1928478  -0.4612656
## 2 -0.43925134 -0.4701800  2.70002464 -0.8349525 -0.9234951   0.2306328
## 3 -0.76022489  0.2796041 -0.47742380 -0.7438022 -0.8107428  -1.2684804
## 4 -0.03142211 -0.4360989 -0.31724852  0.1950459  0.4083915   0.1729746
## 5  1.69558112 -0.1780563 -0.19845823  1.2349879  1.3503431   1.1531640
##      Leverage Rev_Growth Net_Profit_Margin
## 1  1.36644699 -0.6912914    -1.320000179
## 2 -0.14170336 -0.1168459    -1.416514761
## 3  0.06308085  1.5180158    -0.006893899
## 4 -0.27449312 -0.7041516     0.556954446
## 5 -0.46807818  0.4671788     0.591242521
```

```
fviz_cluster(k5,data = Pharma_2) # to Visualize the clusters
```

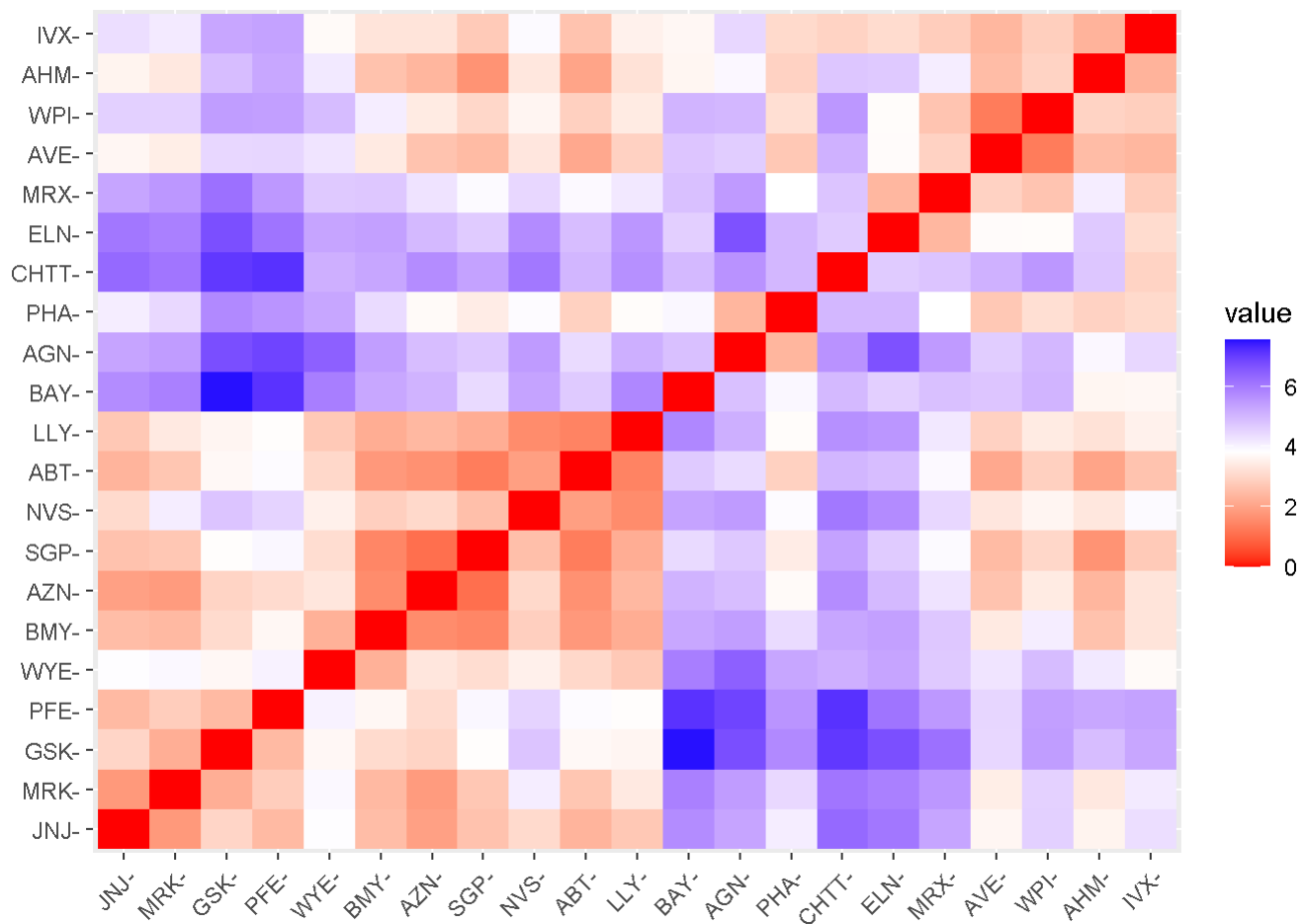
Cluster plot



k5

```
## K-means clustering with 5 clusters of sizes 3, 2, 4, 8, 4
##
## Cluster means:
##      Market_Cap      Beta      PE_Ratio      ROE      ROA Asset_Turnover
## 1 -0.87051511  1.3409869 -0.05284434 -0.6184015 -1.1928478  -0.4612656
## 2 -0.43925134 -0.4701800  2.70002464 -0.8349525 -0.9234951   0.2306328
## 3 -0.76022489  0.2796041 -0.47742380 -0.7438022 -0.8107428  -1.2684804
## 4 -0.03142211 -0.4360989 -0.31724852  0.1950459  0.4083915   0.1729746
## 5  1.69558112 -0.1780563 -0.19845823  1.2349879  1.3503431   1.1531640
##      Leverage Rev_Growth Net_Profit_Margin
## 1  1.36644699 -0.6912914    -1.320000179
## 2 -0.14170336 -0.1168459    -1.416514761
## 3  0.06308085  1.5180158     -0.006893899
## 4 -0.27449312 -0.7041516     0.556954446
## 5 -0.46807818  0.4671788     0.591242521
##
## Clustering vector:
##  ABT  AGN  AHM  AZN  AVE  BAY  BMY  CHTT  ELN  LLY  GSK  IVX  JNJ  MRX  MRK  NVS
##   4   2   4   4   3   1   4   1   3   4   5   1   5   3   5   4
##  PFE  PHA  SGP  WPI  WYE
##   5   2   4   3   4
##
## Within cluster sum of squares by cluster:
## [1] 15.595925  2.803505 12.791257 21.879320  9.284424
## (between_SS / total_SS =  65.4 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
```

```
distance<- dist(Pharma_2, method = "euclidean")
fviz_dist(distance)
```

#K-Means Cluster Analysis was used to fit the data into 5 clusters.

```
fit<-kmeans(Pharma_2,5)
```

#calculating the mean of all quantitative variables in each cluster

```
aggregate(Pharma_2,by=list(fit$cluster),FUN=mean)
```

```
##   Group.1  Market_Cap      Beta  PE_Ratio      ROE      ROA
## 1      1  1.69558112 -0.1780563 -0.1984582  1.2349879  1.3503431
## 2      2 -0.66114002 -0.7233539 -0.3512251 -0.6736441 -0.5915022
## 3      3 -0.96247577  1.1949250 -0.3639982 -0.5200697 -0.9610792
## 4      4 -0.52462814  0.4451409  1.8498439 -1.0404550 -1.1865838
## 5      5  0.08926902 -0.4618336 -0.3208615  0.3260892  0.5396003
##   Asset_Turnover  Leverage Rev_Growth Net_Profit_Margin
## 1  1.153164e+00 -0.4680782  0.4671788      0.5912425
## 2 -1.537552e-01 -0.4040831  0.6917224     -0.4005718
## 3 -1.153164e+00  1.4773718  0.7120120     -0.3688236
## 4  1.480297e-16 -0.3443544 -0.5769454     -1.6095439
## 5  6.589509e-02 -0.2559803 -0.7230135      0.7343816
```

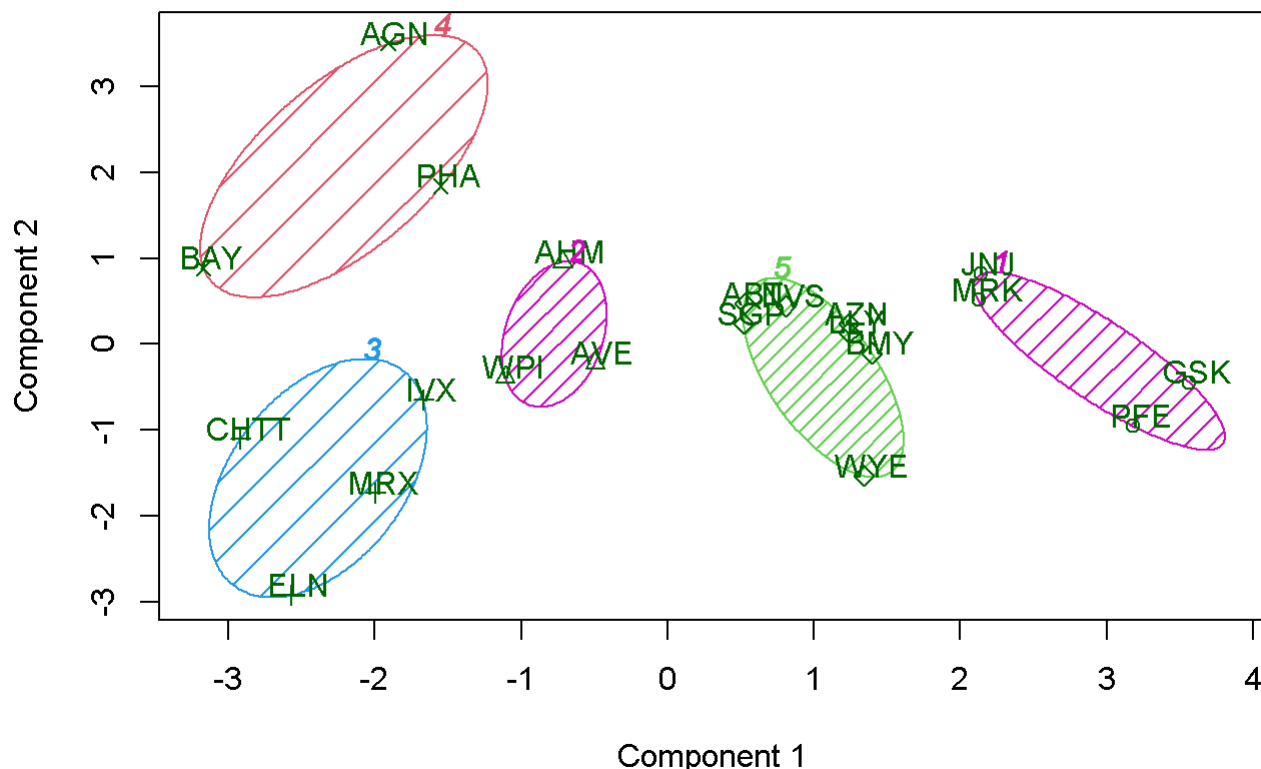
```
Pharma_3<-data.frame(Pharma_2,fit$cluster)
Pharma_3
```

##	Market_Cap	Beta	PE_Ratio	ROE	ROA	Asset_Turnover
## ABT	0.1840960	-0.80125356	-0.04671323	0.04009035	0.2416121	0.0000000
## AGN	-0.8544181	-0.45070513	3.49706911	-0.85483986	-0.9422871	0.9225312
## AHM	-0.8762600	-0.25595600	-0.29195768	-0.72225761	-0.5100700	0.9225312
## AZN	0.1702742	-0.02225704	-0.24290879	0.10638147	0.9181259	0.9225312
## AVE	-0.1790256	-0.80125356	-0.32874435	-0.26484883	-0.5664461	-0.4612656
## BAY	-0.6953818	2.27578267	0.14948233	-1.45146000	-1.7127612	-0.4612656
## BMY	-0.1078688	-0.10015669	-0.70887325	0.59693581	0.8617498	0.9225312
## CHTT	-0.9767669	1.26308721	0.03299122	-0.11237924	-1.1677918	-0.4612656
## ELN	-0.9704532	2.15893320	-1.34037772	-0.70899938	-1.0174553	-1.8450624
## LLY	0.2762415	-1.34655112	0.14948233	0.34502953	0.5610770	-0.4612656
## GSK	1.0999201	-0.68440408	-0.45749769	2.45971647	1.8389364	1.3837968
## IVX	-0.9393967	0.48409069	-0.34100657	-0.29136529	-0.6979905	-0.4612656
## JNJ	1.9841758	-0.25595600	0.18013789	0.18593083	1.0872544	0.9225312
## MRX	-0.9632863	0.87358895	0.19240011	-0.96753478	-0.9610792	-1.8450624
## MRK	1.2782387	-0.25595600	-0.40231769	0.98142435	0.8429577	1.8450624
## NVS	0.6654710	-1.30760129	-0.23677768	-0.52338423	0.1288598	-0.9225312
## PFE	2.4199899	0.48409069	-0.11415545	1.31287998	1.6322239	0.4612656
## PHA	-0.0240846	-0.48965495	1.90298017	-0.81506519	-0.9047030	-0.4612656
## SGP	-0.4018812	-0.06120687	-0.40231769	-0.21181593	0.5234929	0.4612656
## WPI	-0.9281345	-1.11285216	-0.43297324	-1.03382590	-0.6979905	-0.9225312
## WYE	-0.1614497	0.40619104	-0.75792214	1.92938746	0.5422849	-0.4612656
##	Leverage	Rev_Growth	Net_Profit_Margin	fit.cluster		
## ABT	-0.21209793	-0.52776752	0.06168225	5		
## AGN	0.01828430	-0.38113909	-1.55366706	4		
## AHM	-0.40408312	-0.57211809	-0.68503583	2		
## AZN	-0.74965647	0.14744734	0.35122600	5		
## AVE	-0.31449003	1.21638667	-0.42597037	2		
## BAY	-0.74965647	-1.49714434	-1.99560225	4		
## BMY	-0.02011273	-0.96584257	0.74744375	5		
## CHTT	3.74279705	-0.63276071	-1.24888417	3		
## ELN	0.61983791	1.88617085	-0.36501379	3		
## LLY	-0.07130879	-0.64814764	1.17413980	5		
## GSK	-0.31449003	0.76926048	0.82363947	1		
## IVX	1.10620040	0.05603085	-0.71551412	3		
## JNJ	-0.62166634	-0.36213170	0.33598685	1		
## MRX	0.44065173	1.53860717	0.85411776	3		
## MRK	-0.39128411	0.36014907	-0.24310064	1		
## NVS	-0.67286239	-1.45369888	1.02174835	5		
## PFE	-0.54487226	1.10143723	1.44844440	1		
## PHA	-0.30169102	0.14744734	-1.27936246	4		
## SGP	-0.74965647	-0.43544591	0.29026942	5		
## WPI	-0.49367621	1.43089863	-0.09070919	2		
## WYE	0.68383297	-1.17763919	1.49416183	5		

#view of the cluster plot

```
library(cluster)
clusplot(Pharma_2,fit$cluster,color = TRUE,shade = TRUE,labels = 2,lines = 0)
```

CLUSPLOT(Pharma_2)



These two components explain 61.23 % of the point variability.

#b. Interpret the clusters in relation to the numerical variables that were used to form the clusters. By examining the mean values of all quantitative variables within each cluster.

Cluster 1 consists of JNJ, MRK, PFE, and GSK. Cluster 1 has the highest Market_cap, ROA, ROE, Asset_Turnover, and the lowest Beta, PE_Ratio.

Cluster 2 has the highest Rev_Growth and the lowest PE_Ratio, Asset_Turnover.

Cluster 3 has the highest Beta, Leverage, and the lowest Market_Cap, ROE, ROA, Leverage, Rev_Growth, and Net_Profit_Margin.

Cluster 4 has the highest PE_Ratio and the lowest Leverage and Asset_Turnover.

Cluster 5 consists of the following stocks: AZN, ABT, NVS, BMY, WYE, SGP, LLY. Cluster 5 has the highest Net_Profit_Margin and the lowest leverage, Beta.

c.s there a pattern in the clusters with respect to the numerical variables (10 to 12)? (those not used in forming the clusters)

There is a pattern in the clusters for the Media recommendation variable.

Cluster 1 has an equal Hold and Moderate Buy recommendation due to its highest Market_Cap, highest ROE, highest ROA, and highest Asset_Turnover.

Hold is recommended for Cluster 2, which has the lowest PE_Ratio and Asset_Turnover.

Cluster-3, which has the highest Beta and Leverage, has mostly Moderate Buy Recommendation.

Hold is recommended for Cluster 4, which has the highest PE_Ratio.

Cluster 5 has the highest Net_Profit_Margin and is mostly recommended to hold.

In terms of variables, I've noticed a trend among the clusters (10 to 12).

Clusters 1–3 have a majority of Moderate Buy Recommendation.

Clusters 1,2,4,5 are recommended to be held.

d. Provide an appropriate name for each cluster using any or all of the variables in the dataset.

Cluster-1 - Buy (or) Hold Moderately.

Cluster-2 - Low PE_Ratio, Asset_Turnover, or Hold.

Cluster-3 - High Beta, Buy Cluster (or Leverage Cluster).

Cluster-4 is a high PE_Ratio (or high Hold) cluster.

Cluster-5 is a high net profit margin (or high hold) cluster.