

ASSIGNMENT 4 FML

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```
Pharmaceuticals <- read.csv("C://Users//heere//Downloads//ASSIGNMENT 4 FML//Pharmaceuticals (2).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
##
##
```

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

```
## Warning: package 'factoextra' was built under R version 4.2.2
```

```
## 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
```

#Task-a. Use only the numerical variables (1 to 9) to cluster the 21 firms. Justify the various choices

#Remove missing data and rescale variables for comparability before clustering data.

```
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
```

#To cluster the 21 firms, just the quantitative variables (1-9) need be collected.

```
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

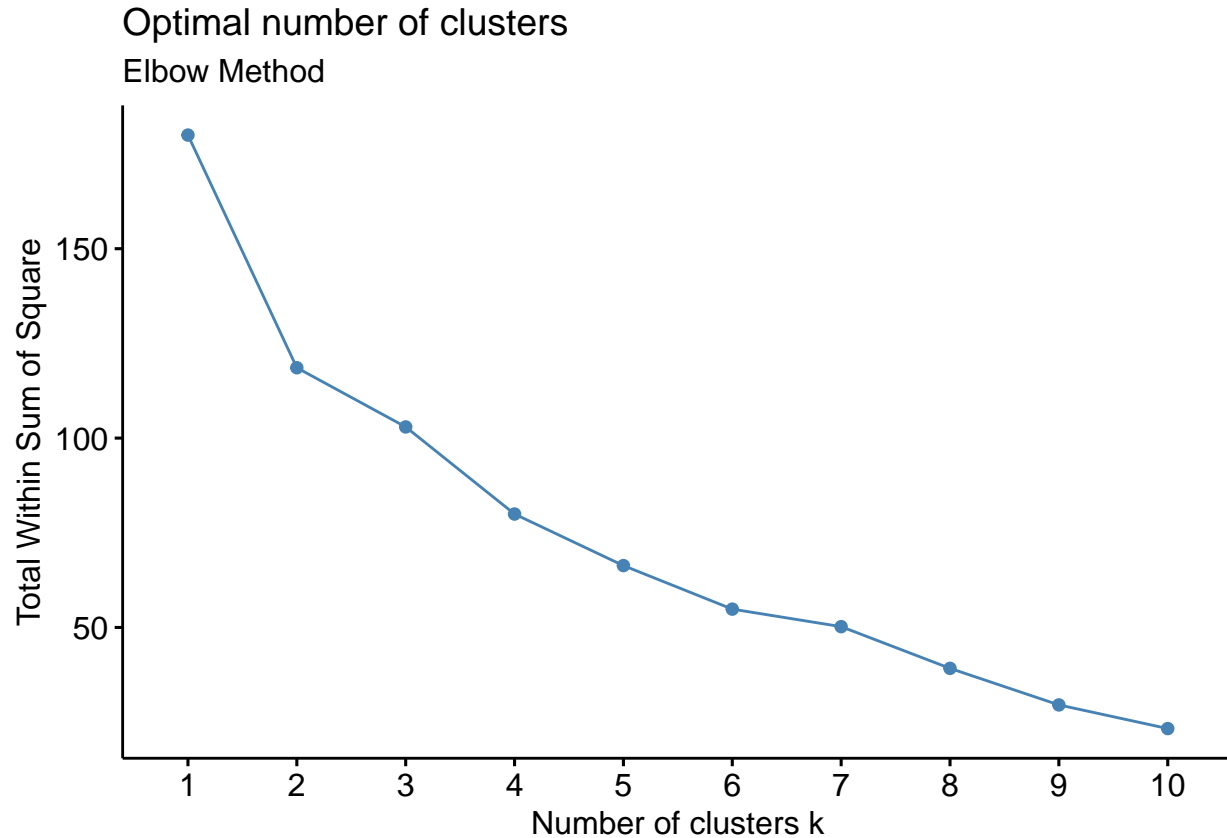
```
Pharma_2<-scale(Pharma_1)
head(Pharma_2)
```

```
##      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
```

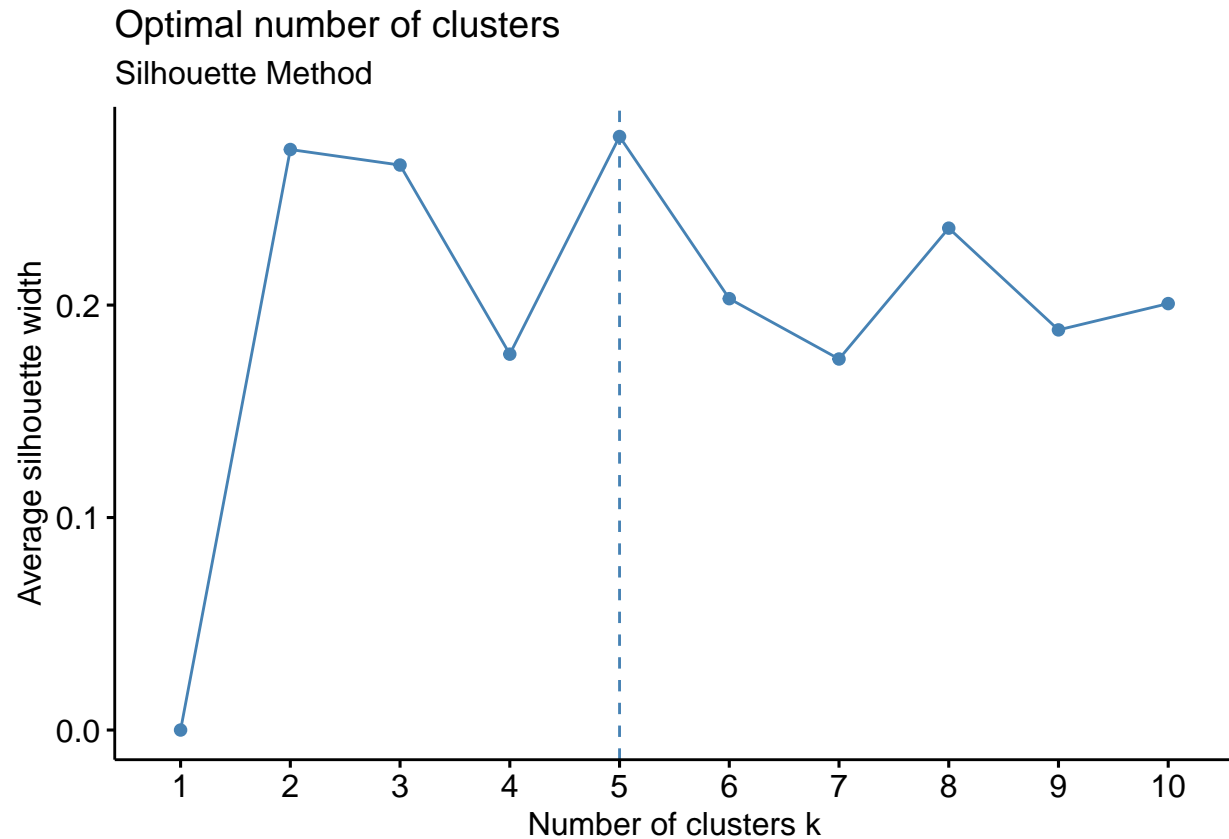
#Determining the no of clusters to do the cluster analysis using Elbow Method

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



#Using Silhouette method for determining no of clusters

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

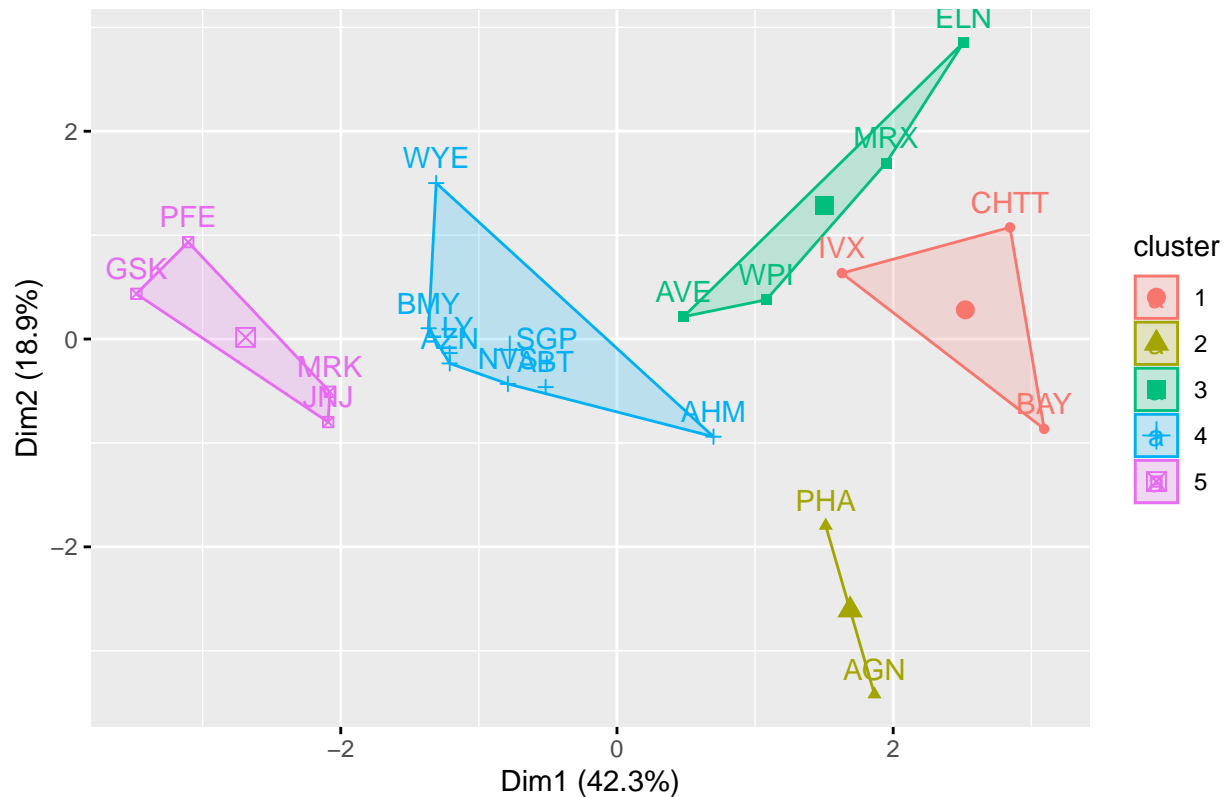
#Visualizing the output

```
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

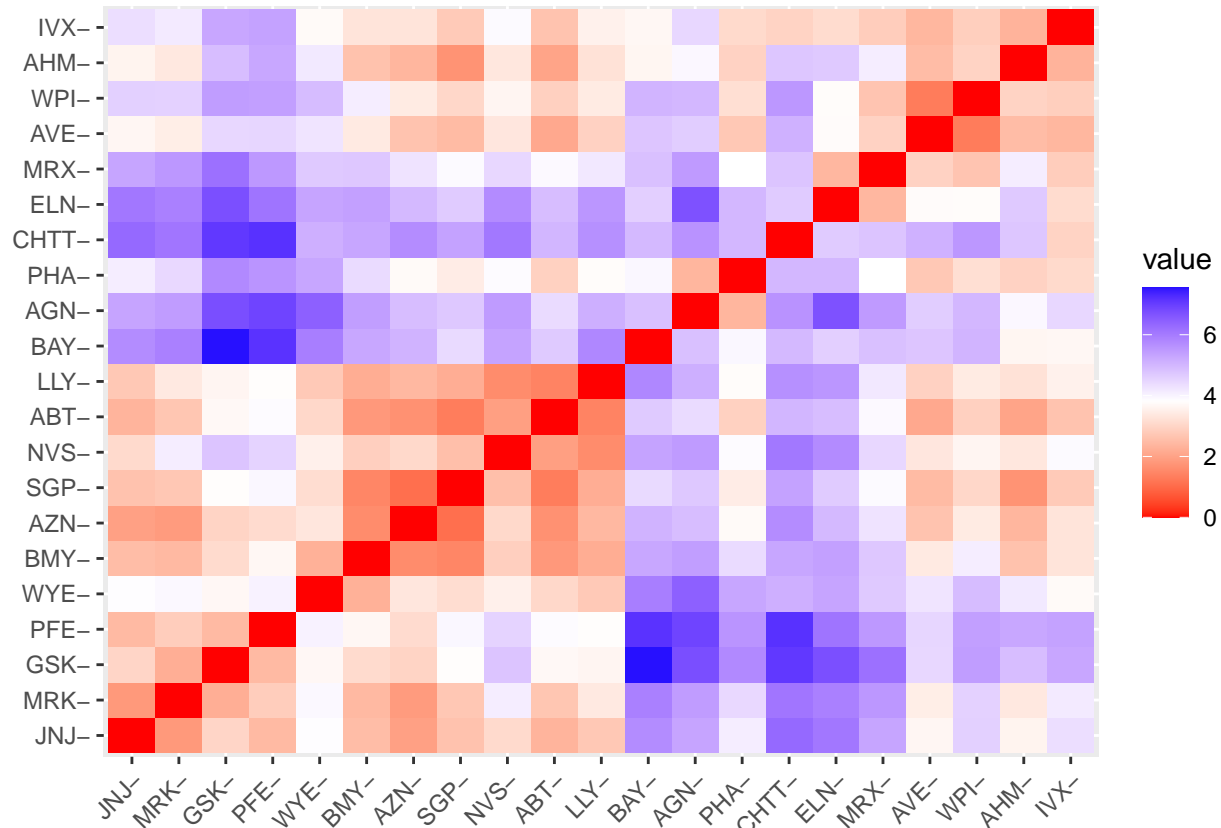


k5

```
## 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)
```



#Using K-Means Cluster Analysis- to Fit the data with 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
```

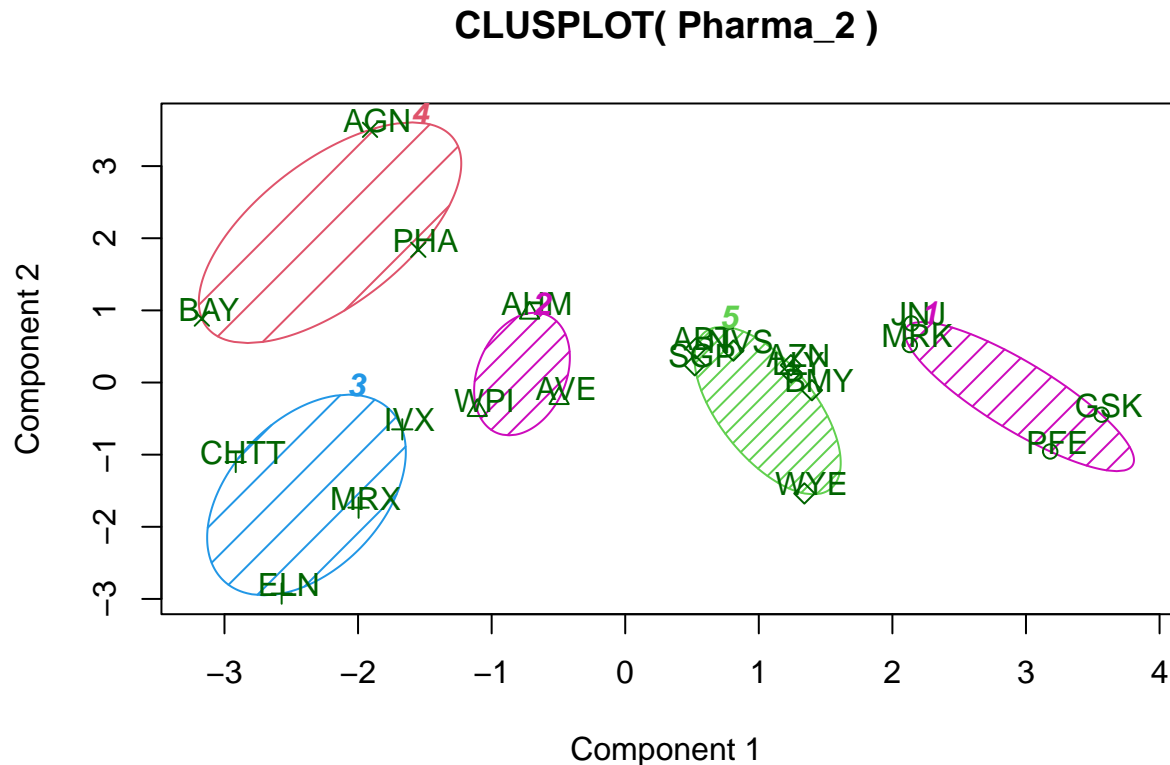
##	Market_Cap	Beta	PE_Ratio	ROE	ROA	Asset_Turnover
## ABT	0.1840960	-0.80125356	-0.04671323	0.04009035	0.2416121	0.0000000
## AGN	-0.8544181	-0.45070513	3.49706911	-0.85483986	-0.9422871	0.9225312
## AHM	-0.8762600	-0.25595600	-0.29195768	-0.72225761	-0.5100700	0.9225312
## AZN	0.1702742	-0.02225704	-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(Pharma_3)
```

```
#view of the cluster plot
```

```
library(cluster)
```

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



These two components explain 61.23 % of the point variability.

```
#Task-b. Interpret the clusters with respect to the numerical variables used in forming the clusters.  
#By looking at the mean values of all quantitative variables in each cluster.
```

```
#Cluster 1 - JNJ, MRK, PFE, GSK
```

```
#Cluster 2 - AHM, WPI, AVE
```

```
#Cluster 3 - CHTT, ELN, MRX, IVX
```

```
#Cluster 4 - BAY, PHA, AGN
```

```
#Cluster 5 - AZN, ABT, NVS, BMY, WYE, SGP, LLY
```

```
#Cluster 1 has highest Market_cap, ROA, ROE, Asset_Turnover and lowest is Beta, PE_Ratio.
```

```
#Cluster 2 has highest Rev_Growth and lowest PE_Ratio, Asset_Turnover.
```

#Cluster 3 has highest Beta, Leverage and lowest Market_Cap, ROE, ROA, Leverage, Rev_Growth, Net_Profit.

#Cluster 4 has highest PE_Ratio and lowest Leverage, Asset_Turnover.

#Cluster 5 has highest Net_Profit_Margin and lowest leverage,Beta.

#Task-c. Is there a pattern in the clusters with respect to the numerical variables (10 to 12)? (those forming the clusters)

#With respect to the Media recommendation variable, there is a pattern in the clusters.

#Cluster 1 with highest ROA, highest Asset_Turnover , highest Market_cap, highest ROE has equal Hold and Buy Recommendation.

#Cluster 2 with lowest Asset_Turnover and lowest PE_Ratio we considered as undervalued stock but moderate Recommendation.

#Cluster-3 with highest Leverage, highest Beta has mostly too risky nature it is less Recommendation than Buy.

#Cluster 4 with highest PE_Ratio that means all stocks in this cluster are overvalued so it advisable to sell.

#Cluster 5 with highest Net_Profit_Margin has mostly Hold Recommendation.

#In terms of variables, I have seen a pattern among the clusters (10 to 12)

#Clusters 1,3 has mostly Moderate Buy Recommendation

#Clusters 1,2,4,5 has Hold Recommendation

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

#Cluster-1 - strong Buy Recommendation (less risky nature)..

#Cluster-2 - Moderately Hold cluster

#Cluster-3 - high sell Recommendation cluster (high risky nature)

#Cluster-4 - low sell Recommendation cluster

#Cluster-5 - strong Hold cluster.