

# Assignment\_4

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```
library(factoextra)
```

```
## Warning: package 'factoextra' was built under R version 4.1.3
```

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

```
## Warning: package 'ggplot2' was built under R version 4.1.3
```

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

```
library(ggplot2)
```

```
Pharmaanalyst<-read.csv("C:/Users/ravin/Downloads/Pharmaceuticals.csv")
```

```
summary(Pharmaanalyst)
```

```
##      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) Use only the numerical variables (1 to 9) to cluster the 21 firms. Justify the various choices made in conducting the cluster analysis, such as weights for different variables, the specific clustering algorithm(s) used, the number of clusters formed, and so on. Prior to clustering data, remove the missing data and rescale variables for comparability.

```
x <- na.omit(Pharmaanalyst) #gives the data after removing the incomplete cases.
x
```

##	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
```

collect only the quantitative variables(1-9) to cluster the 21 firms

```
row.names(x)<- x[,1]
Pharma1<- x[,3:11]
head(Pharma1)
```

```
##      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 quantitative variables in the dataframe

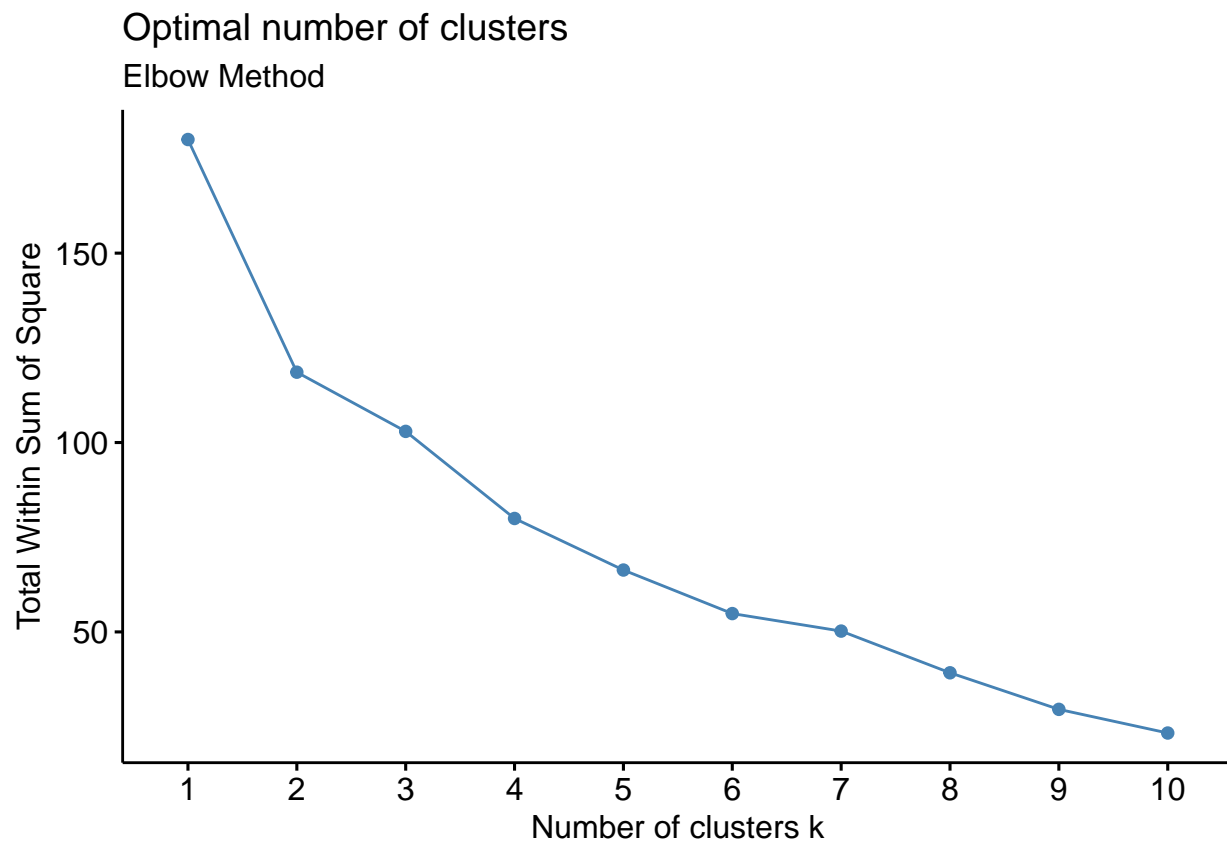
```
Pharma2<-scale(Pharma1)
head(Pharma2)
```

```
##      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
```

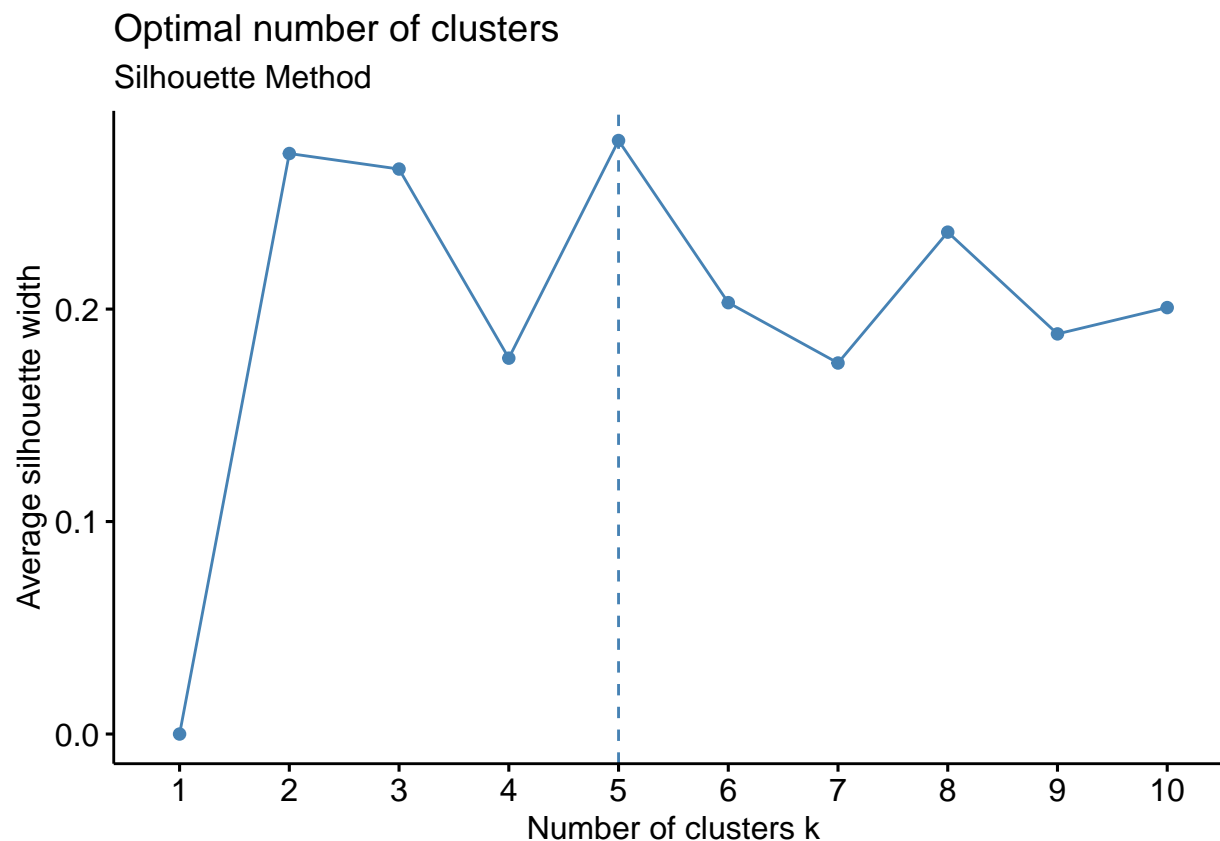
To determine the no of clusters to do the cluster analysis using Elbow Method

```
fviz_nbclust(Pharma2, kmeans, method = "wss") + labs(subtitle = "Elbow Method")
```



Silhouette method for determining no of clusters

```
fviz_nbclust(Pharma2, kmeans, method = "silhouette")+ labs(subtitle = "Silhouette Method")
```

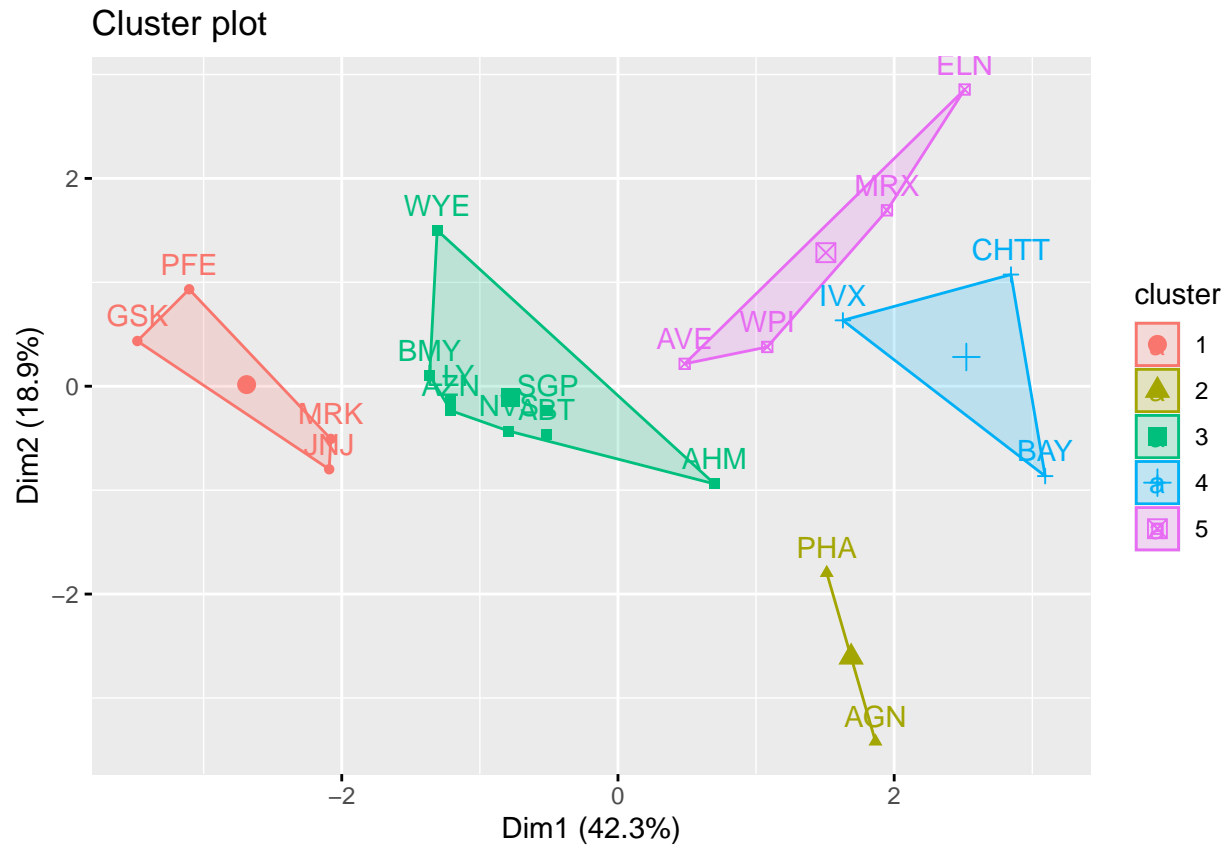


from the above plots, it is clear that the no of clusters are 5 and they are enough to show the variations that are present in the data

```
set.seed(120)
k5<- kmeans(Pharma2,centers=5,nstart = 25)
#Visualize the output
k5$centers #centroids
```

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

```
fviz_cluster(k5,data = Pharma2) # to Visualize the clusters
```



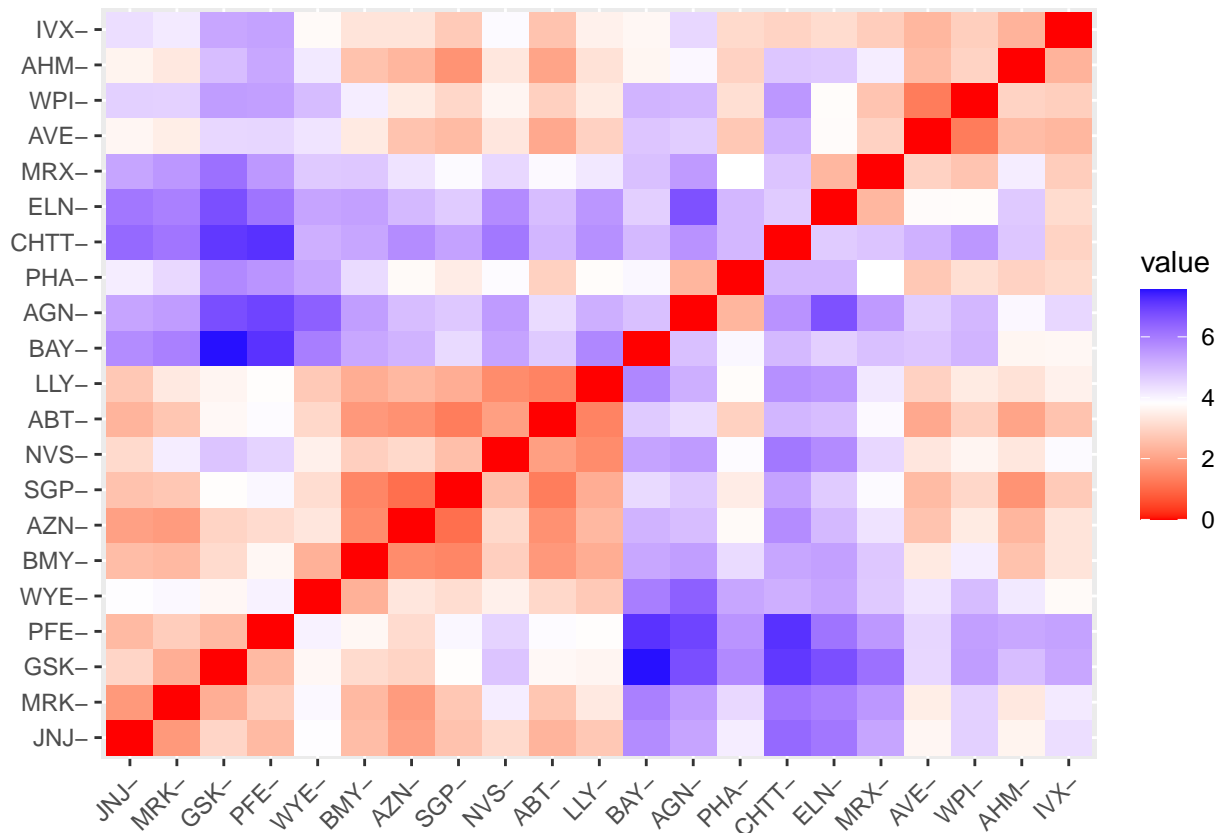
k5

```
## K-means clustering with 5 clusters of sizes 4, 2, 8, 3, 4
##
## Cluster means:
##   Market_Cap      Beta    PE_Ratio      ROE      ROA Asset_Turnover
## 1  1.69558112 -0.1780563 -0.19845823  1.2349879  1.3503431    1.1531640
## 2 -0.43925134 -0.4701800  2.70002464 -0.8349525 -0.9234951    0.2306328
## 3 -0.03142211 -0.4360989 -0.31724852  0.1950459  0.4083915    0.1729746
## 4 -0.87051511  1.3409869 -0.05284434 -0.6184015 -1.1928478   -0.4612656
## 5 -0.76022489  0.2796041 -0.47742380 -0.7438022 -0.8107428   -1.2684804
##   Leverage Rev_Growth Net_Profit_Margin
## 1 -0.46807818  0.4671788    0.591242521
## 2 -0.14170336 -0.1168459   -1.416514761
## 3 -0.27449312 -0.7041516    0.556954446
## 4  1.36644699 -0.6912914   -1.320000179
## 5  0.06308085  1.5180158   -0.006893899
##
## Clustering vector:
##  ABT  AGN  AHM  AZN  AVE  BAY  BMY  CHTT  ELN  LLY  GSK  IVX  JNJ  MRX  MRK  NVS
##   3    2    3    3    5    4    3    4    5    3    1    4    1    5    1    3
##  PFE  PHA  SGP  WPI  WYE
##   1    2    3    5    3
```

```
##
## Within cluster sum of squares by cluster:
## [1] 9.284424 2.803505 21.879320 15.595925 12.791257
## (between_SS / total_SS = 65.4 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"

```

```
distance<- dist(Pharma2, method = "euclidean")
fviz_dist(distance)
```



K-Means Cluster Analysis- Fit the data with 5 clusters

```
fit<-kmeans(Pharma2,5)
```

Finding the mean value of all quantitative variables for each cluster

```
aggregate(Pharma2,by=list(fit$cluster),FUN=mean)
```

```
##   Group.1 Market_Cap      Beta  PE_Ratio      ROE      ROA
## 1      1 -0.87051511  1.3409869 -0.05284434 -0.6184015 -1.1928478
## 2      2  0.08926902 -0.4618336 -0.32086149  0.3260892  0.5396003
## 3      3 -0.96686975  1.5162611 -0.57398880 -0.8382671 -0.9892673
```

```
## 4      4  1.69558112 -0.1780563 -0.19845823  1.2349879  1.3503431
## 5      5 -0.57238455 -0.6220844  0.86927480 -0.7381675 -0.7242993
## Asset_Turnover Leverage Rev_Growth Net_Profit_Margin
## 1 -4.612656e-01  1.3664470 -0.6912914      -1.3200002
## 2  6.589509e-02 -0.2559803 -0.7230135      0.7343816
## 3 -1.845062e+00  0.5302448  1.7123890      0.2445520
## 4  1.153164e+00 -0.4680782  0.4671788      0.5912425
## 5  1.776140e-16 -0.2991312  0.3682951      -0.8069490
```

```
Pharma3<-data.frame(Pharma2,fit$cluster)
Pharma3
```

```
##      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      2
## AGN  0.01828430 -0.38113909     -1.55366706      5
## AHM -0.40408312 -0.57211809     -0.68503583      5
## AZN -0.74965647  0.14744734      0.35122600      2
## AVE -0.31449003  1.21638667     -0.42597037      5
## BAY -0.74965647 -1.49714434     -1.99560225      1
## BMY -0.02011273 -0.96584257      0.74744375      2
## CHTT 3.74279705 -0.63276071     -1.24888417      1
## ELN  0.61983791  1.88617085     -0.36501379      3
## LLY -0.07130879 -0.64814764      1.17413980      2
## GSK -0.31449003  0.76926048      0.82363947      4
## IVX  1.10620040  0.05603085     -0.71551412      1
## JNJ -0.62166634 -0.36213170      0.33598685      4
## MRX  0.44065173  1.53860717      0.85411776      3
## MRK -0.39128411  0.36014907     -0.24310064      4
## NVS -0.67286239 -1.45369888      1.02174835      2
## PFE -0.54487226  1.10143723      1.44844440      4
## PHA -0.30169102  0.14744734     -1.27936246      5
## SGP -0.74965647 -0.43544591      0.29026942      2
```



```
## WPI -0.49367621 1.43089863 -0.09070919 5
## WYE 0.68383297 -1.17763919 1.49416183 2
```

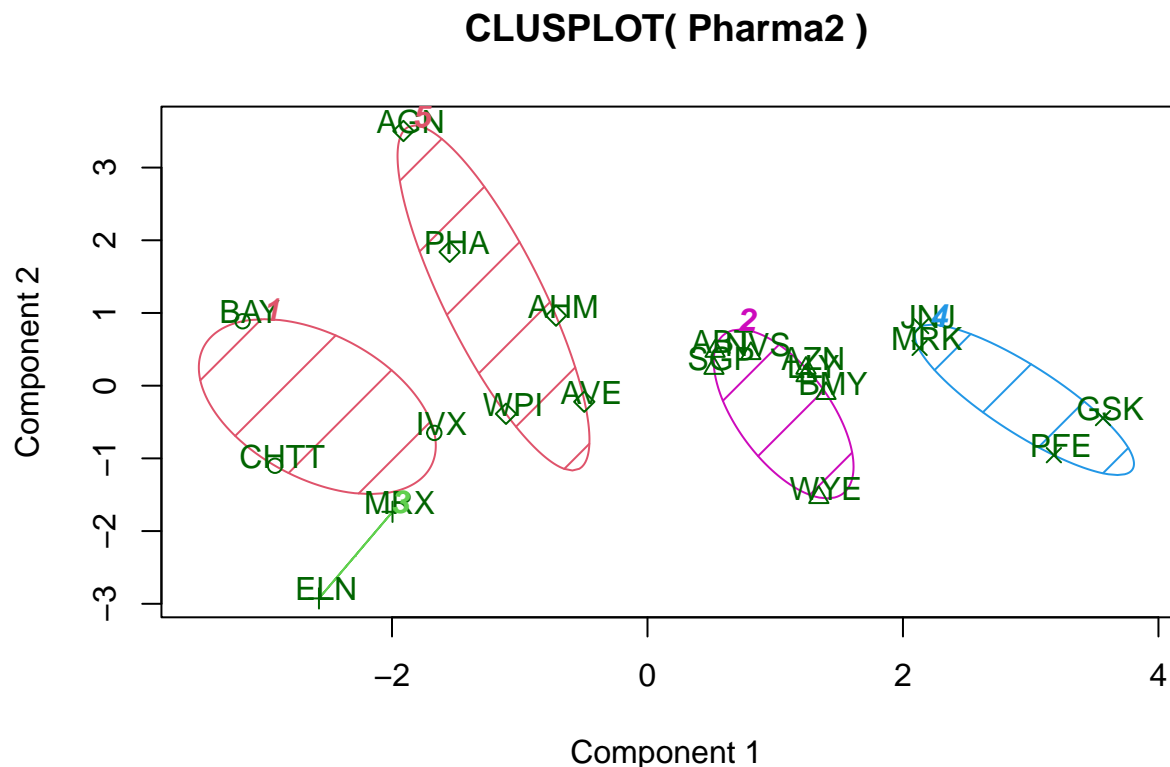
```
View(Pharma3)
```

To view the cluster plot

```
library(cluster)
```

```
## Warning: package 'cluster' was built under R version 4.1.3
```

```
clusplot(Pharma2,fit$cluster,color = TRUE,shade = TRUE,labels = 2,lines = 0)
```



These two components explain 61.23 % of the point variability.

#b) Interpret the clusters with respect to the numerical variables used in forming the clusters.

By observing the mean values of all quantitative variables for each cluster

Cluster 1 - BAY, CHTT, IVX

Cluster 2 - ABT, AZN, BMY, LLY, NVS, SGP, WYE

Cluster 3 - ELN, MRX

Cluster 4 - JNJ, MRK, PFE, GSK

Cluster 5 - AGN, AHM, AVE, PHA, WPI

Cluster 1 has highest Beta , Leverage and lowest Market\_Cap, ROE, ROA, Leverage, Rev\_Growth, Net\_Profit\_Margin Cluster 2 has highest Net\_Profit\_Margin and lowest Beta. Cluster 3 has highest Rev\_Growth and lowest PE\_Ratio, Asset\_Turnover. Cluster 4 has highest Market\_Cap, ROE, ROA,Asset\_Turnover Cluster 5 has highest PE\_Ratio.

#c)Is 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 with respect to Media recommendation variable.

Cluster 1 with highest Beta, highest Leverage has mostly Moderate Buy Recommendation.

Cluster 2 with highest Net\_Profit\_Margin has mostly Hold Recommendation

Cluster 3 with lowest PE\_Ratio and lowest Asset\_Turnover has Hold Recommendation

Cluster 4 with highest Market\_Cap, highest ROE, highest ROA, highest Asset\_Turnover has equal Hold and Moderate Buy Recommendation

Cluster 5 with highest PE\_Ratio has the Strong Buy Recommendation, because high PE\_Ratio indicates the company is growing fast.

Could see a pattern among the clusters with respect to variables(10 to 12)

Clusters 1,4 has mostly Moderate Buy Recommendation

Clusters 2,3,4 has Hold Recommendation

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

Cluster1 - high Beta, Leverage cluster (or) Buy Cluster.

Cluster2 - high Net\_Profit\_Margin cluster (or) high hold cluster.

Cluster3 - Low PE\_Ratio, Asset\_Turnover cluster (or) hold cluster.

Cluster4 - Moderate Buy cluster

Cluster5 - high PE\_Ratio cluster (or) high Buy cluster.