

# FML Assignment 4

JYOTHIRMAI MOPARTHI

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```
library(readr)
Pharmaceuticals_Read <- read.csv("./Pharmaceuticals.csv")
View(Pharmaceuticals_Read)
```

```
library(ggplot2)
library(factoextra)
```

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

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

```
library(flexclust)
```

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

```
## Loading required package: grid
```

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

```
## Loading required package: modeltools
```

```
## Loading required package: stats4
```

```
library(cluster)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.3      v stringr    1.5.0
## v forcats    1.0.0      v tibble     3.2.1
## v lubridate  1.9.2      v tidyr      1.3.0
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()    masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
summary(Pharmaceuticals_Read)
```

```
##      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
##
##
##
```

**Question 1 :** 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.

```
K <- na.omit(Pharmaceuticals_Read)
K
```

```
##      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
```

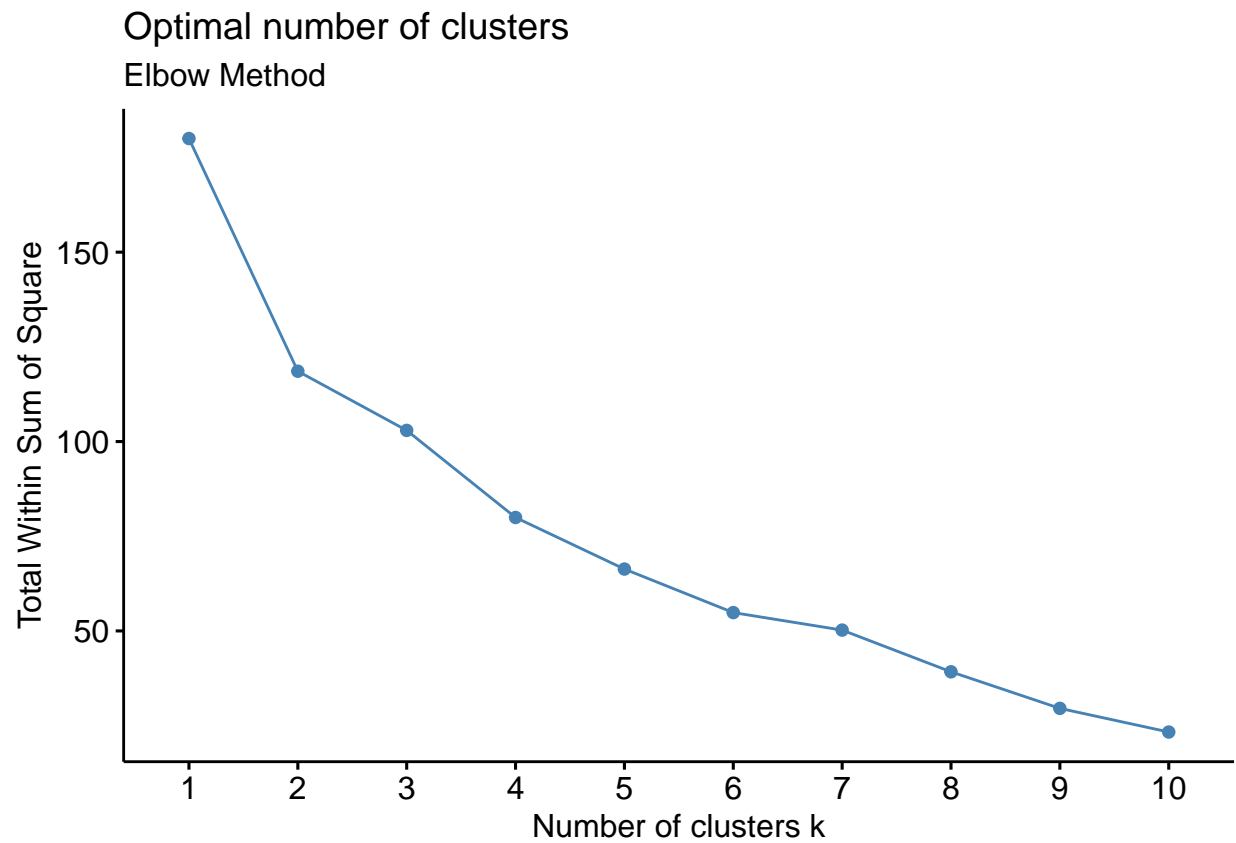
```
row.names <- K[,1]
pharma1 <- K[,3:11]
head(pharma1)
```

```
##   Market_Cap Beta PE_Ratio ROE ROA Asset_Turnover Leverage Rev_Growth
## 1    68.44 0.32    24.7 26.4 11.8          0.7    0.42      7.54
## 2     7.58 0.41    82.5 12.9  5.5          0.9    0.60      9.16
## 3     6.30 0.46    20.7 14.9  7.8          0.9    0.27      7.05
## 4    67.63 0.52    21.5 27.4 15.4          0.9    0.00     15.00
## 5    47.16 0.32    20.1 21.8  7.5          0.6    0.34     26.81
## 6    16.90 1.11    27.9  3.9  1.4          0.6    0.00     -3.17
##   Net_Profit_Margin
## 1             16.1
## 2              5.5
## 3             11.2
## 4             18.0
## 5             12.9
## 6              2.6
```

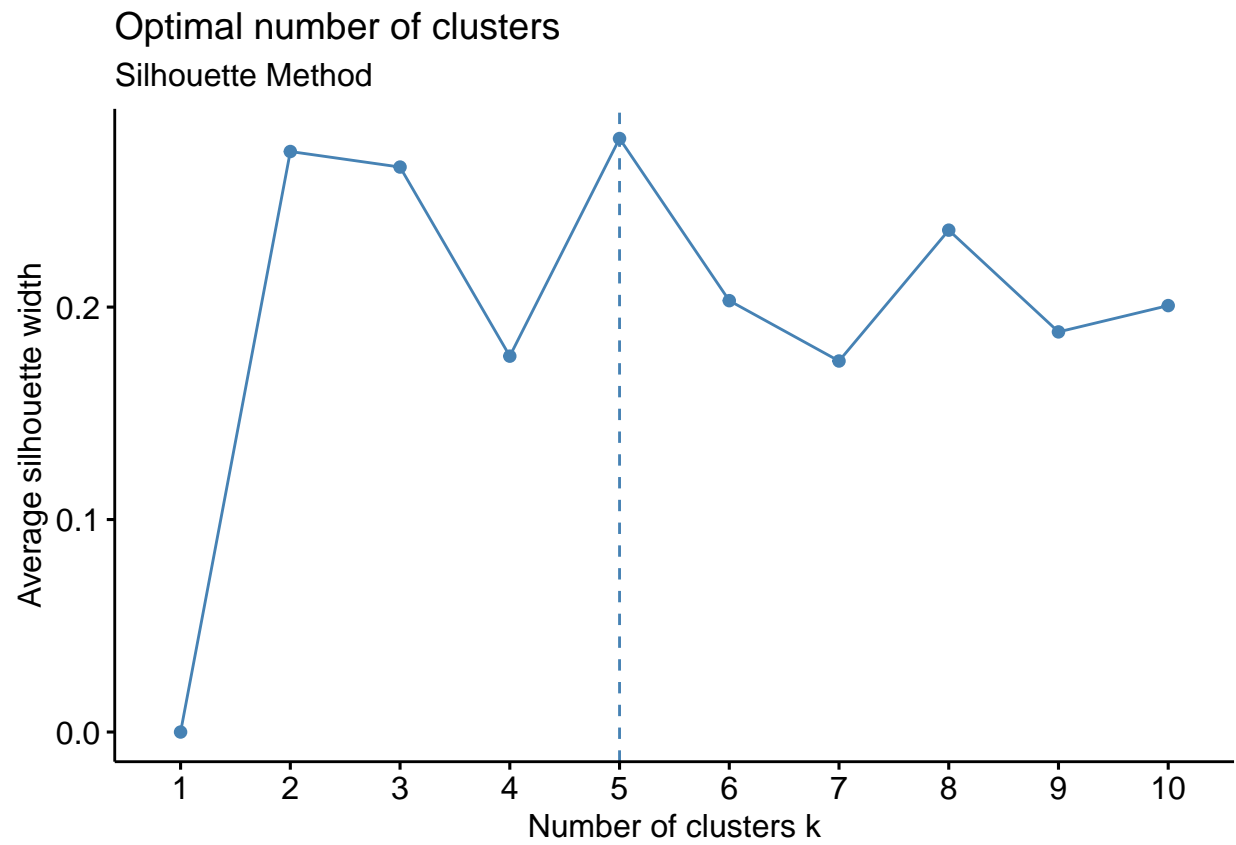
```
pharma2 <- scale(pharma1)
head(pharma2)
```

```
##   Market_Cap      Beta    PE_Ratio      ROE      ROA Asset_Turnover
## 1  0.1840960 -0.80125356 -0.04671323  0.04009035  0.2416121  0.0000000
## 2 -0.8544181 -0.45070513  3.49706911 -0.85483986 -0.9422871  0.9225312
## 3 -0.8762600 -0.25595600 -0.29195768 -0.72225761 -0.5100700  0.9225312
## 4  0.1702742 -0.02225704 -0.24290879  0.10638147  0.9181259  0.9225312
## 5 -0.1790256 -0.80125356 -0.32874435 -0.26484883 -0.5664461 -0.4612656
## 6 -0.6953818  2.27578267  0.14948233 -1.45146000 -1.7127612 -0.4612656
##   Leverage Rev_Growth Net_Profit_Margin
## 1 -0.2120979 -0.5277675      0.06168225
## 2  0.0182843 -0.3811391     -1.55366706
## 3 -0.4040831 -0.5721181     -0.68503583
## 4 -0.7496565  0.1474473      0.35122600
## 5 -0.3144900  1.2163867     -0.42597037
## 6 -0.7496565 -1.4971443     -1.99560225
```

```
fviz_nbclust(pharma2, kmeans, method = "wss") + labs(subtitle = "Elbow Method")
```



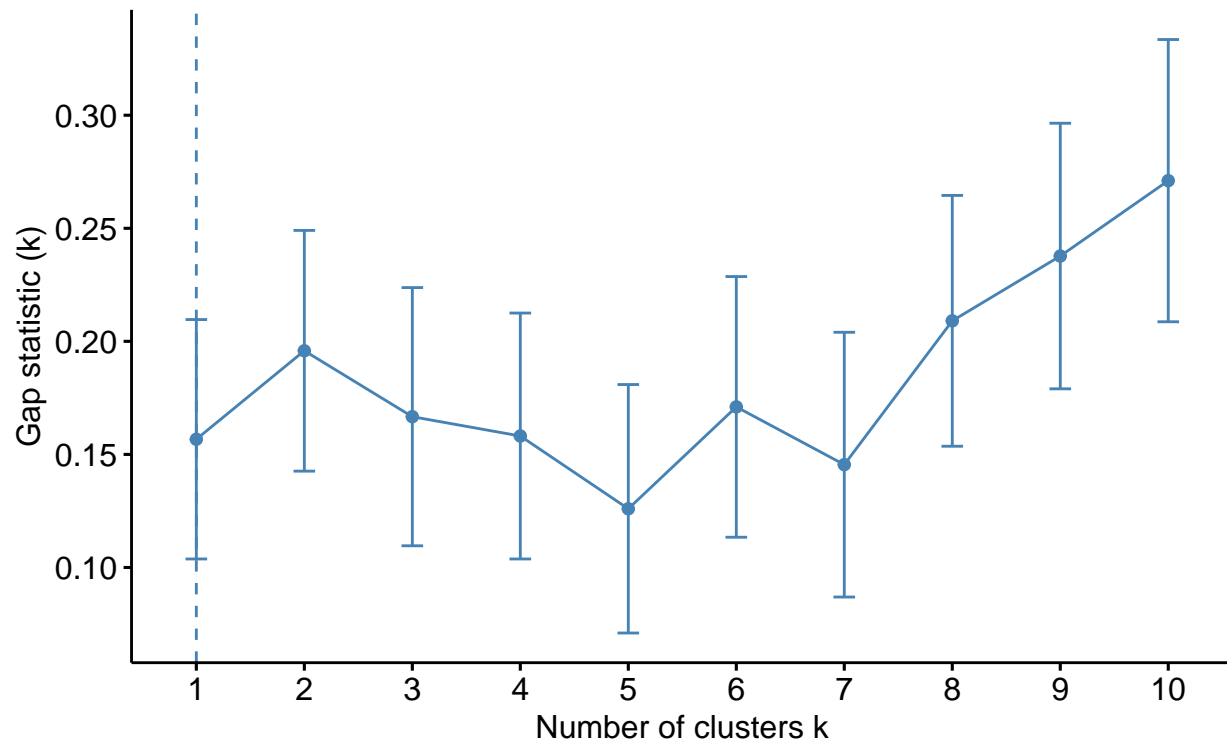
```
fviz_nbclust(pharma2, kmeans, method = "silhouette") + labs(subtitle = "Silhouette Method")
```



```
fviz_nbclust(pharma2, kmeans, method = "gap_stat") + labs(subtitle= "Gap Stat Method")
```

## Optimal number of clusters

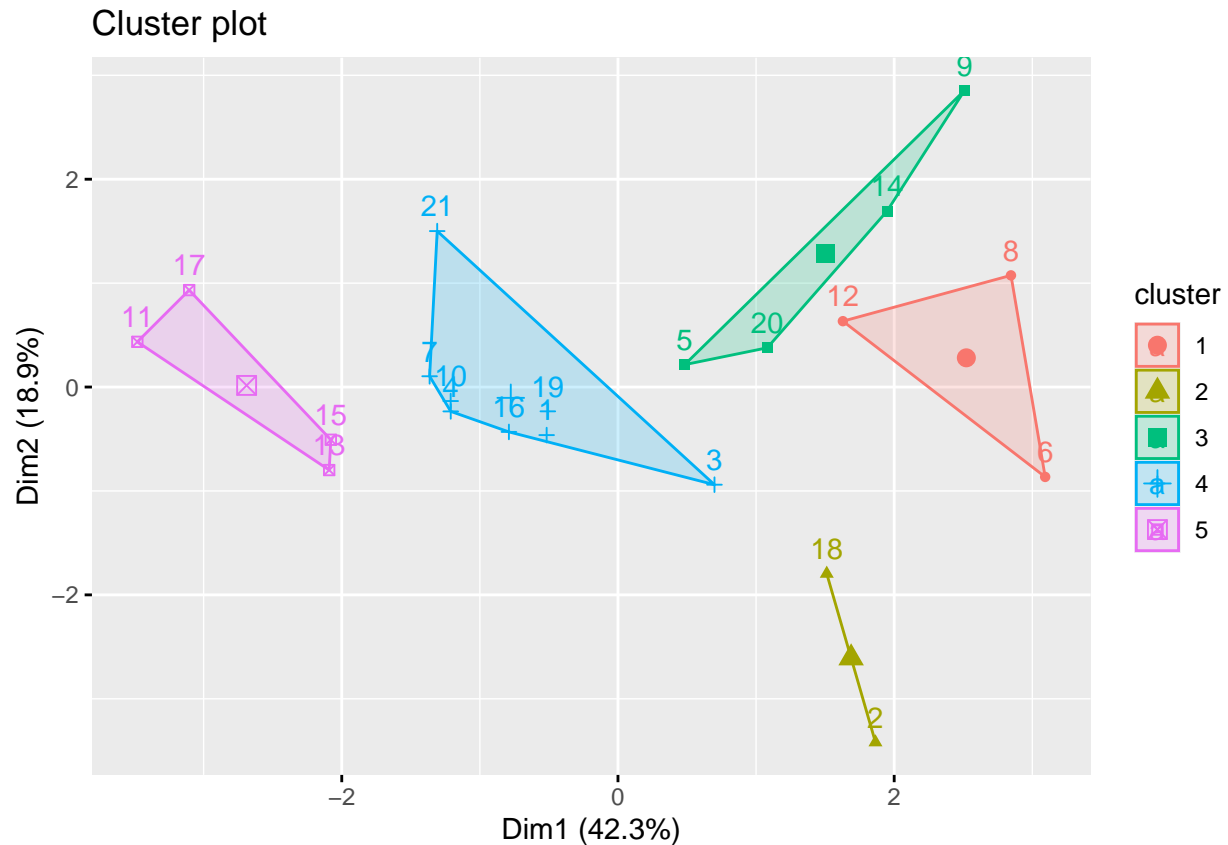
Gap Stat Method



```
set.seed(64060)
k5 <- kmeans (pharma2, centers = 5, nstart =25)
k5$centers
```

```
##      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= pharma2)
```



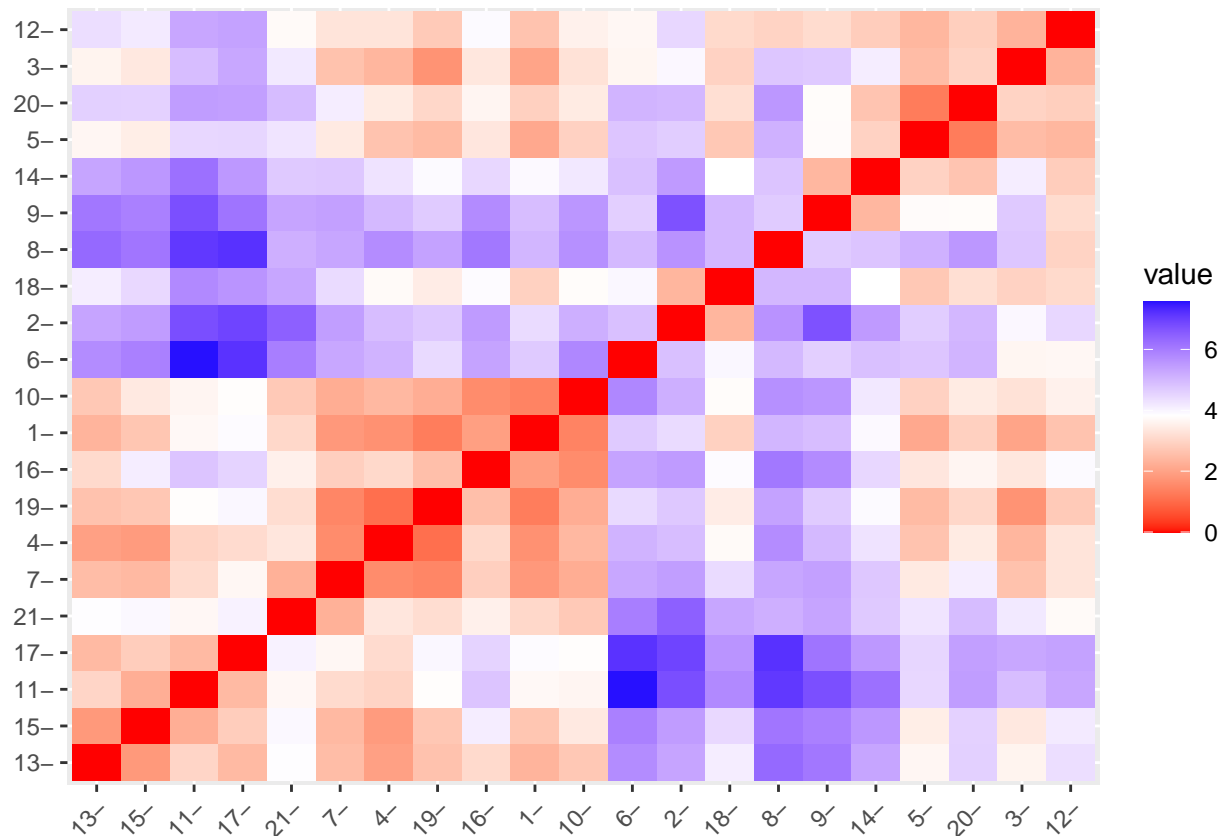
k5

```
## 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:
##  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21
##  4  2  4  4  3  1  4  1  3  4  5  1  5  3  5  4  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(pharma2,method = "euclidian")
fviz_dist(Distance)
```



```
Fitting <- kmeans(pharma2,5)
aggregate(pharma2,by = list(Fitting$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

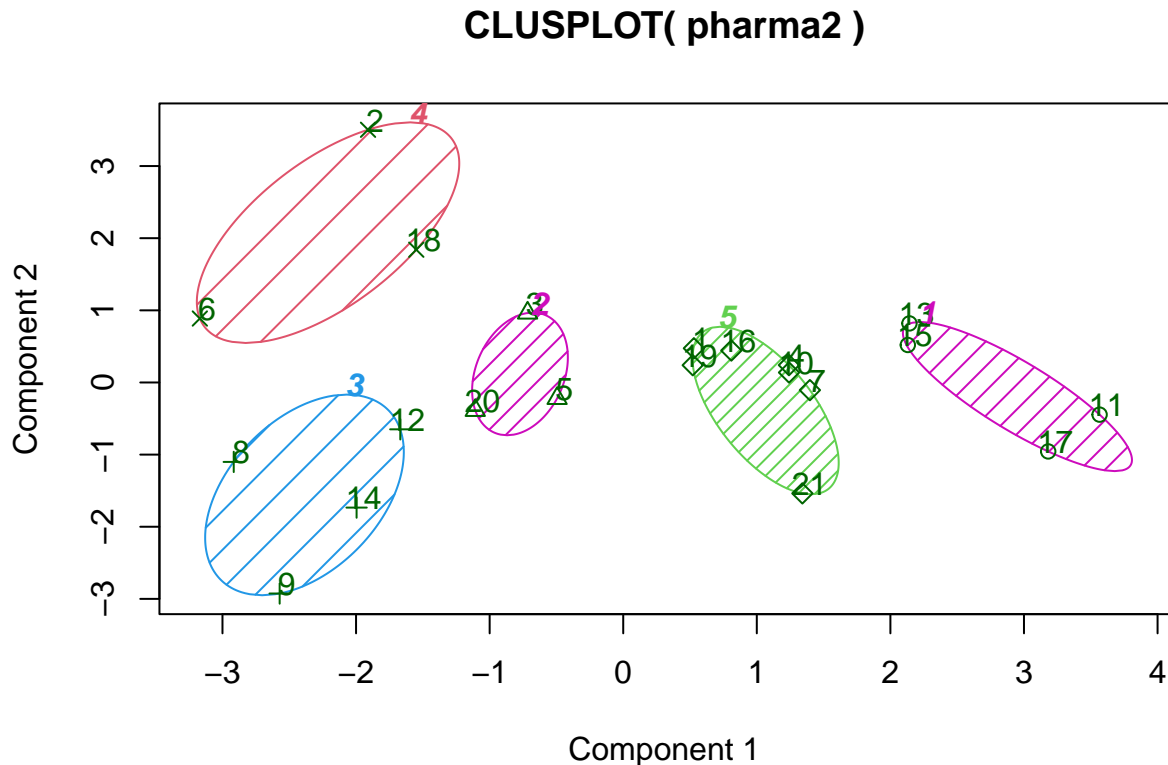
  

##	Asset_Turnover	Leverage	Rev_Growth	Net_Profit_Margin
## 1	1.153164e+00	-0.4680782	0.4671788	0.5912425
## 2	-1.537552e-01	-0.4040831	0.6917224	-0.4005718
## 3	-1.153164e+00	1.4773718	0.7120120	-0.3688236
## 4	1.480297e-16	-0.3443544	-0.5769454	-1.6095439
## 5	6.589509e-02	-0.2559803	-0.7230135	0.7343816

```
pharma3 <- data.frame(pharma2,Fitting$cluster)
pharma3
```

##	Market_Cap	Beta	PE_Ratio	ROE	ROA	Asset_Turnover
## 1	0.1840960	-0.80125356	-0.04671323	0.04009035	0.2416121	0.0000000
## 2	-0.8544181	-0.45070513	3.49706911	-0.85483986	-0.9422871	0.9225312
## 3	-0.8762600	-0.25595600	-0.29195768	-0.72225761	-0.5100700	0.9225312
## 4	0.1702742	-0.02225704	-0.24290879	0.10638147	0.9181259	0.9225312
## 5	-0.1790256	-0.80125356	-0.32874435	-0.26484883	-0.5664461	-0.4612656
## 6	-0.6953818	2.27578267	0.14948233	-1.45146000	-1.7127612	-0.4612656
## 7	-0.1078688	-0.10015669	-0.70887325	0.59693581	0.8617498	0.9225312
## 8	-0.9767669	1.26308721	0.03299122	-0.11237924	-1.1677918	-0.4612656
## 9	-0.9704532	2.15893320	-1.34037772	-0.70899938	-1.0174553	-1.8450624
## 10	0.2762415	-1.34655112	0.14948233	0.34502953	0.5610770	-0.4612656
## 11	1.0999201	-0.68440408	-0.45749769	2.45971647	1.8389364	1.3837968
## 12	-0.9393967	0.48409069	-0.34100657	-0.29136529	-0.6979905	-0.4612656
## 13	1.9841758	-0.25595600	0.18013789	0.18593083	1.0872544	0.9225312
## 14	-0.9632863	0.87358895	0.19240011	-0.96753478	-0.9610792	-1.8450624
## 15	1.2782387	-0.25595600	-0.40231769	0.98142435	0.8429577	1.8450624
## 16	0.6654710	-1.30760129	-0.23677768	-0.52338423	0.1288598	-0.9225312
## 17	2.4199899	0.48409069	-0.11415545	1.31287998	1.6322239	0.4612656
## 18	-0.0240846	-0.48965495	1.90298017	-0.81506519	-0.9047030	-0.4612656
## 19	-0.4018812	-0.06120687	-0.40231769	-0.21181593	0.5234929	0.4612656
## 20	-0.9281345	-1.11285216	-0.43297324	-1.03382590	-0.6979905	-0.9225312
## 21	-0.1614497	0.40619104	-0.75792214	1.92938746	0.5422849	-0.4612656
##	Leverage	Rev_Growth	Net_Profit_Margin	Fitting.cluster		
## 1	-0.21209793	-0.52776752	0.06168225	5		
## 2	0.01828430	-0.38113909	-1.55366706	4		
## 3	-0.40408312	-0.57211809	-0.68503583	2		
## 4	-0.74965647	0.14744734	0.35122600	5		
## 5	-0.31449003	1.21638667	-0.42597037	2		
## 6	-0.74965647	-1.49714434	-1.99560225	4		
## 7	-0.02011273	-0.96584257	0.74744375	5		
## 8	3.74279705	-0.63276071	-1.24888417	3		
## 9	0.61983791	1.88617085	-0.36501379	3		
## 10	-0.07130879	-0.64814764	1.17413980	5		
## 11	-0.31449003	0.76926048	0.82363947	1		
## 12	1.10620040	0.05603085	-0.71551412	3		
## 13	-0.62166634	-0.36213170	0.33598685	1		
## 14	0.44065173	1.53860717	0.85411776	3		
## 15	-0.39128411	0.36014907	-0.24310064	1		
## 16	-0.67286239	-1.45369888	1.02174835	5		
## 17	-0.54487226	1.10143723	1.44844440	1		
## 18	-0.30169102	0.14744734	-1.27936246	4		
## 19	-0.74965647	-0.43544591	0.29026942	5		
## 20	-0.49367621	1.43089863	-0.09070919	2		
## 21	0.68383297	-1.17763919	1.49416183	5		

```
library(cluster)
clusplot(pharma2,Fitting$cluster,color=TRUE, shade = TRUE, labels = 2, lines=0)
```



These two components explain 61.23 % of the point variability.

**Question 2:** Interpret the clusters with respect to the numerical variables used in forming the clusters. Is there a pattern in the clusters with respect to the numerical variables (10 to 12)? (those not used in forming the clusters)

```
aggregate(pharma2, by=list(Fitting$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

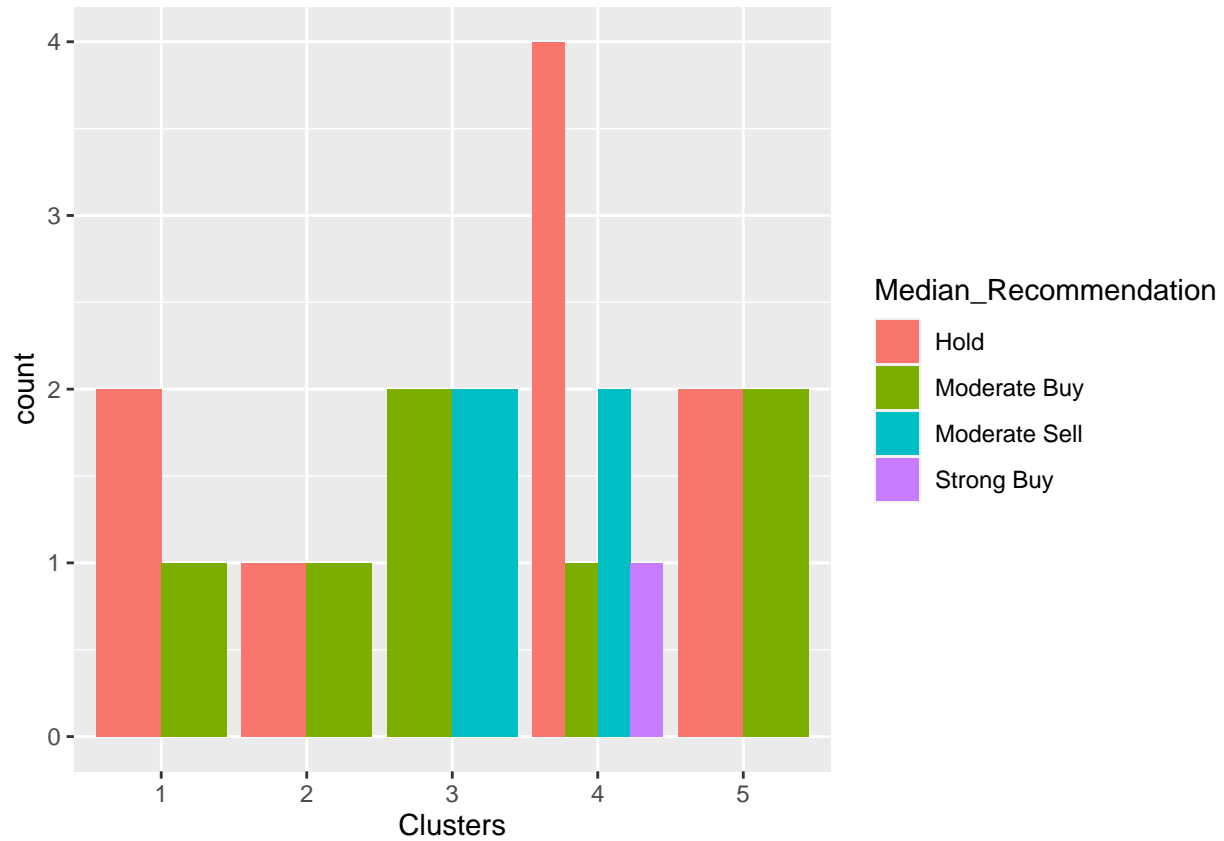
##	Asset_Turnover	Leverage	Rev_Growth	Net_Profit_Margin
## 1	1.153164e+00	-0.4680782	0.4671788	0.5912425
## 2	-1.537552e-01	-0.4040831	0.6917224	-0.4005718
## 3	-1.153164e+00	1.4773718	0.7120120	-0.3688236
## 4	1.480297e-16	-0.3443544	-0.5769454	-1.6095439
## 5	6.589509e-02	-0.2559803	-0.7230135	0.7343816

```
pharmacy <- data.frame(pharma2,k5$cluster)
pharmacy
```

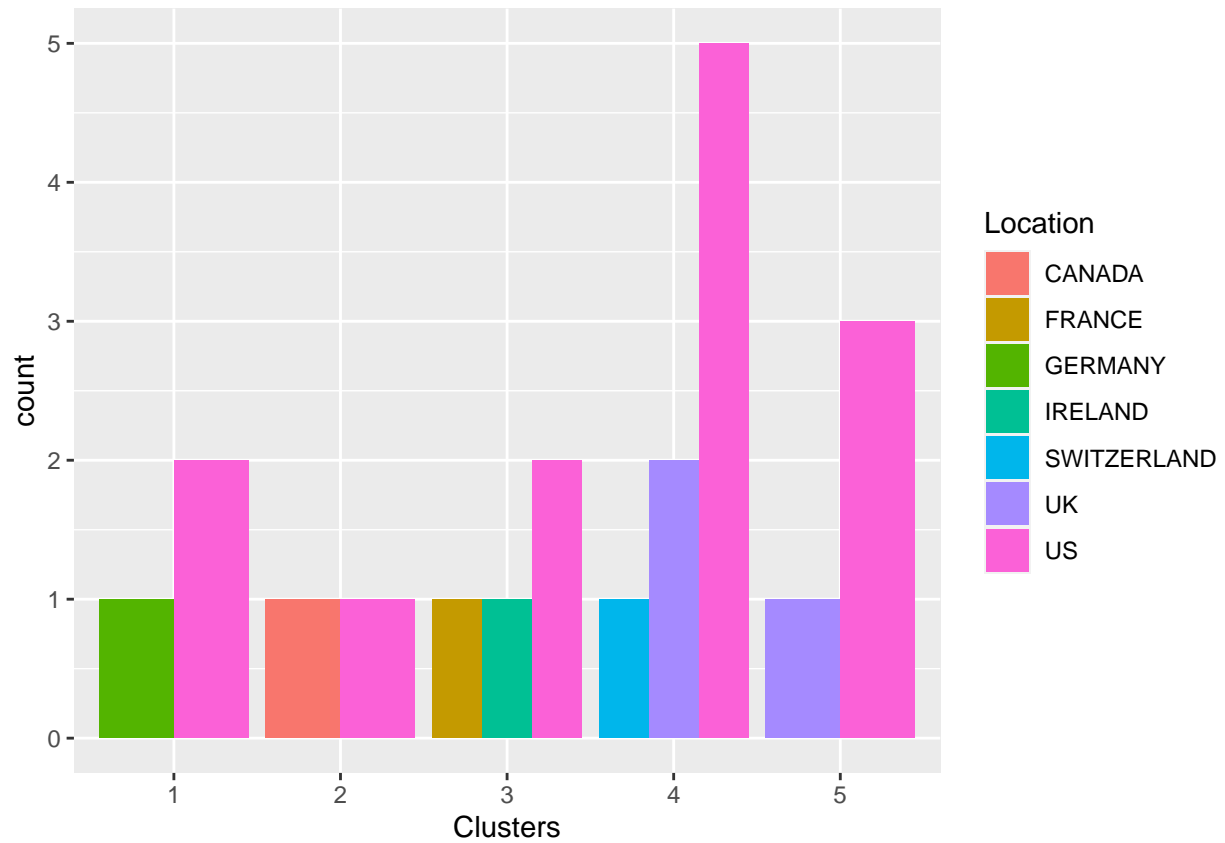
##	Market_Cap	Beta	PE_Ratio	ROE	ROA	Asset_Turnover
## 1	0.1840960	-0.80125356	-0.04671323	0.04009035	0.2416121	0.0000000
## 2	-0.8544181	-0.45070513	3.49706911	-0.85483986	-0.9422871	0.9225312
## 3	-0.8762600	-0.25595600	-0.29195768	-0.72225761	-0.5100700	0.9225312
## 4	0.1702742	-0.02225704	-0.24290879	0.10638147	0.9181259	0.9225312
## 5	-0.1790256	-0.80125356	-0.32874435	-0.26484883	-0.5664461	-0.4612656
## 6	-0.6953818	2.27578267	0.14948233	-1.45146000	-1.7127612	-0.4612656
## 7	-0.1078688	-0.10015669	-0.70887325	0.59693581	0.8617498	0.9225312
## 8	-0.9767669	1.26308721	0.03299122	-0.11237924	-1.1677918	-0.4612656
## 9	-0.9704532	2.15893320	-1.34037772	-0.70899938	-1.0174553	-1.8450624
## 10	0.2762415	-1.34655112	0.14948233	0.34502953	0.5610770	-0.4612656
## 11	1.0999201	-0.68440408	-0.45749769	2.45971647	1.8389364	1.3837968
## 12	-0.9393967	0.48409069	-0.34100657	-0.29136529	-0.6979905	-0.4612656
## 13	1.9841758	-0.25595600	0.18013789	0.18593083	1.0872544	0.9225312
## 14	-0.9632863	0.87358895	0.19240011	-0.96753478	-0.9610792	-1.8450624
## 15	1.2782387	-0.25595600	-0.40231769	0.98142435	0.8429577	1.8450624
## 16	0.6654710	-1.30760129	-0.23677768	-0.52338423	0.1288598	-0.9225312
## 17	2.4199899	0.48409069	-0.11415545	1.31287998	1.6322239	0.4612656
## 18	-0.0240846	-0.48965495	1.90298017	-0.81506519	-0.9047030	-0.4612656
## 19	-0.4018812	-0.06120687	-0.40231769	-0.21181593	0.5234929	0.4612656
## 20	-0.9281345	-1.11285216	-0.43297324	-1.03382590	-0.6979905	-0.9225312
## 21	-0.1614497	0.40619104	-0.75792214	1.92938746	0.5422849	-0.4612656
##	Leverage	Rev_Growth	Net_Profit_Margin	k5.cluster		
## 1	-0.21209793	-0.52776752	0.06168225	4		
## 2	0.01828430	-0.38113909	-1.55366706	2		
## 3	-0.40408312	-0.57211809	-0.68503583	4		
## 4	-0.74965647	0.14744734	0.35122600	4		
## 5	-0.31449003	1.21638667	-0.42597037	3		
## 6	-0.74965647	-1.49714434	-1.99560225	1		
## 7	-0.02011273	-0.96584257	0.74744375	4		
## 8	3.74279705	-0.63276071	-1.24888417	1		
## 9	0.61983791	1.88617085	-0.36501379	3		
## 10	-0.07130879	-0.64814764	1.17413980	4		
## 11	-0.31449003	0.76926048	0.82363947	5		
## 12	1.10620040	0.05603085	-0.71551412	1		
## 13	-0.62166634	-0.36213170	0.33598685	5		
## 14	0.44065173	1.53860717	0.85411776	3		
## 15	-0.39128411	0.36014907	-0.24310064	5		
## 16	-0.67286239	-1.45369888	1.02174835	4		
## 17	-0.54487226	1.10143723	1.44844440	5		
## 18	-0.30169102	0.14744734	-1.27936246	2		
## 19	-0.74965647	-0.43544591	0.29026942	4		
## 20	-0.49367621	1.43089863	-0.09070919	3		
## 21	0.68383297	-1.17763919	1.49416183	4		

#Cluster 1:- JNJ, MRK, GSK, PFE #Highest Market\_Cap and lowest Beta/PE Ratio #Cluster 2:- AHM, WPI, AVE #Highest Revenue Growth and lowest PE/Asset Turnover Ratio #Cluster 3:- CHTT, IVX, MRX, ELN #Highest Beta/leverage/Asset Turnover Ratio and lowest Net\_Profit\_Margin, PE ratio and Market #Cluster 4:- AGN,BAY, PHA #Highest PE ratio and lowest Leverage/Asset\_Turnover #Cluster 5:- ABT, WYE, AZN, SGP, BMY, NVS, LLY #Highest Net\_Profit\_Margin and lowest Leverage

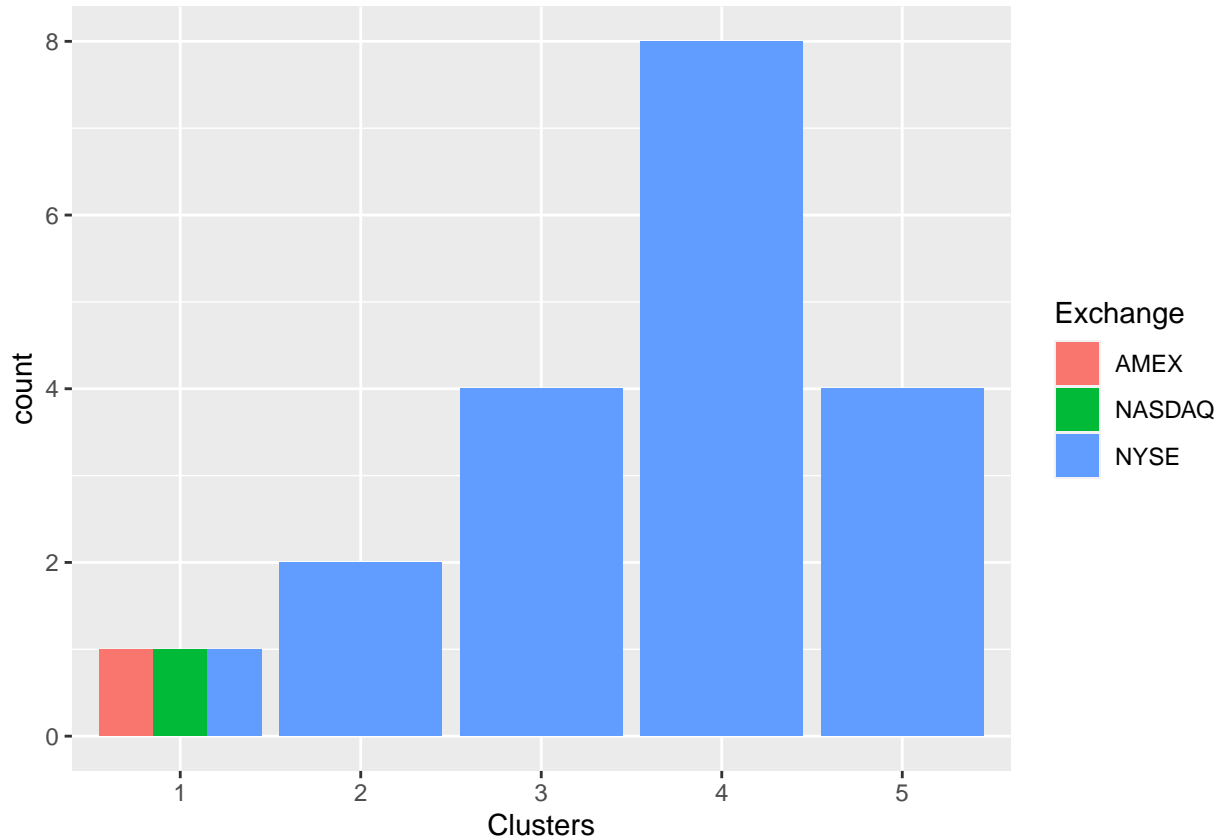
```
Read <- Pharmaceuticals_Read[12:14] %>% mutate(Clusters=k5$cluster)
ggplot(Read,mapping= aes(factor(Clusters),
                           fill = Median_Recommendation))+
  geom_bar(position = 'dodge')+labs(x= 'Clusters')
```



```
ggplot(Read,mapping= aes(factor(Clusters),
                           fill = Location))+
  geom_bar(position = 'dodge')+labs(x= 'Clusters')
```



```
ggplot(Read,mapping= aes(factor(Clusters),  
                           fill = Exchange))+  
  geom_bar(position = 'dodge')+labs(x= 'Clusters')
```



#The above graphs indicate that there is a slim pattern in the clusters.

#The firms are evenly distributed throughout AMEX, NASDAQ, and NYSE, despite the fact that cluster 1 has a different Hold and Moderate Buy median, a different count from the US and Germany, and a distinct nation count.

#The cluster 2 is equally distributed throughout the US and Canada, has equal Hold and Moderate Buy medians, and is completely listed on the NYSE.

#The Cluster 3 is listed on the NYSE and has equal Moderate Buy and Sell medians and also a separate count from France, Ireland, and the US.

#In Cluster 4, the Hold median is the highest, followed by the medians for Moderate Buy, Strong Buy, and Hold. They are listed on the NYSE and originate from the US, the UK, and Switzerland.

#The Cluster 5 has the same hold and moderate buy medians, is listed on the NYSE, and distributes throughout the US and the UK.

**Question 3: Provide an appropriate name for each cluster using any or all of the variables in the dataset.**

#Cluster1 :- Buy Cluster #Cluster2 :- Sceptical Cluster #Cluster3 :- Moderate Buy Cluster #Cluster4 :- Hold Cluster #Cluster5 :- High Hold Cluster