

Assignment_4

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Description: "The purpose of this assignment is to use k-Means for clustering". load the required libraries

```
library(caret)
```

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

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

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

```
## Warning: package 'factoextra' was built under R version 4.3.3
```

```
## 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 4.3.3
```

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

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

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

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

```
## Warning: package 'tidyverse' was built under R version 4.3.3
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.4      v stringr    1.5.1
```

```
## v forcats   1.0.0      v tibble     3.2.1
```

```
## v lubridate 1.9.3      v tidyr      1.3.1
```

```
## v purrr     1.0.2
```

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

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

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

```
## x purrr::lift()    masks caret::lift()
```

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

Get the current working directory

```
getwd()
```

```
## [1] "C:/Users/tarun/Downloads"
```

Set the working directory to “C:/Users/tarun/Downloads”

```
setwd("C:/Users/tarun/Downloads")
```

Read the CSV file “Pharmaceuticals.csv” and store it in the data frame named “Pharma_Assignment4”

```
Pharma_Assignment4 <- read.csv("Pharmaceuticals.csv")
```

Print the structure of the data frame

```
str(Pharma_Assignment4)
```

```
## 'data.frame':    21 obs. of  14 variables:
##  $ Symbol      : chr  "ABT" "AGN" "AHM" "AZN" ...
##  $ Name        : chr  "Abbott Laboratories" "Allergan, Inc." "Amersham plc" "AstraZeneca PL
##  $ Market_Cap  : num  68.44 7.58 6.3 67.63 47.16 ...
##  $ Beta        : num  0.32 0.41 0.46 0.52 0.32 1.11 0.5 0.85 1.08 0.18 ...
##  $ PE_Ratio    : num  24.7 82.5 20.7 21.5 20.1 27.9 13.9 26 3.6 27.9 ...
##  $ ROE         : num  26.4 12.9 14.9 27.4 21.8 3.9 34.8 24.1 15.1 31 ...
##  $ ROA         : num  11.8 5.5 7.8 15.4 7.5 1.4 15.1 4.3 5.1 13.5 ...
##  $ Asset_Turnover : num  0.7 0.9 0.9 0.9 0.6 0.6 0.9 0.6 0.3 0.6 ...
##  $ Leverage    : num  0.42 0.6 0.27 0 0.34 0 0.57 3.51 1.07 0.53 ...
##  $ Rev_Growth   : num  7.54 9.16 7.05 15 26.81 ...
##  $ Net_Profit_Margin : num  16.1 5.5 11.2 18 12.9 2.6 20.6 7.5 13.3 23.4 ...
##  $ Median_Recommendation: chr  "Moderate Buy" "Moderate Buy" "Strong Buy" "Moderate Sell" ...
##  $ Location     : chr  "US" "CANADA" "UK" "UK" ...
##  $ Exchange     : chr  "NYSE" "NYSE" "NYSE" "NYSE" ...
```

Print a summary of the data frame

```
summary(Pharma_Assignment4)
```

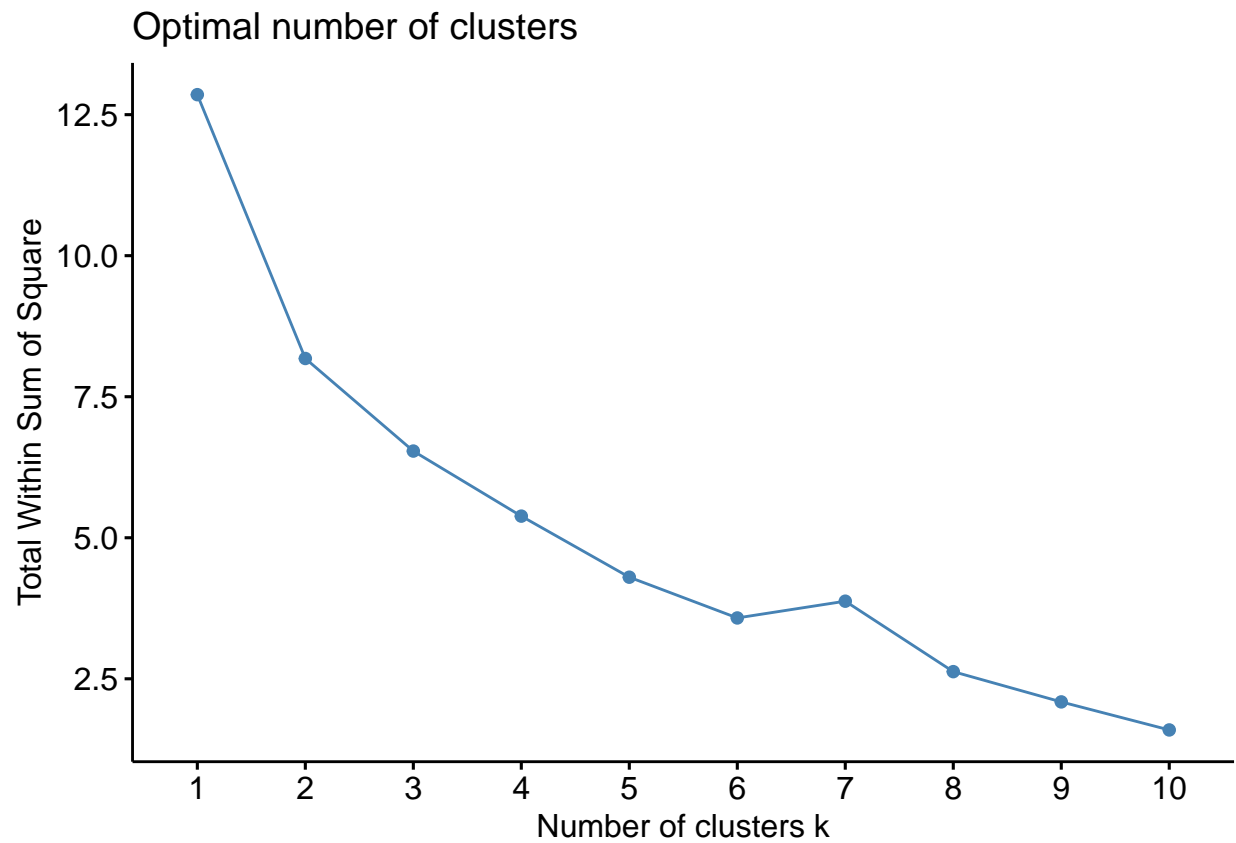
```
##      Symbol      Name      Market_Cap      Beta
## Length:21      Length:21      Min.   : 0.41      Min.   :0.1800
## Class :character Class :character 1st Qu.: 6.30      1st Qu.:0.3500
## Mode  :character Mode  :character Median : 48.19      Median :0.4600
##                                     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. # Select only the numerical variables from the data frame # Preprocess the numerical variables by scaling them using the range method

```
norm_mean<- Pharma_Assignment4 %>% select('Market_Cap', 'Beta', 'PE_Ratio', 'ROE', 'ROA', 'Asset_Turnover')
#scaling the data.
norm_train <- preProcess(norm_mean, method = "range")
predictions<-predict(norm_train, norm_mean)
```

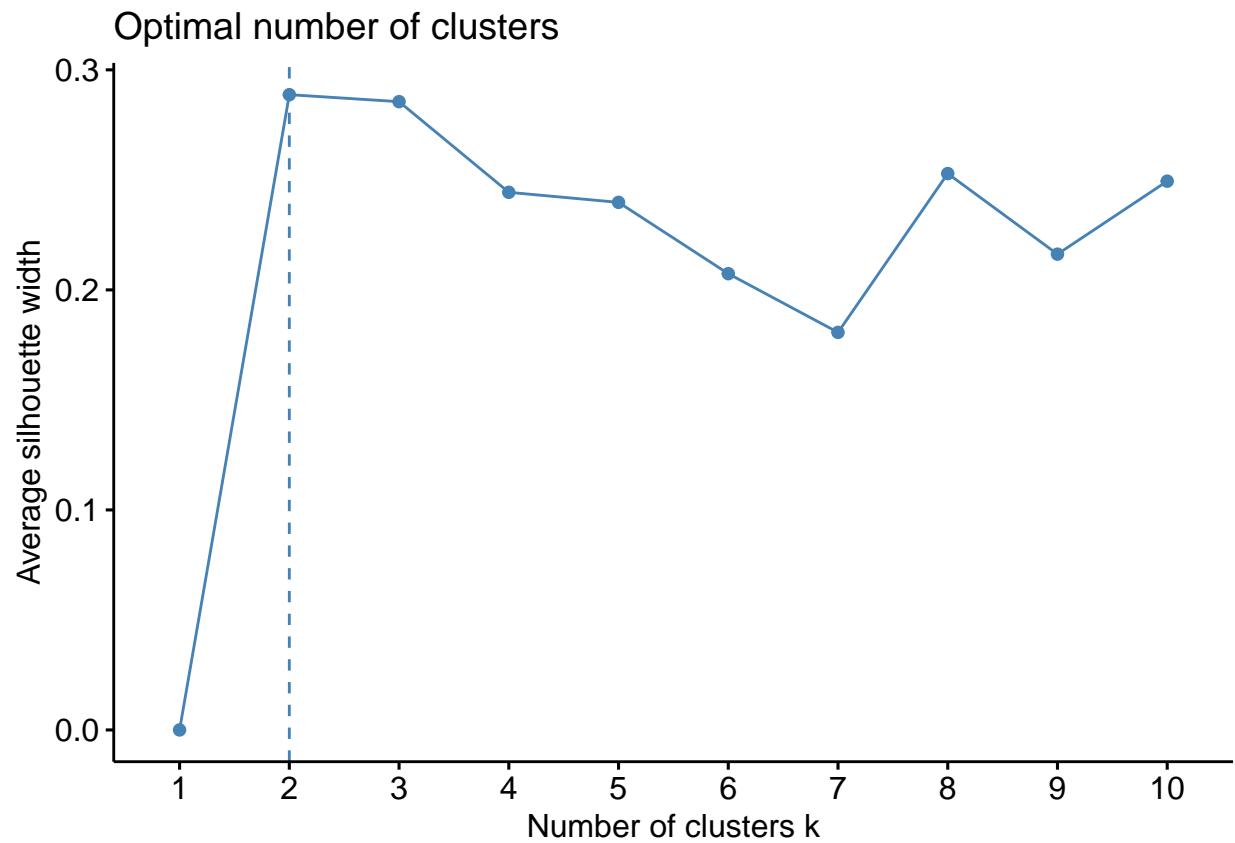
Determine the optimal number of clusters using the Elbow method

```
fviz_nbclust(predictions, kmeans, method = "wss")
```



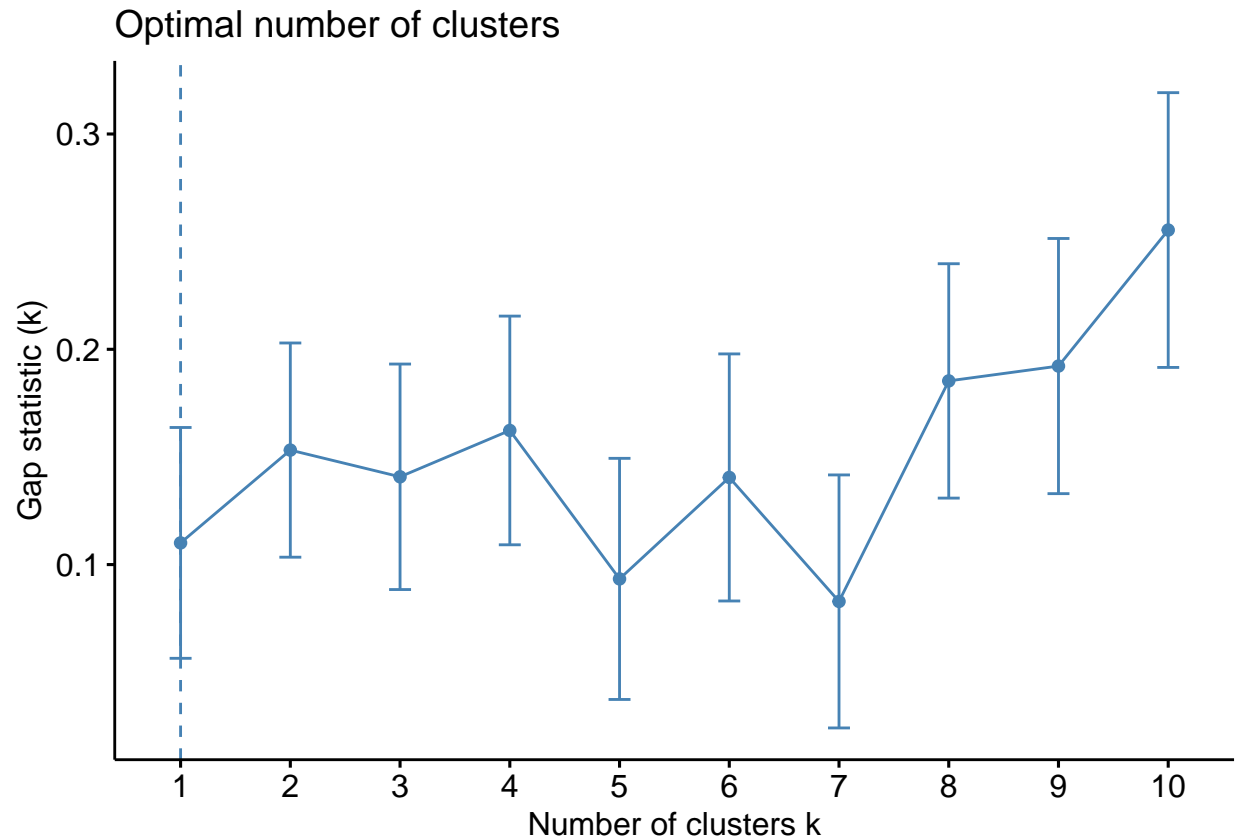
Determine the optimal number of clusters using the Silhouette method

```
fviz_nbclust(predictions, kmeans, method = "silhouette")
```



Determine the optimal number of clusters using the Gap static method

```
fviz_nbclust(predictions, kmeans, method = "gap_stat")
```



#From above we calculate kmeans optimal being k=5

```
k_means_5 <- kmeans(predictions, centers = 5, nstart = 25)
k_means_5$centers
```

```
##   Market_Cap      Beta  PE_Ratio      ROE      ROA Asset_Turnover  Leverage
## 1  0.11535885  0.2616487  0.6290663  0.1672316  0.2610229      0.6250000  0.11585945
## 2  0.03128035  0.7419355  0.2661597  0.2129944  0.1463845      0.3750000  0.47103514
## 3  0.78673516  0.3225806  0.2360583  0.6868644  0.8624339      0.8125000  0.06267806
## 4  0.31383933  0.2442396  0.2107550  0.4544794  0.6341648      0.5178571  0.10989011
## 5  0.06374962  0.4489247  0.1783904  0.1809322  0.2539683      0.1562500  0.18091168
##   Rev_Growth Net_Profit_Margin
## 1  0.3631175      0.2358079
## 2  0.2381844      0.1935953
## 3  0.5805912      0.7412664
## 4  0.2288084      0.7822832
## 5  0.8911851      0.5698690
```

```
k_means_5 <- kmeans(predictions, centers = 5, nstart = 25)
k_means_5$centers
```

```
##      Market_Cap      Beta  PE_Ratio      ROE      ROA Asset_Turnover  Leverage
## 1 0.78673516 0.3225806 0.2360583 0.6868644 0.8624339      0.8125000 0.06267806
## 2 0.03128035 0.7419355 0.2661597 0.2129944 0.1463845      0.3750000 0.47103514
## 3 0.31383933 0.2442396 0.2107550 0.4544794 0.6341648      0.5178571 0.10989011
## 4 0.06374962 0.4489247 0.1783904 0.1809322 0.2539683      0.1562500 0.18091168
## 5 0.11535885 0.2616487 0.6290663 0.1672316 0.2610229      0.6250000 0.11585945
##      Rev_Growth Net_Profit_Margin
## 1 0.5805912      0.7412664
## 2 0.2381844      0.1935953
## 3 0.2288084      0.7822832
## 4 0.8911851      0.5698690
## 5 0.3631175      0.2358079
```

Display the Pharmaceuticals data frame

```
Pharmaceuticals <- data.frame(predictions, cluster = k_means_5$cluster)
Pharmaceuticals
```

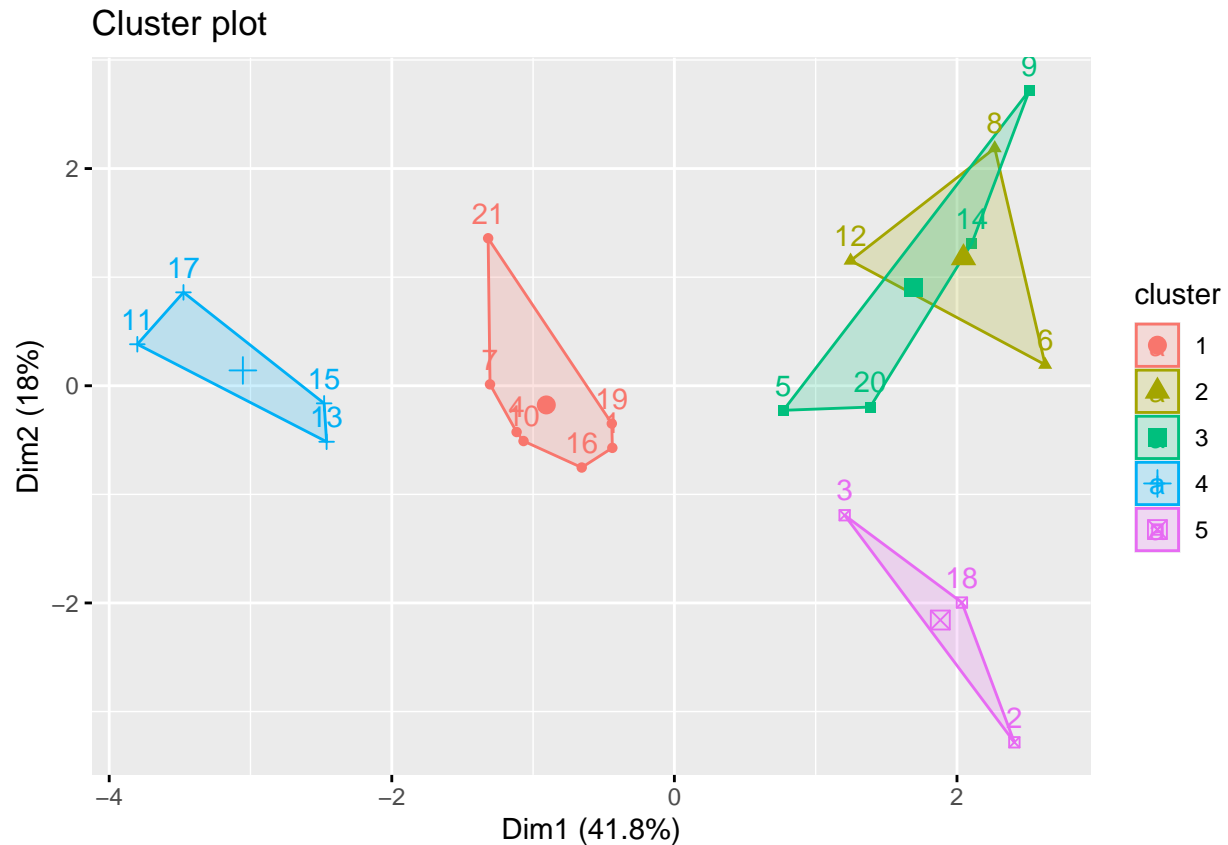
```
##      Market_Cap      Beta  PE_Ratio      ROE      ROA Asset_Turnover
## 1 0.341756254 0.15053763 0.2674271 0.3813559 0.5502646      0.500
## 2 0.036019291 0.24731183 1.0000000 0.1525424 0.2169312      0.750
## 3 0.029589069 0.30107527 0.2167300 0.1864407 0.3386243      0.750
## 4 0.337687130 0.36559140 0.2268695 0.3983051 0.7407407      0.750
## 5 0.234853813 0.15053763 0.2091255 0.3033898 0.3227513      0.375
## 6 0.082839345 1.00000000 0.3079848 0.0000000 0.0000000      0.375
## 7 0.255802271 0.34408602 0.1305450 0.5237288 0.7248677      0.750
## 8 0.000000000 0.72043011 0.2839037 0.3423729 0.1534392      0.375
## 9 0.001858736 0.96774194 0.0000000 0.1898305 0.1957672      0.000
## 10 0.368883754 0.00000000 0.3079848 0.4593220 0.6402116      0.375
## 11 0.611373455 0.18279570 0.1825095 1.0000000 1.0000000      0.875
## 12 0.011001708 0.50537634 0.2065906 0.2966102 0.2857143      0.375
## 13 0.871696976 0.30107527 0.3143219 0.4186441 0.7883598      0.750
## 14 0.003968653 0.61290323 0.3168568 0.1237288 0.2116402      0.000
## 15 0.663870190 0.30107527 0.1939163 0.6220339 0.7195767      1.000
## 16 0.483472320 0.01075269 0.2281369 0.2372881 0.5185185      0.250
## 17 1.000000000 0.50537634 0.2534854 0.7067797 0.9417989      0.625
## 18 0.280468201 0.23655914 0.6704689 0.1627119 0.2275132      0.375
## 19 0.169245454 0.35483871 0.1939163 0.3169492 0.6296296      0.625
## 20 0.014317291 0.06451613 0.1875792 0.1067797 0.2857143      0.250
## 21 0.240028132 0.48387097 0.1204056 0.8644068 0.6349206      0.375
##      Leverage Rev_Growth Net_Profit_Margin cluster
## 1 0.11965812 0.28651685      0.5895197      3
## 2 0.17094017 0.32985554      0.1266376      5
## 3 0.07692308 0.27340824      0.3755459      5
## 4 0.00000000 0.48608882      0.6724891      3
## 5 0.09686610 0.80203317      0.4497817      4
## 6 0.00000000 0.00000000      0.0000000      2
## 7 0.16239316 0.15703585      0.7860262      3
## 8 1.00000000 0.25548422      0.2139738      2
## 9 0.30484330 1.00000000      0.4672489      4
## 10 0.15099715 0.25093633      0.9082969      3
```

```
## 11 0.09686610 0.66987694      0.8078603      1
## 12 0.41310541 0.45906902      0.3668122      2
## 13 0.02849003 0.33547352      0.6681223      1
## 14 0.26495726 0.89727127      0.8165939      4
## 15 0.07977208 0.54895666      0.5021834      1
## 16 0.01709402 0.01284109      0.8646288      3
## 17 0.04558405 0.76805778      0.9868996      1
## 18 0.09971510 0.48608882      0.2052402      5
## 19 0.00000000 0.31380417      0.6550218      3
## 20 0.05698006 0.86543606      0.5458515      4
## 21 0.31908832 0.09443553      1.0000000      3
```

```
set.seed(1515)
k5 <- kmeans(Pharmaceuticals, centers = 5, nstart = 25)
k5$centers
```

```
##   Market_Cap      Beta PE_Ratio      ROE      ROA Asset_Turnover  Leverage
## 1 0.31383933 0.2442396 0.2107550 0.4544794 0.6341648    0.5178571 0.10989011
## 2 0.03128035 0.7419355 0.2661597 0.2129944 0.1463845    0.3750000 0.47103514
## 3 0.06374962 0.4489247 0.1783904 0.1809322 0.2539683    0.1562500 0.18091168
## 4 0.78673516 0.3225806 0.2360583 0.6868644 0.8624339    0.8125000 0.06267806
## 5 0.11535885 0.2616487 0.6290663 0.1672316 0.2610229    0.6250000 0.11585945
##   Rev_Growth Net_Profit_Margin cluster
## 1 0.2288084      0.7822832      3
## 2 0.2381844      0.1935953      2
## 3 0.8911851      0.5698690      4
## 4 0.5805912      0.7412664      1
## 5 0.3631175      0.2358079      5
```

```
fviz_cluster(k5, data = Pharmaceuticals)
```

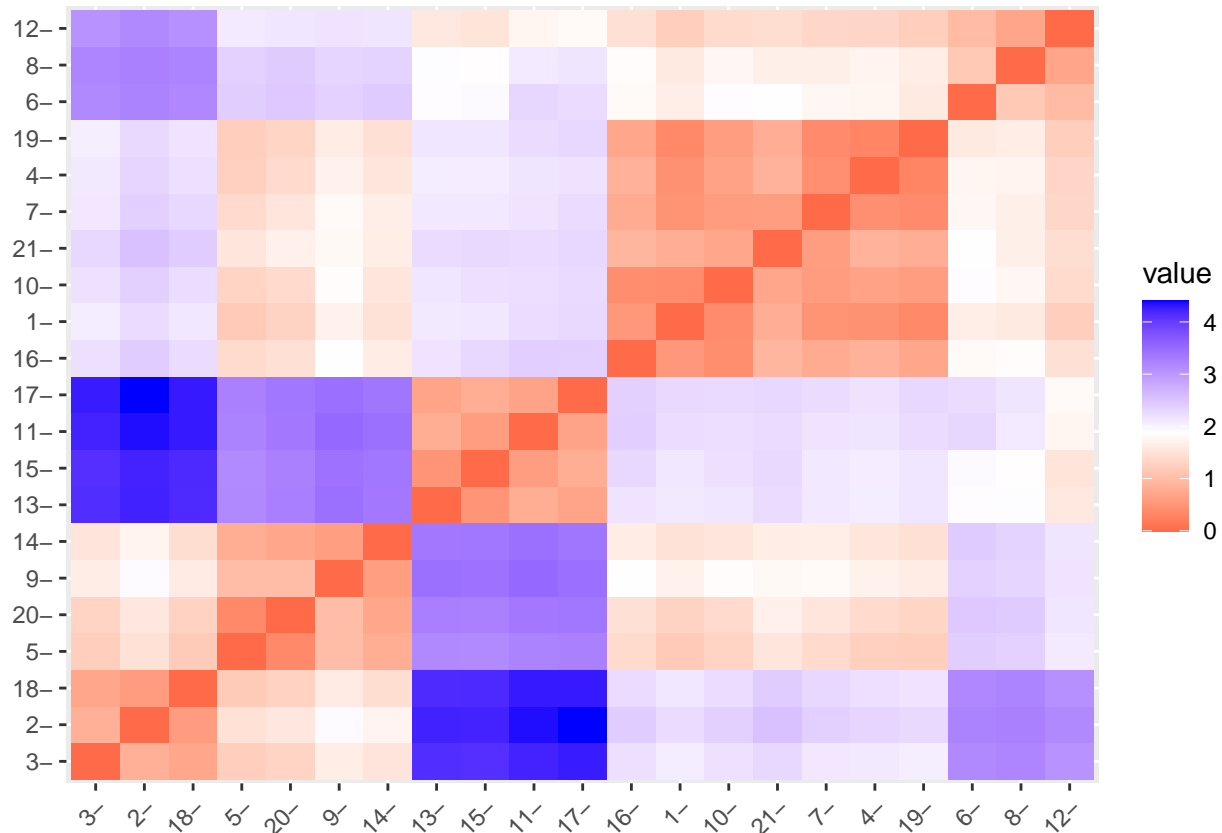



k5

```
## K-means clustering with 5 clusters of sizes 7, 3, 4, 4, 3
##
## Cluster means:
##   Market_Cap      Beta  PE_Ratio      ROE      ROA  Asset_Turnover  Leverage
## 1 0.31383933 0.2442396 0.2107550 0.4544794 0.6341648      0.5178571 0.10989011
## 2 0.03128035 0.7419355 0.2661597 0.2129944 0.1463845      0.3750000 0.47103514
## 3 0.06374962 0.4489247 0.1783904 0.1809322 0.2539683      0.1562500 0.18091168
## 4 0.78673516 0.3225806 0.2360583 0.6868644 0.8624339      0.8125000 0.06267806
## 5 0.11535885 0.2616487 0.6290663 0.1672316 0.2610229      0.6250000 0.11585945
##   Rev_Growth Net_Profit_Margin cluster
## 1 0.2288084      0.7822832      3
## 2 0.2381844      0.1935953      2
## 3 0.8911851      0.5698690      4
## 4 0.5805912      0.7412664      1
## 5 0.3631175      0.2358079      5
##
## Clustering vector:
## [1] 1 5 5 1 3 2 1 2 3 1 4 2 4 3 4 1 4 5 1 3 1
##
## Within cluster sum of squares by cluster:
## [1] 1.1915705 0.9214221 0.9159833 0.7017406 0.5175390
## (between_SS / total_SS = 91.1 %)
##
## Available components:
```

```
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"       "
```

```
Distance <- dist(Pharmaceuticals, method = "euclidian")
fviz_dist(Distance)
```



```
Fitting <- kmeans(Pharmaceuticals,3)
aggregate(Pharmaceuticals,by = list(Fitting$cluster), FUN = mean)
```

```
##   Group.1 Market_Cap      Beta PE_Ratio      ROE      ROA Asset_Turnover
## 1      1  0.78673516 0.3225806 0.2360583 0.6868644 0.8624339 0.8125000
## 2      2  0.08586786 0.3686636 0.3715372 0.1750605 0.2569917 0.3571429
## 3      3  0.22907164 0.3935484 0.2273764 0.3820339 0.4878307 0.4750000
##   Leverage Rev_Growth Net_Profit_Margin cluster
## 1 0.06267806 0.5805912      0.7412664 1.000000
## 2 0.15303215 0.6648704      0.4266999 4.428571
## 3 0.21823362 0.2316212      0.6056769 2.700000
```

```
Pharma <- data.frame(Pharmaceuticals,Fitting$cluster)
Pharma
```

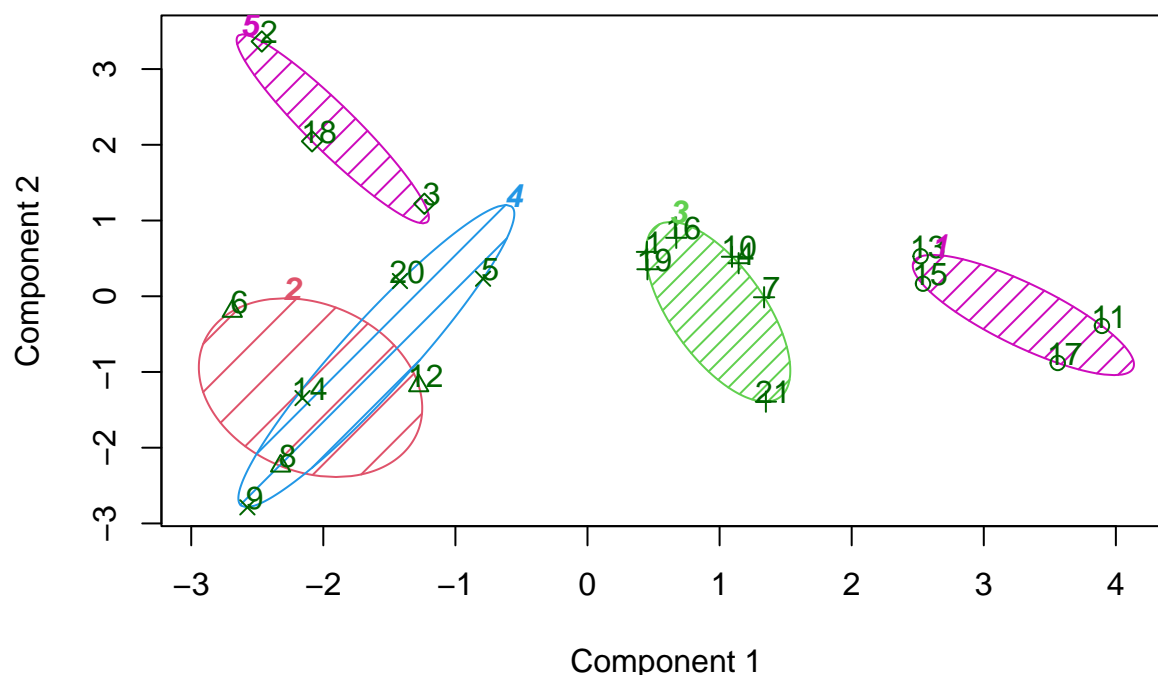
```
##   Market_Cap      Beta PE_Ratio      ROE      ROA Asset_Turnover
## 1 0.341756254 0.15053763 0.2674271 0.3813559 0.5502646 0.500
```

## 2	0.036019291	0.24731183	1.0000000	0.1525424	0.2169312	0.750
## 3	0.029589069	0.30107527	0.2167300	0.1864407	0.3386243	0.750
## 4	0.337687130	0.36559140	0.2268695	0.3983051	0.7407407	0.750
## 5	0.234853813	0.15053763	0.2091255	0.3033898	0.3227513	0.375
## 6	0.082839345	1.00000000	0.3079848	0.0000000	0.0000000	0.375
## 7	0.255802271	0.34408602	0.1305450	0.5237288	0.7248677	0.750
## 8	0.000000000	0.72043011	0.2839037	0.3423729	0.1534392	0.375
## 9	0.001858736	0.96774194	0.0000000	0.1898305	0.1957672	0.000
## 10	0.368883754	0.00000000	0.3079848	0.4593220	0.6402116	0.375
## 11	0.611373455	0.18279570	0.1825095	1.0000000	1.0000000	0.875
## 12	0.011001708	0.50537634	0.2065906	0.2966102	0.2857143	0.375
## 13	0.871696976	0.30107527	0.3143219	0.4186441	0.7883598	0.750
## 14	0.003968653	0.61290323	0.3168568	0.1237288	0.2116402	0.000
## 15	0.663870190	0.30107527	0.1939163	0.6220339	0.7195767	1.000
## 16	0.483472320	0.01075269	0.2281369	0.2372881	0.5185185	0.250
## 17	1.000000000	0.50537634	0.2534854	0.7067797	0.9417989	0.625
## 18	0.280468201	0.23655914	0.6704689	0.1627119	0.2275132	0.375
## 19	0.169245454	0.35483871	0.1939163	0.3169492	0.6296296	0.625
## 20	0.014317291	0.06451613	0.1875792	0.1067797	0.2857143	0.250
## 21	0.240028132	0.48387097	0.1204056	0.8644068	0.6349206	0.375
##	Leverage	Rev_Growth	Net_Profit_Margin	cluster	Fitting.cluster	
## 1	0.11965812	0.28651685	0.5895197	3	3	
## 2	0.17094017	0.32985554	0.1266376	5	2	
## 3	0.07692308	0.27340824	0.3755459	5	2	
## 4	0.00000000	0.48608882	0.6724891	3	3	
## 5	0.09686610	0.80203317	0.4497817	4	2	
## 6	0.00000000	0.00000000	0.0000000	2	3	
## 7	0.16239316	0.15703585	0.7860262	3	3	
## 8	1.00000000	0.25548422	0.2139738	2	3	
## 9	0.30484330	1.00000000	0.4672489	4	2	
## 10	0.15099715	0.25093633	0.9082969	3	3	
## 11	0.09686610	0.66987694	0.8078603	1	1	
## 12	0.41310541	0.45906902	0.3668122	2	3	
## 13	0.02849003	0.33547352	0.6681223	1	1	
## 14	0.26495726	0.89727127	0.8165939	4	2	
## 15	0.07977208	0.54895666	0.5021834	1	1	
## 16	0.01709402	0.01284109	0.8646288	3	3	
## 17	0.04558405	0.76805778	0.9868996	1	1	
## 18	0.09971510	0.48608882	0.2052402	5	2	
## 19	0.00000000	0.31380417	0.6550218	3	3	
## 20	0.05698006	0.86543606	0.5458515	4	2	
## 21	0.31908832	0.09443553	1.0000000	3	3	

Create a cluster plot

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

CLUSPLOT(Pharmaceuticals)



These two components explain 59.81 % of the point variability.

#b. Interpret the clusters with respect to the numerical variables used in forming the clusters. # Combine cluster assignments with Location and Exchange columns

```
aggregate(Pharmaceuticals, by = list(Fitting$cluster), FUN = mean)
```

```
##      Group.1 Market_Cap      Beta PE_Ratio      ROE      ROA Asset_Turnover
## 1          1  0.78673516 0.3225806 0.2360583 0.6868644 0.8624339      0.8125000
## 2          2  0.08586786 0.3686636 0.3715372 0.1750605 0.2569917      0.3571429
## 3          3  0.22907164 0.3935484 0.2273764 0.3820339 0.4878307      0.4750000
##      Leverage Rev_Growth Net_Profit_Margin cluster
## 1 0.06267806  0.5805912      0.7412664 1.000000
## 2 0.15303215  0.6648704      0.4266999 4.428571
## 3 0.21823362  0.2316212      0.6056769 2.700000
```

```
Pharmacy <- data.frame(Pharmaceuticals,k5$cluster)
Pharmacy
```

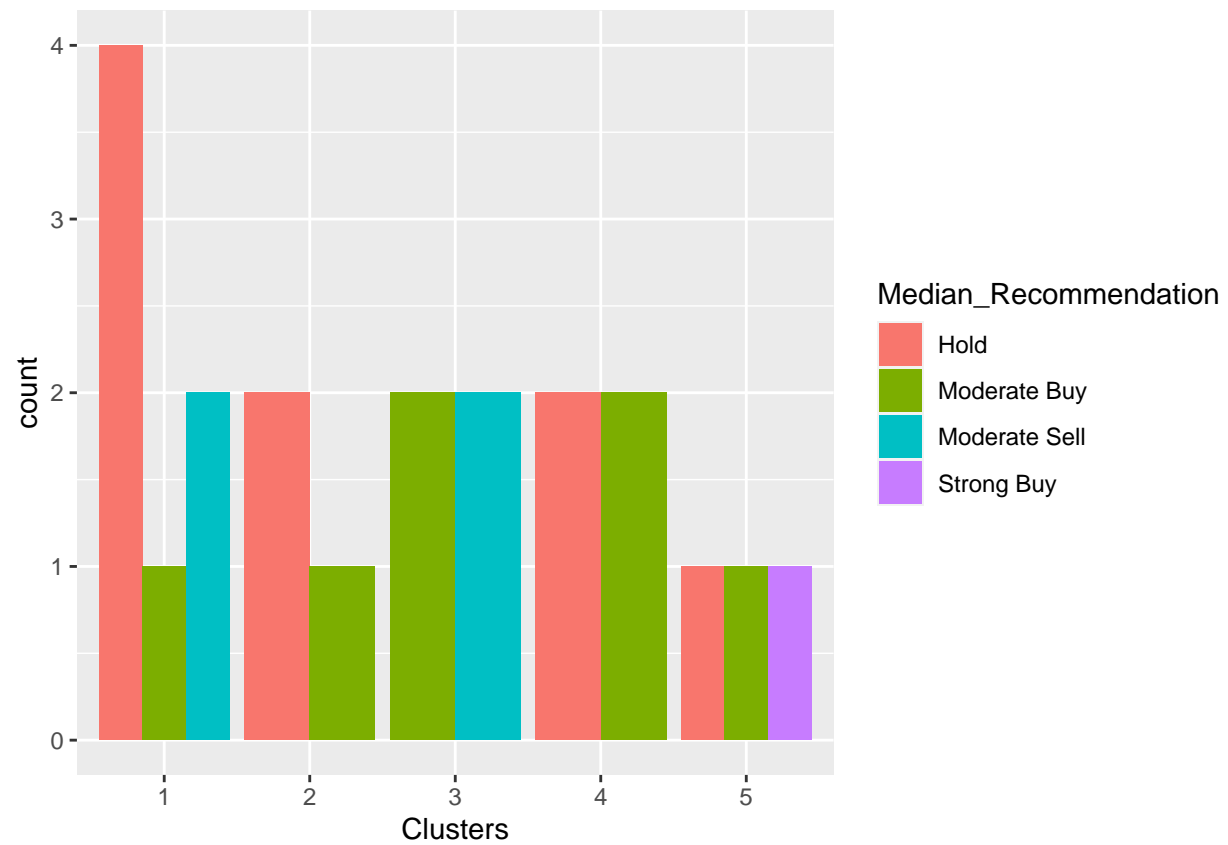
```
##      Market_Cap      Beta PE_Ratio      ROE      ROA Asset_Turnover
## 1 0.341756254 0.15053763 0.2674271 0.3813559 0.5502646      0.500
## 2 0.036019291 0.24731183 1.0000000 0.1525424 0.2169312      0.750
## 3 0.029589069 0.30107527 0.2167300 0.1864407 0.3386243      0.750
## 4 0.337687130 0.36559140 0.2268695 0.3983051 0.7407407      0.750
## 5 0.234853813 0.15053763 0.2091255 0.3033898 0.3227513      0.375
## 6 0.082839345 1.00000000 0.3079848 0.0000000 0.0000000      0.375
## 7 0.255802271 0.34408602 0.1305450 0.5237288 0.7248677      0.750
```

## 8	0.000000000	0.72043011	0.2839037	0.3423729	0.1534392	0.375
## 9	0.001858736	0.96774194	0.0000000	0.1898305	0.1957672	0.000
## 10	0.368883754	0.00000000	0.3079848	0.4593220	0.6402116	0.375
## 11	0.611373455	0.18279570	0.1825095	1.0000000	1.0000000	0.875
## 12	0.011001708	0.50537634	0.2065906	0.2966102	0.2857143	0.375
## 13	0.871696976	0.30107527	0.3143219	0.4186441	0.7883598	0.750
## 14	0.003968653	0.61290323	0.3168568	0.1237288	0.2116402	0.000
## 15	0.663870190	0.30107527	0.1939163	0.6220339	0.7195767	1.000
## 16	0.483472320	0.01075269	0.2281369	0.2372881	0.5185185	0.250
## 17	1.000000000	0.50537634	0.2534854	0.7067797	0.9417989	0.625
## 18	0.280468201	0.23655914	0.6704689	0.1627119	0.2275132	0.375
## 19	0.169245454	0.35483871	0.1939163	0.3169492	0.6296296	0.625
## 20	0.014317291	0.06451613	0.1875792	0.1067797	0.2857143	0.250
## 21	0.240028132	0.48387097	0.1204056	0.8644068	0.6349206	0.375
##	Leverage	Rev_Growth	Net_Profit_Margin	cluster	k5.cluster	
## 1	0.11965812	0.28651685	0.5895197	3	1	
## 2	0.17094017	0.32985554	0.1266376	5	5	
## 3	0.07692308	0.27340824	0.3755459	5	5	
## 4	0.00000000	0.48608882	0.6724891	3	1	
## 5	0.09686610	0.80203317	0.4497817	4	3	
## 6	0.00000000	0.00000000	0.0000000	2	2	
## 7	0.16239316	0.15703585	0.7860262	3	1	
## 8	1.00000000	0.25548422	0.2139738	2	2	
## 9	0.30484330	1.00000000	0.4672489	4	3	
## 10	0.15099715	0.25093633	0.9082969	3	1	
## 11	0.09686610	0.66987694	0.8078603	1	4	
## 12	0.41310541	0.45906902	0.3668122	2	2	
## 13	0.02849003	0.33547352	0.6681223	1	4	
## 14	0.26495726	0.89727127	0.8165939	4	3	
## 15	0.07977208	0.54895666	0.5021834	1	4	
## 16	0.01709402	0.01284109	0.8646288	3	1	
## 17	0.04558405	0.76805778	0.9868996	1	4	
## 18	0.09971510	0.48608882	0.2052402	5	5	
## 19	0.00000000	0.31380417	0.6550218	3	1	
## 20	0.05698006	0.86543606	0.5458515	4	3	
## 21	0.31908832	0.09443553	1.0000000	3	1	

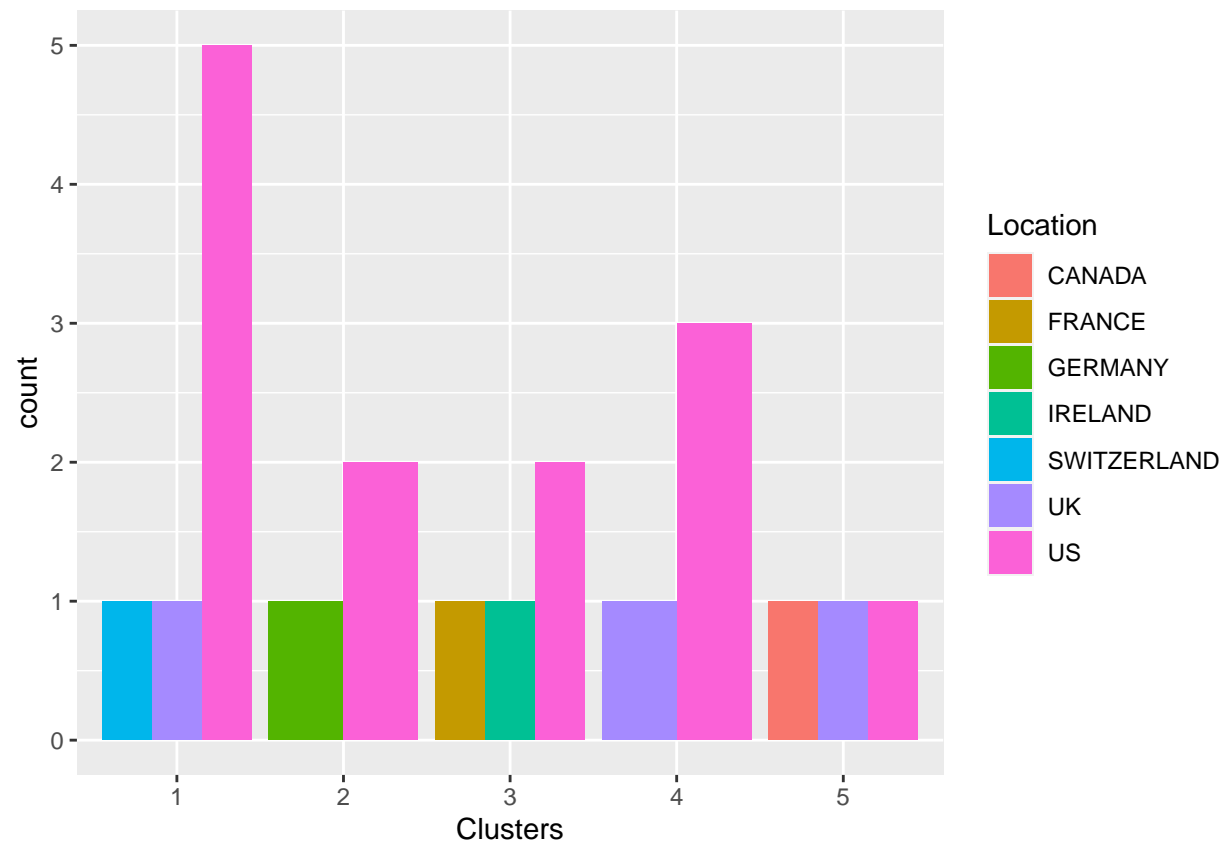
c. Is there a pattern in the clusters with respect to the numerical variables (10 to 12)

Create a bar plot using ggplot(fill=Location)

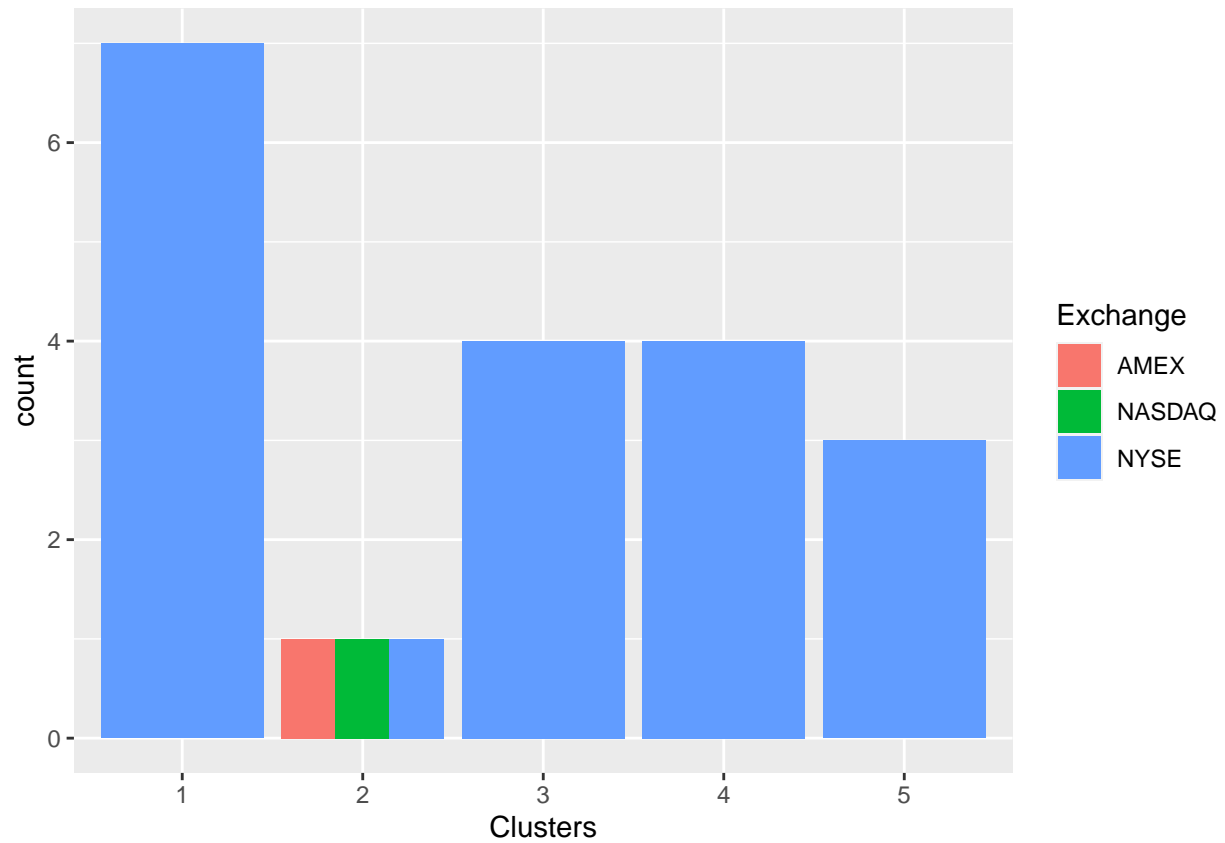
```
Assignment4 <- Pharma_Assignment4[12:14] %>% mutate(Clusters=k5$cluster)
ggplot(Assignment4, mapping = aes(factor(Clusters), fill=Median_Recommendation))+geom_bar(position='dodge')
```



```
ggplot(Assignment4, mapping = aes(factor(Clusters), fill = Location)) + geom_bar(position = 'dodge') + labs(x = 'Clusters', y = 'count')
```



```
ggplot(Assignment4, mapping = aes(factor(Clusters),fill = Exchange))+geom_bar(position = 'dodge')+labs(
```



#d. Provide an appropriate name for each cluster using any or all of the variables in the dataset. Cluster 1 represents firms with strong buy potential, Cluster 2 consists of firms with uncertain investment potential, and Cluster 3 comprises firms with moderate buy potential.