MichaelBasta_Assignment_4

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2022-10-30

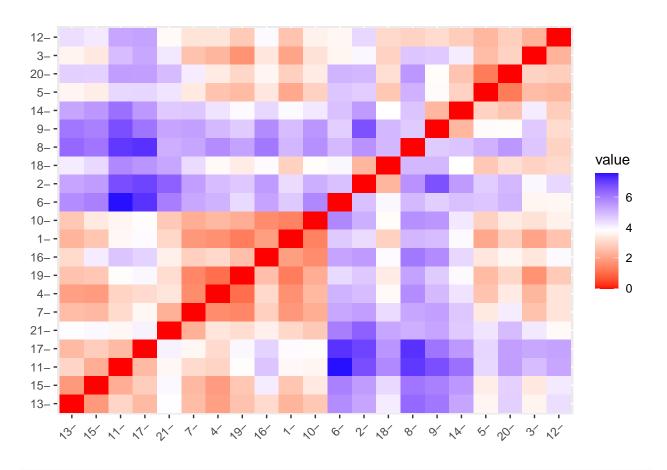
```
Pharma <- read.csv("/Users/michaelbasta/Documents/Fundmentals of Machine Learning /Module 6/Pharmaceuti
library(tidyverse)
## -- Attaching packages -----
                                            ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6 v purrr
                               0.3.5
## v tibble 3.1.8
                    v dplyr 1.0.10
## v tidyr 1.2.1
                    v stringr 1.4.1
## v readr 2.1.3
                     v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library(factoextra)
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
library(ISLR)
#install.packages("flexclust")
library(flexclust)
## Loading required package: grid
## Loading required package: lattice
## Loading required package: modeltools
## Loading required package: stats4
set.seed(123)
# a.
df <- Pharma[,3:11]</pre>
summary(df)
                                                      ROE
##
     Market_Cap
                                     PE_Ratio
                       Beta
## Min. : 0.41 Min. :0.1800 Min. : 3.60 Min. : 3.9
## 1st Qu.: 6.30 1st Qu.:0.3500 1st Qu.:18.90 1st Qu.:14.9
## Median: 48.19 Median: 0.4600 Median: 21.50 Median: 22.6
```

Mean : 57.65 Mean :0.5257 Mean :25.46 Mean :25.8 ## 3rd Qu.: 73.84 3rd Qu.:0.6500 3rd Qu.:27.90 3rd Qu.:31.0 ## Max. :199.47 Max. :1.1100 Max. :82.50 Max. :62.9

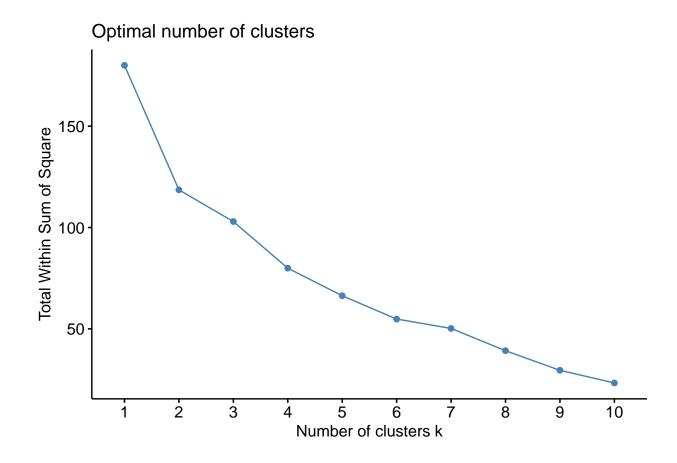
```
Asset_Turnover
##
        ROA
                                  Leverage
                                                 Rev_Growth
## Min. : 1.40
                 Min. :0.3
                               Min.
                                      :0.0000
                                               Min. :-3.17
                1st Qu.:0.6
  1st Qu.: 5.70
                               1st Qu.:0.1600
                                               1st Qu.: 6.38
## Median :11.20 Median :0.6
                               Median :0.3400
                                               Median: 9.37
## Mean :10.51
                  Mean :0.7
                               Mean :0.5857
                                               Mean :13.37
## 3rd Qu.:15.00
                  3rd Qu.:0.9
                               3rd Qu.:0.6000
                                               3rd Qu.:21.87
## Max.
        :20.30
                  Max. :1.1
                               Max. :3.5100
                                               Max. :34.21
## Net_Profit_Margin
## Min. : 2.6
## 1st Qu.:11.2
## Median :16.1
## Mean :15.7
## 3rd Qu.:21.1
## Max. :25.5
```

It doesn't seem that there's a direct correlation between variables so Euclidean distance should be s # Also since it is scale dependant we had to scale it before applying df <- scale(df)</pre> distance <- get_dist(df)</pre>

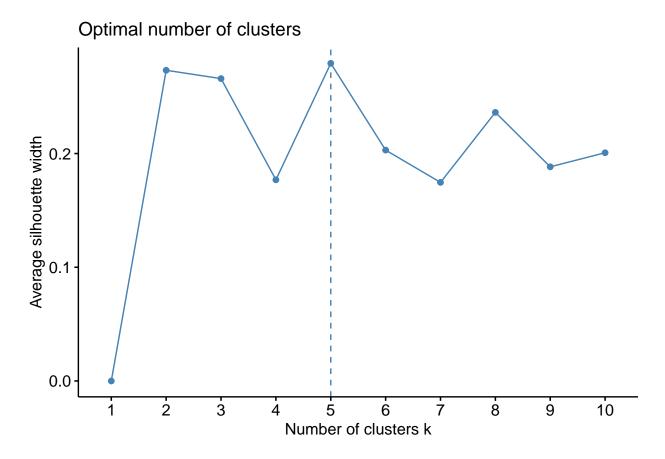
fviz_dist(distance)



fviz_nbclust(df, kmeans, method="wss")

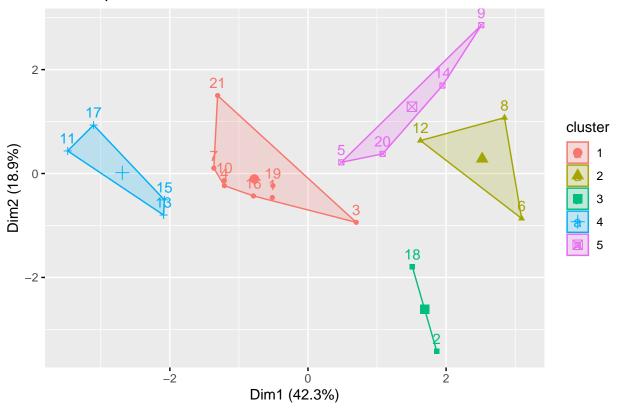


fviz_nbclust(df, kmeans, method="silhouette")



From the earlier step it shows that 5 clusters is the best choice
k4 <- kmeans(df, centers=5, nstart=25)
fviz_cluster(k4, data = df)</pre>

Cluster plot

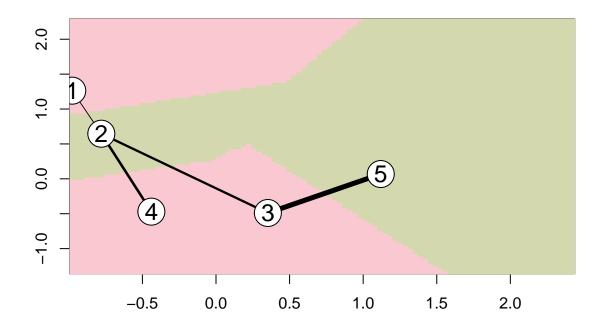


k4\$size

[1] 8 3 2 4 4

k4\$centers

```
##
     Market_Cap
                            PE_Ratio
                                           ROE
                                                     ROA Asset_Turnover
                     Beta
## 1 -0.03142211 -0.4360989 -0.31724852 0.1950459 0.4083915
                                                             0.1729746
## 2 -0.87051511 1.3409869 -0.05284434 -0.6184015 -1.1928478
                                                            -0.4612656
## 3 -0.43925134 -0.4701800 2.70002464 -0.8349525 -0.9234951
                                                             0.2306328
## 4 1.69558112 -0.1780563 -0.19845823 1.2349879 1.3503431
                                                             1.1531640
-1.2684804
       Leverage Rev_Growth Net_Profit_Margin
##
## 1 -0.27449312 -0.7041516
                               0.556954446
## 2 1.36644699 -0.6912914
                              -1.320000179
## 3 -0.14170336 -0.1168459
                              -1.416514761
## 4 -0.46807818 0.4671788
                               0.591242521
## 5 0.06308085 1.5180158
                              -0.006893899
k4 = kcca(df, k=5, kccaFamily("kmeans"))
clusters_index = predict(k4)
image(k4)
```



```
## b.

## Looking at the data from the centroids of the clusters

## Cluster 1: (Market_cap, Beta, PE_Ratio, Leverage, Rev_Growth) are Lower than average (negative value ## Cluster 2: (Market_cap, PE_Ratio, ROE, ROA, Asset_Turnover, Rev_Growth, Net_Profit_Margin) are Lower ## Cluster 3: (Market_cap, Beta, ROE, ROA, Leverage, Rev_Growth) are Lower than average (negative values ## Cluster 4: (Beta, PE_Ratio, Leverage) are Lower than average (negative values) and (Market_Cap, ROE, ## Cluster 5: (Market_Cap, PE_Ratio, ROE, ROA, Asset_Turnover, Net_Profit_Margin) are Lower than average ## c.

## I don't see a pattern in the clusters with respect to the non-numerical variables

## d.

## Cluster 1: High returns and turnover

## Cluster 2: High leverage

## Cluster 3: High Price/earnings ratio

## Cluster 4: High Market Capitalization and returns

## Cluster 5: High Revenew Growth
```