HackBio Bio-Data Science Task 3

In this stage, I will continue to improve on my R programming skills. I have two tasks. First, I will perform K-means clustering and Hierachical clustering on 'mtcars built in data set in R' and secondly, the biological data set 'microbial stationary phase", that I performed pca on last week.

Consider the following explanations and plots.

1. Data set for this first analysis is 'mtcars'

K-means clustering

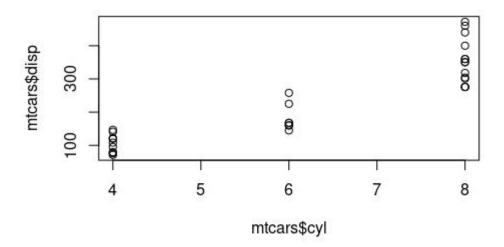
2. Our data is mtcars, an inbuilt data in R data()

data(mtcars)

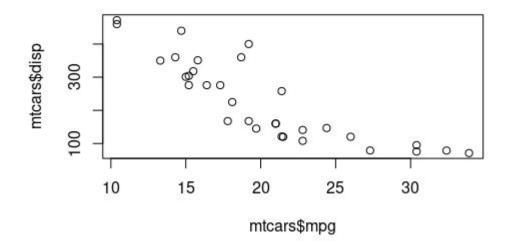
View(mtcars)

3. For the purpose of visualization, I will like to use any two columns that correlate cor(mtcars)

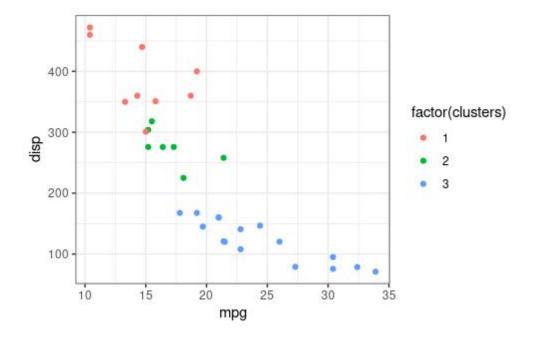
plot (mtcars\$cyl, mtcars\$disp)



plot (mtcars\$mpg, mtcars\$disp)

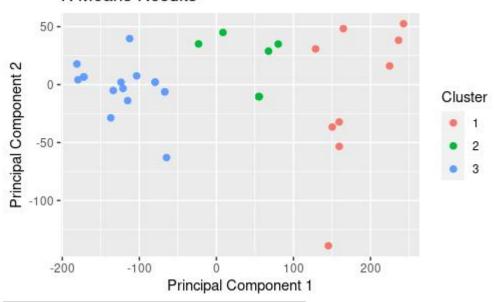


- 4. So we can have same results. I will set seed which is an arbitrary number. Since I want to get a consistent result from my K-Mean clustering (which is unsupervised). Picking '102' means that I have chosen to index to 102. set.seed(102)
- 5. I will now perform my KMEAN Clustering for now. I want to pick 3 centers (clusters) mtcarsK3 <- kmeans (x = mtcars, centers = 3)
- 6. Well, I can also add this cluster information to my dataset mtcars\$clusters <- c(mtcarsK3\$cluster)
- 7. I want to visualize this cluster information for each car ggplot(mtcars, aes(x = mpg, y = disp, color = factor(clusters))) + geom_point() + theme_bw()

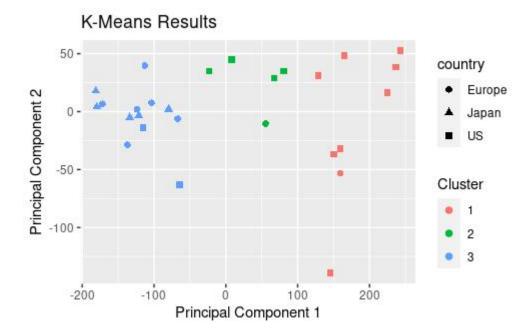


- 8. I need to install "useful" for my K-Means Clustering install.packages('useful') library ('useful')
- 9. I will create a new column for country mtcars.country <- c(rep("Japan", 3), rep("US", 4), rep("Europe", 7), rep("US", 3), "Europe", rep("Japan", 3), rep("US", 4), rep("Europe", 4), rep("US", 3)) mtcars\$country <- c(mtcars.country)
- 10. Let's now plot plot(mtcarsK3, data = mtcars)

K-Means Results



plot(mtcarsK3, data = mtcars, class = 'country')



11. Choosing the right number of clusters: I will set my maximum cluster to 10.

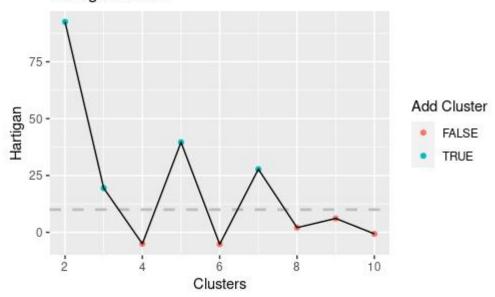
NB: I need to re-run my data set since the columns of 'country' and 'cluster' which I previously added will not all our code to run because it doesn't require alphabets.

mtcarsBEST <- FitKMeans(mtcars, max.clusters=10)

mtcarsBEST

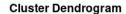
NB: Once I got a FALSE, I stopped counting for clusters PlotHartigan(mtcarsBEST)

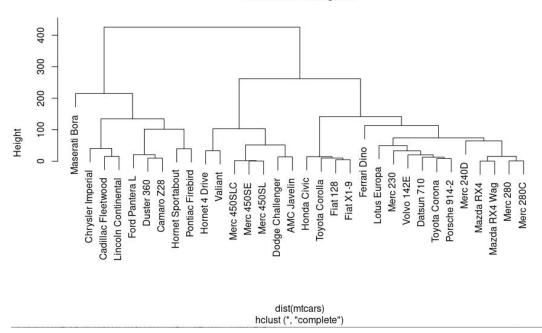
Hartigan's Rule



- 12. I will now plot again, this time using the number of clusters that I determined. Since I determined 3 clusters just like previously, there is no need to repeat all over.
- 13. Hierachical Clustering is used to cluster clusters into clusters. Let's see how to implement it: I will start by calculating the distance between the rows. Next, I will pick a method. In this case I will be using the "complete method". "hclust and dist" are inbuilt functions in R.

hcmtcars <- hclust(d=dist(mtcars), method='complete') plot(hcmtcars)





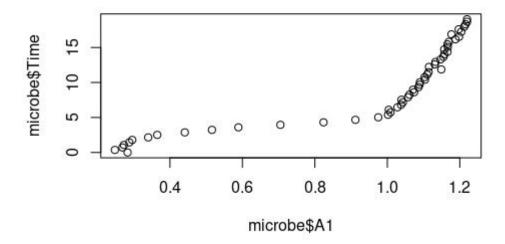
Second K-means clustering

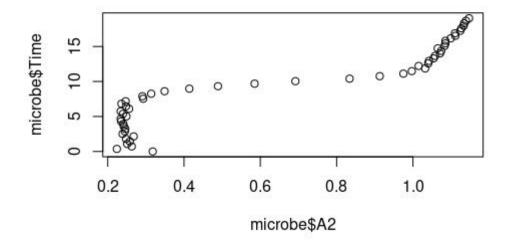
14. Our second data set is 'microbial_stationary_phase.csv' from last week's task. Import .csv file in R

microbe <- read.csv(file.choose())
microbe

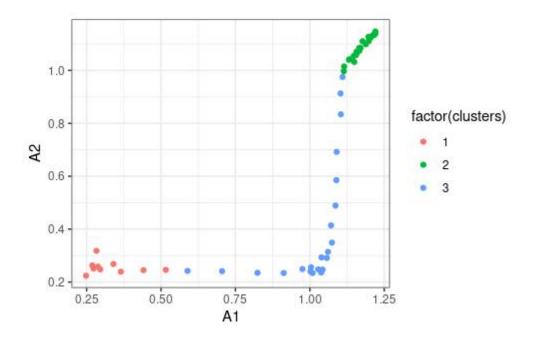
15. For the purpose of visualization, I will like to use any two columns that correlate cor(microbe)

plot (microbe\$A1, microbe\$Time)





- 16. So we can have same results. I will set seed which is an arbitrary number. Since I want to get a consistent result from my K-Mean clustering (which is unsupervised). Picking '102' means that I have chosen to index to 102. set.seed(102)
- 17. I will now perform my KMEAN Clustering for now. I want to pick 3 centers (clusters) microbeK3 <- kmeans (x = microbe, centers = 3)
- 18. Well, I can also add this cluster information to my dataset microbe\$clusters <- c(microbeK3\$cluster)
- 19. I want to visualize this cluster information for each microbe ggplot(microbe, aes(x = A1, y = A2, color = factor(clusters))) + geom_point() + theme bw()



20. Install useful install.packages('useful') library ('useful')

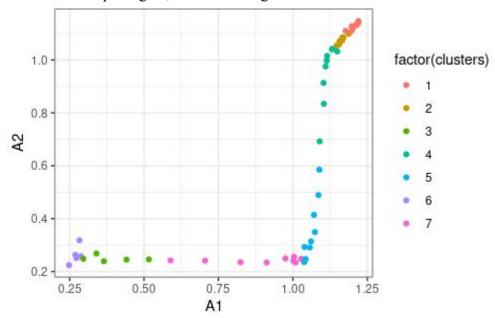
21. Let's plot plot(microbeK3, data = microbe)

K-Means Results 1.0 - Cluster 0.0 - 1 2 3 Principal Component 1

- 22. Choosing the right number of clusters: I will set my maximum cluster to 10. microbeBEST <- FitKMeans(microbe, max.clusters=10)
- 23. Once I got a FALSE, I stopped counting for clusters PlotHartigan(microbeBEST)

Hartigan's Rule 150 100 Add Cluster FALSE TRUE Clusters

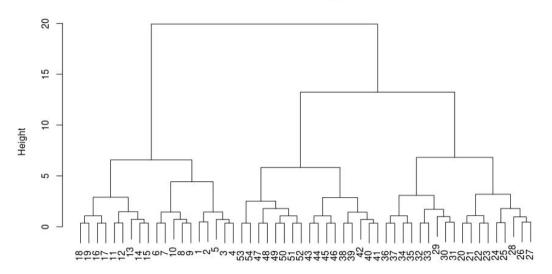
24. I will now plot again, this time using the number of clusters that I determined



25. Hierachical Clustering is used to cluster clusters into clusters. Let's see how to implement it: I will start by calculating the distance between the rows. Next, I will pick a method. In this case I will be using the "complete method". "hclust and dist" are inbuilt functions in R.

hcmicrobe <- hclust(d=dist(microbe), method='complete')
plot(hcmicrobe)</pre>

Cluster Dendrogram



dist(microbe) hclust (*, "complete")