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Vellore Institute of Technology
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Essentials of Data Analytics - (CSE3506)

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Lab-7

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Tasks for Week-7: Partitioning Based Clustering

Understand the following operations/functions on 'iris' data and perform similar operations on 'USArrests' dataset based on given instructions.

AIM

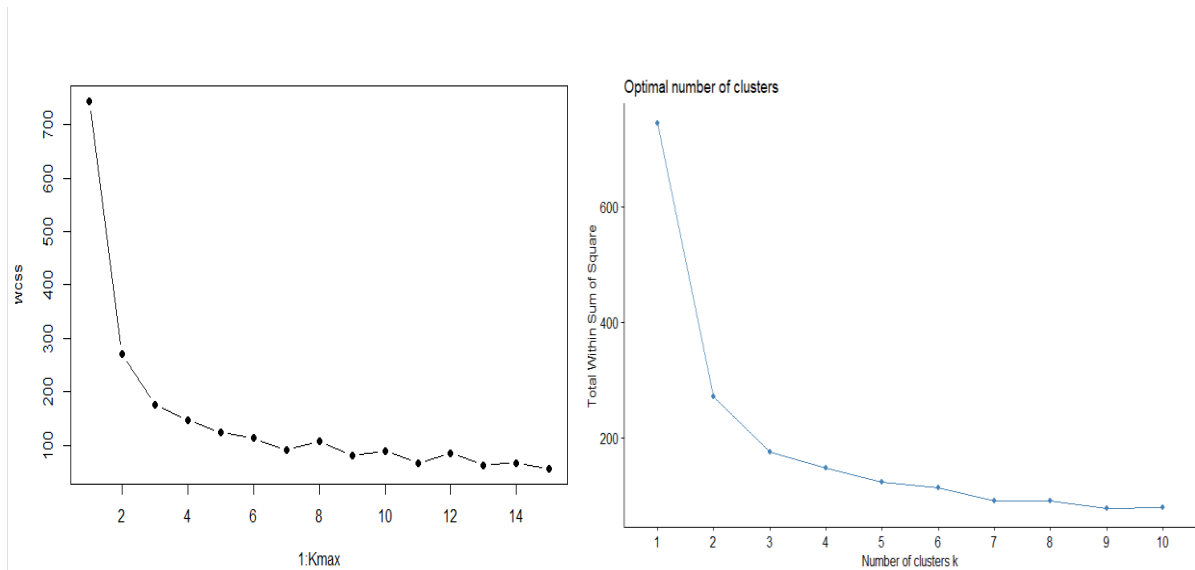
To Understand and perform operations/functions on 'iris' data and perform similar operations on 'USArrests' dataset based on given instructions.

Algorithm

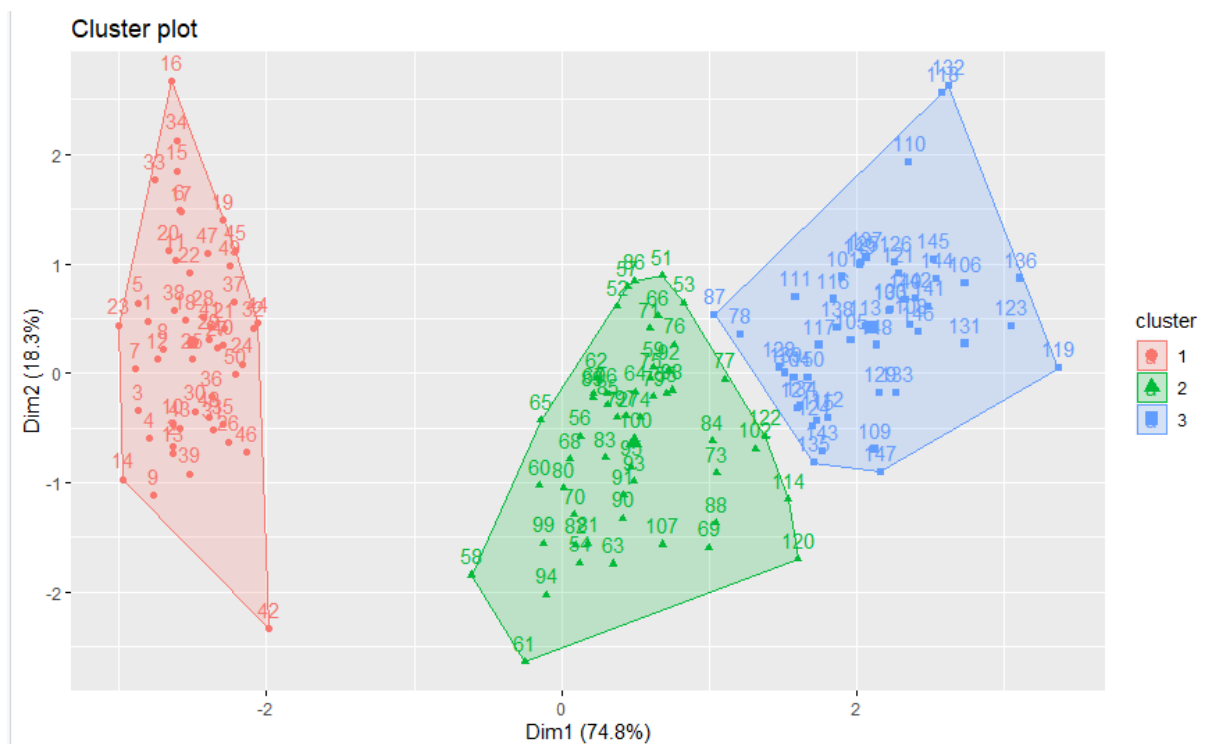
1. Start
2. Use `rm(list=ls())` to clear environment variables.
3. Read the dataset from directory.
4. Make a scaled version of the original data.
5. Train the Kmeans model and go over all of its terms.
6. By using `fit$cluster` we can find the cluster values.
7. By using `fit$size` we can find the size of each cluster.
8. By using `fit$withinss` we can find within cluster sum of squares for each cluster.
9. By using `fit$tot.withinss` we can find within cluster sum of squares with respective to all clusters.
10. Find WCSS values and keep a list of them.
11. To find the elbow point, plot the graph.
12. Train the K-Medoids model and go over all of its terms.
13. By using the `fitm$medoid` we can find no of medoid
14. Using `fviz_cluster` function we can find the medoids
15. Stop.

Result

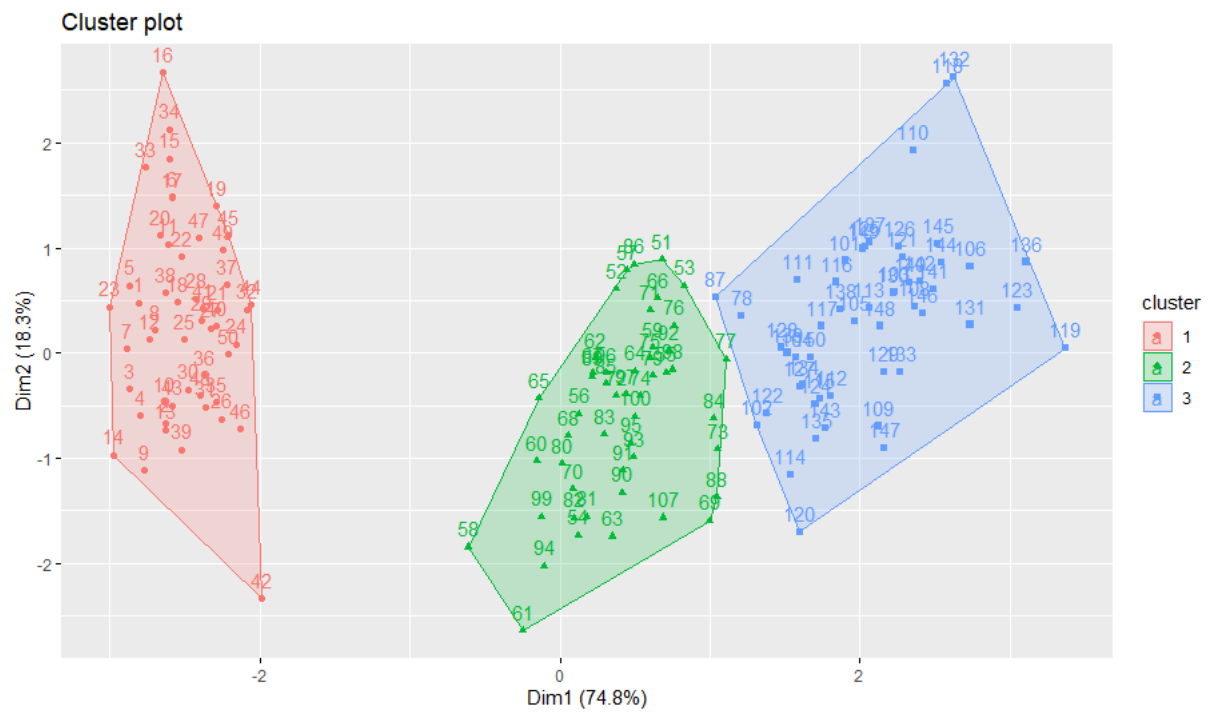
Case 1: Iris dataset



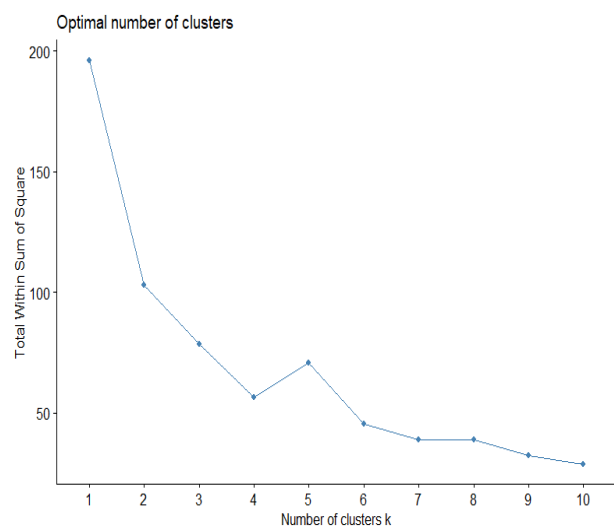
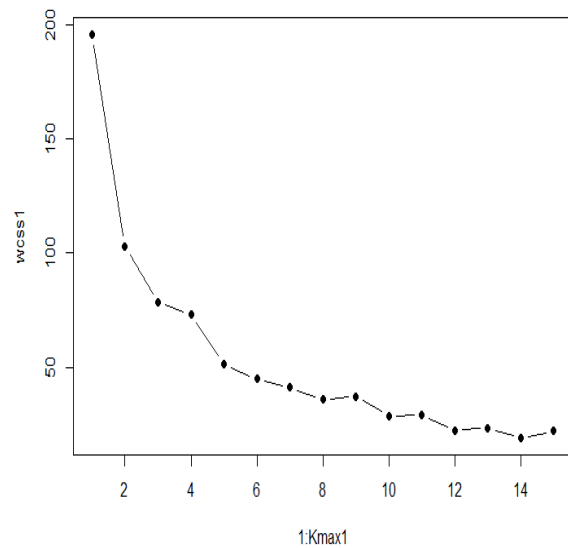
K-means centers:



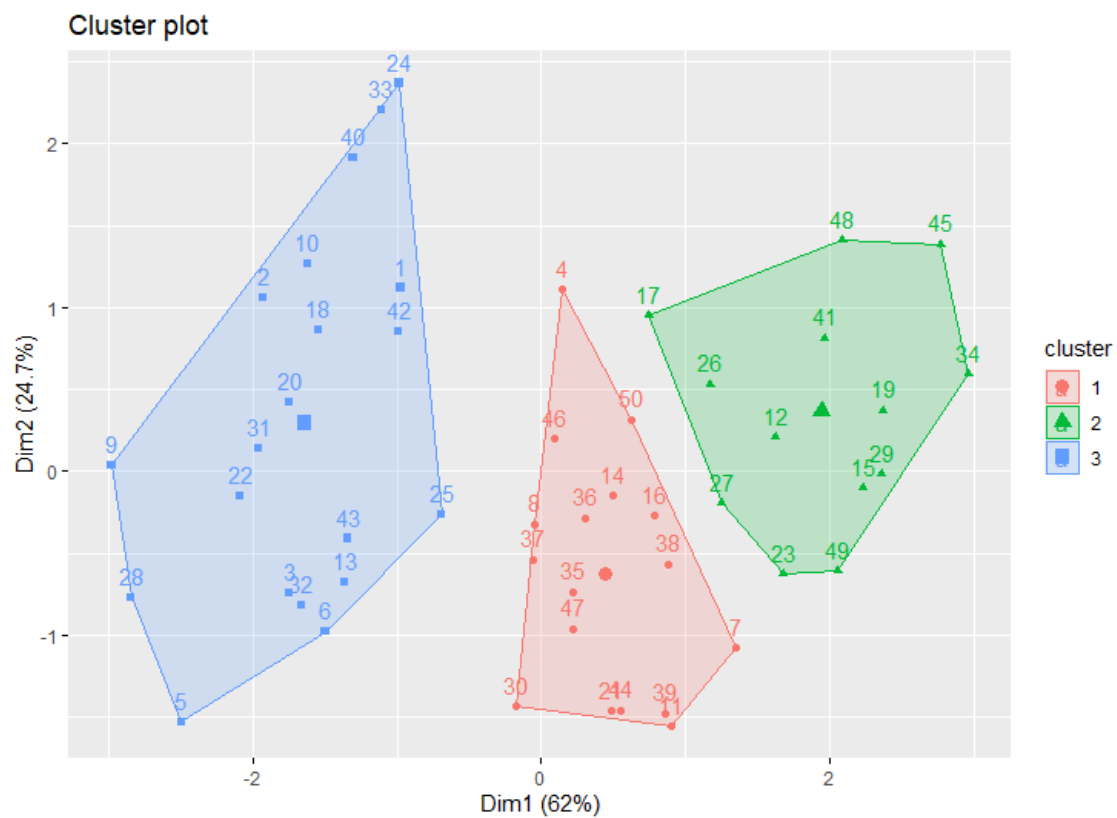
K-medoid centers:



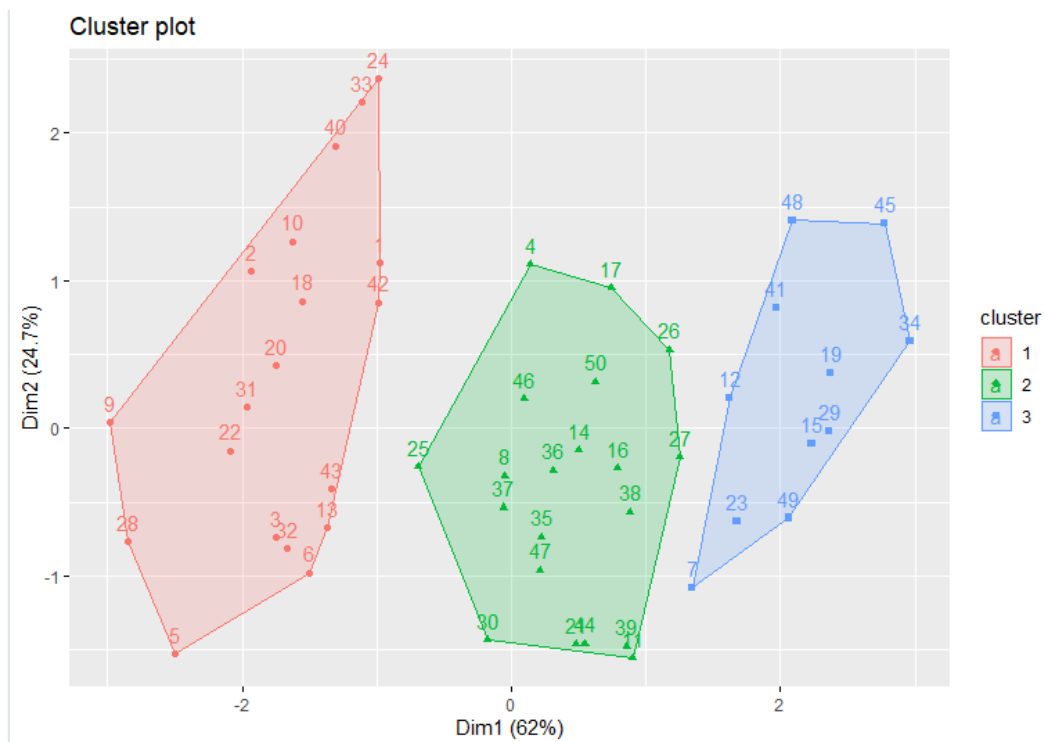
Case II: USArrests Dataset.



K-means centers:



K-medoid centers:



Statistics

Case 1: Iris dataset

K-means centers:

```
> fit$center
      x Sepal.Length Sepal.width Petal.Length Petal.width
1 -1.15087068 -1.01119138  0.85041372  -1.3006301  -1.2507035
2  0.07534946  0.03881135 -0.73324663   0.3059615   0.2137533
3  1.13936197  1.03196952 -0.07784286   1.0386287   1.0894947
```

K-medoid centers:

```
      x Sepal.Length Sepal.width Petal.Length
[1,] -1.07030973  -0.7769106   0.7861738  -1.3357516
[2,] -0.08056095   0.3099591  -0.5903951   0.1370873
[3,]  0.95522266   0.7930124  -0.1315388   0.9868021
      Petal.width
[1,] -1.3110521
[2,]  0.1320673
[3,]  0.7880307
```

Case II: USArrests Dataset.

K-means centers:

```
> fit1$center
      Murder    Assault  UrbanPop      Rape
1 -0.4469795 -0.3465138  0.4788049 -0.2571398
2 -0.9615407 -1.1066010 -0.9301069 -0.9667633
3  1.0049340  1.0138274  0.1975853  0.8469650
```

K-medoid centers:

```
      Murder    Assault  UrbanPop      Rape
[1,]  0.8292944  1.3708088  0.3081225  1.1603196
[2,] -0.2727580 -0.2371077  0.1699510 -0.1315342
[3,] -1.2829727 -1.3770485 -0.5899924 -1.0603878
```

Program

Case 1: Iris dataset

```
rm(list=ls())

setwd("D:/6th Sem Works/A2- EDA/LAB/Lab7")

data1<-read.csv("D:/6th Sem Works/A2- EDA/LAB/Lab7/iris.csv")

View(data1)

df<-scale(data1)

fit<-kmeans(df,centers=2)

fit$cluster

fit$size

fit$withinss

fit$tot.withinss

Kmax<-15

wcss<-rep(NA,Kmax)

nClust<- list()

for(i in 1:Kmax){

  fit<-kmeans(df,i)

  wcss[i]<-fit$tot.withinss

  nClust[[i]]<-fit$size

}

plot(1:Kmax,wcss,type="b",pch=19)

fit<-kmeans(df,centers=3)

fit$cluster

fit$size

fit$center

library(factoextra)

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

fviz_cluster(fit, data1)

library(cluster)

fitm <- pam(df, 3, metric = "manhattan")

fitm

fitm$medoids

fviz_cluster(fitm, data1)
```

Case II: USArrests Dataset.

```
rm(list=ls())

setwd("D:/6th Sem Works/A2- EDA/LAB/Lab7")

data2<-read.csv("D:/6th Sem Works/A2- EDA/LAB/Lab7/USArrests.csv")

view(data2)

data2<-data2[,-1]

df1<-scale(data2)

fit1<-kmeans(df1,centers=2)

fit1$cluster

fit1$size

fit1$withinss

fit1$tot.withinss

Kmax1<-15

wcss1<-rep(NA,Kmax1)

nClust1<- list()

for(i in 1:Kmax1){

  fit1<-kmeans(df1,i)

  wcss1[i]<-fit1$tot.withinss

  nClust1[[i]]<-fit1$size

}

plot(1:Kmax1,wcss1,type="b",pch=19)

fit1<-kmeans(df1,centers=3)

fit1$cluster

fit1$size

fit1$center

library(factoextra)

fviz_nbclust(df1, kmeans, method = "wss")

fviz_cluster(fit1, data2)

library(cluster)

fitm1 <- pam(df1, 3, metric = "manhattan")

fitm1

fitm1$medoids

fviz_cluster(fitm1, data2)
```