## 1\_clustering.R

## win10

## 2021-05-14

```
rm(list = ls())
options(digits = 5)
# if (!is.null(dev.list())){dev.off()}
# Clustering is a form of UNsupervised learning.
# k-means clustering
library(factoextra)
## Loading required package: ggplot2
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
library(cluster)
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
      filter, lag
## The following objects are masked from 'package:base':
##
      intersect, setdiff, setequal, union
# import (iris) dataset as data-frame
df <- as.data.frame(iris)</pre>
head(df)
##
    Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1
            5.1
                       3.5
                                  1.4
                                             0.2 setosa
## 2
            4.9
                       3.0
                                  1.4
                                             0.2 setosa
## 3
            4.7
                       3.2
                                  1.3
                                             0.2 setosa
## 4
                                             0.2 setosa
            4.6
                       3.1
                                  1.5
## 5
            5.0
                       3.6
                                  1.4
                                             0.2 setosa
## 6
            5.4
                       3.9
                                  1.7
                                             0.4 setosa
#plot(df)
#The na.omit R function removes all incomplete cases of a data object
df <- na.omit(df)</pre>
#scale/normalize/standardize
```

```
#scale(function)
#cast as data frame
#drop 5th column because non-numeric, cannot scale non-numeric
df_scaled <- as.data.frame(scale(df[,1:4]))</pre>
set.seed(123)
#kmeans(data, centers, nstart)
#start of run
centers_value = 2
run_cluster <- kmeans(df_scaled[,1:2], centers = centers_value, nstart = 5)</pre>
a_2_12 <- run_cluster$tot.withinss</pre>
run_cluster<- kmeans(df_scaled[,1:3], centers = centers_value, nstart = 5)</pre>
a_2_13 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,1:4], centers = centers_value, nstart = 5)</pre>
a_2_14 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,2:3], centers = centers_value, nstart = 5)</pre>
a_2_23 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,2:4], centers = centers_value, nstart = 5)</pre>
a_2_24 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,3:4], centers = centers_value, nstart = 5)</pre>
a_2_34 <- run_cluster$tot.withinss</pre>
center_results <- c(a_2_34, a_2_24, a_2_23, a_2_14, a_2_13, a_2_12)
center_labels <- c("a_2_34", "a_2_24", "a_2_23", "a_2_14", "a_2_13", "a_2_12")
center_2_df <- data.frame(center_labels, center_results)</pre>
center_2_df
##
     center_labels center_results
## 1
            a_2_34
                            53.808
## 2
            a_2_24
                           148.567
## 3
            a_2_23
                           115.881
## 4
            a_2_14
                           220.879
            a_2_13
## 5
                           189.199
## 6
                           165.839
            a_2_12
#compare to original response
table(run_cluster$cluster, iris$Species)
##
##
       setosa versicolor virginica
                       50
                                  50
##
            0
                        0
                                   0
##
     2
           50
#end of run
#start of run
centers_value = 3
run_cluster <- kmeans(df_scaled[,1:2], centers = centers_value, nstart = 5)</pre>
```

```
a_3_12 <- run_cluster$tot.withinss</pre>
# tot.withinss = Total within-cluster sum of squares, i.e.sum(withinss).
run_cluster<- kmeans(df_scaled[,1:3], centers = centers_value, nstart = 5)</pre>
a_3_13 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,1:4], centers = centers_value, nstart = 5)</pre>
a 3 14 <- run cluster$tot.withinss
run_cluster <- kmeans(df_scaled[,2:3], centers = centers_value, nstart = 5)</pre>
a_3_23 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,2:4], centers = centers_value, nstart = 5)</pre>
a_3_24 <- run_cluster$tot.withinss
run_cluster <- kmeans(df_scaled[,3:4], centers = centers_value, nstart = 5)</pre>
a_3_34 <- run_cluster$tot.withinss</pre>
center_results <- c(a_3_34, a_3_24, a_3_23, a_3_14, a_3_13, a_3_12)
center_labels <- c("a_3_34", "a_3_24", "a_3_23", "a_3_14", "a_3_13", "a_3_12")
center_3_df <- data.frame(center_labels, center_results)</pre>
#compare to original response
table(run_cluster$cluster, iris$Species)
##
##
       setosa versicolor virginica
            0
                                  46
##
                        2
     1
##
            0
                       48
                                   4
                        0
                                   0
##
    3
           50
#end of run
#start of run
centers_value = 4
run_cluster <- kmeans(df_scaled[,1:2], centers = centers_value, nstart = 5)</pre>
a_4_12 <- run_cluster$tot.withinss</pre>
run cluster<- kmeans(df scaled[,1:3], centers = centers value, nstart = 5)</pre>
a_4_13 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,1:4], centers = centers_value, nstart = 5)</pre>
a_4_14 <- run_cluster$tot.withinss</pre>
run cluster <- kmeans(df scaled[,2:3], centers = centers value, nstart = 5)
a_4_23 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,2:4], centers = centers_value, nstart = 5)</pre>
a_4_24 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,3:4], centers = centers_value, nstart = 5)</pre>
a_4_34 <- run_cluster$tot.withinss
```

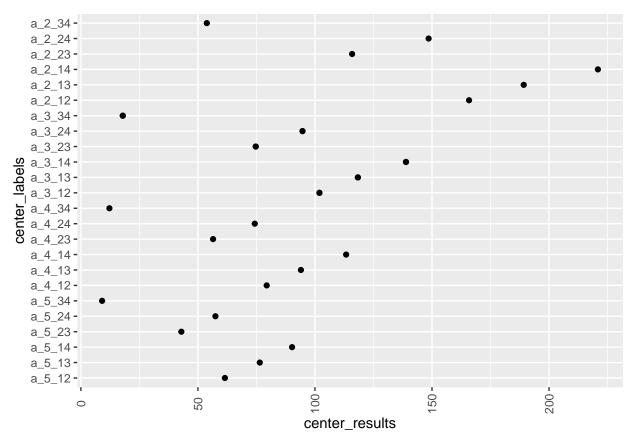
```
center_results <- c(a_4_34, a_4_24, a_4_23, a_4_14, a_4_13, a_4_12)
center_labels <- c("a_4_34", "a_4_24", "a_4_23", "a_4_14", "a_4_13", "a_4_12")
center_4_df <- data.frame(center_labels, center_results)</pre>
#end of run
#start of run
centers_value = 5
run_cluster <- kmeans(df_scaled[,1:2], centers = centers_value, nstart = 5)</pre>
a_5_12 <- run_cluster$tot.withinss</pre>
run_cluster<- kmeans(df_scaled[,1:3], centers = centers_value, nstart = 5)</pre>
a_5_13 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,1:4], centers = centers_value, nstart = 5)</pre>
a_5_14 <- run_cluster$tot.withinss</pre>
run_cluster <- kmeans(df_scaled[,2:3], centers = centers_value, nstart = 5)</pre>
a_5_23 <- run_cluster$tot.withinss
run_cluster <- kmeans(df_scaled[,2:4], centers = centers_value, nstart = 5)</pre>
a_5_24 <- run_cluster$tot.withinss
run_cluster <- kmeans(df_scaled[,3:4], centers = centers_value, nstart = 5)</pre>
a_5_34 <- run_cluster$tot.withinss</pre>
center_results <- c(a_5_34, a_5_24, a_5_23, a_5_14, a_5_13, a_5_12)
center_labels <- c("a_5_34", "a_5_24", "a_5_23", "a_5_14", "a_5_13", "a_5_12")
center_5_df <- data.frame(center_labels, center_results)</pre>
#compare to original response
table(run_cluster$cluster, iris$Species)
##
       setosa versicolor virginica
##
##
     1
            0
                       24
##
     2
            0
                        3
                                  20
            0
                        0
                                  26
##
     3
           50
                        0
                                   0
##
     4
                                   0
     5
                       23
##
#end of run
all_results <- rbind2(center_5_df, center_4_df)</pre>
all_results <- rbind2(all_results, center_3_df)</pre>
all_results <- rbind2(all_results, center_3_df)</pre>
all_results <- rbind2(all_results, center_2_df)</pre>
all_results
```

## center\_labels center\_results

```
9.078
## 1
             a_5_34
## 2
             a_5_24
                             57.490
             a_5_23
## 3
                             42.934
## 4
                             90.228
             a_5_14
## 5
             a_5_13
                             76.444
## 6
                             61.530
             a_5_12
## 7
             a 4 34
                             12.201
                             74.289
## 8
             a_4_24
## 9
             a_4_23
                             56.482
## 10
             a_4_14
                            113.332
## 11
             a_4_13
                             94.010
## 12
             a_4_12
                             79.388
                             17.907
## 13
             a_3_34
## 14
                             94.683
             a_3_24
## 15
             a_3_23
                             74.708
## 16
             a_3_14
                            138.888
## 17
                            118.345
             a_3_13
## 18
             a_3_12
                            101.931
## 19
             a_3_34
                             17.907
## 20
             a_3_24
                             94.683
## 21
             a_3_23
                             74.708
## 22
             a_3_14
                            138.888
## 23
             a_3_13
                            118.345
## 24
             a_3_12
                            101.931
## 25
             a_2_34
                             53.808
## 26
             a_2_24
                            148.567
## 27
             a_2_23
                            115.881
## 28
                            220.879
             a_2_14
## 29
             a_{2}13
                            189.199
                            165.839
## 30
             a_2_12
all_results_total_withinss <- all_results[order(all_results$center_results),][1,]
all_results_total_withinss
##
     center_labels center_results
## 1
            a_5_34
                             9.078
#based on results above, select run a_5_34
a_5_34
## [1] 9.078
# rerun the clustering so that selected configuration is now active in "run_cluster" variable
run_cluster <- kmeans(df_scaled[,3:4], centers = centers_value, nstart = 5)</pre>
a_5_34 <- run_cluster$tot.withinss</pre>
#compare to original response
table(run_cluster$cluster, iris$Species)
##
##
       setosa versicolor virginica
##
     1
            0
                       24
##
     2
            0
                        0
                                 26
##
     3
            0
                       23
                                  0
##
     4
           50
                        0
                                  0
                        3
##
     5
            0
                                 20
```

```
all_results <- arrange(all_results, center_results)

ggplot(
    all_results, aes(y = center_labels, x = center_results)) +
    geom_point() +
    theme(axis.text.x = element_text(angle = 90, vjust = 0.5)
)</pre>
```



```
all_results <- arrange(all_results, center_labels)

ggplot(
   all_results, aes(x = center_labels, y = center_results)) +
   geom_point() +
   theme(axis.text.x = element_text(angle = 90, vjust = 0.5)
   )</pre>
```

