Coursework project

on the module 1 "The Datascientist's Toolbox"

done by Alexandra Avdan

July 2023

This project is available in the public repository on GitHub. You can check it following this link.

The first step of this project is to load and analyze three datasets from the pre-installed R package 'datasets'. The full list of packages, available at the named package, is available in the link to the R documentation. From the named library it was decided to pick 3 datasets by my own preferences to analyze data each of the selected datasets contains.

Respecting to the fact, that our knowledge is not deep enough to build complex analysis, in my work is represented only brief look on each of datasets, within the basics of the descriptive analysis with the elements of exploratory approach in order to find some possible dependencies of the given data.

• The first dataset is pretty classical dataset for this subject area, 'iris'. This dataset represents measurements of sepal length, sepal width, petal length and petal width of the three iris species: setosa, virginica and versicolor. There are 50 rows per each of the iris species in the dataset.

```
# get the data of the whole dataset
library(datasets)
data("iris")
```

On the image below are represented top 12 rows of the iris dataset, ordered by index at the ascending order.

÷	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	4.6	3.1	1.5	0.2	setosa
5	5.0	3.6	1.4	0.2	setosa
6	5.4	3.9	1.7	0.4	setosa
7	4.6	3.4	1.4	0.3	setosa
8	5.0	3.4	1.5	0.2	setosa
9	4.4	2.9	1.4	0.2	setosa
10	4.9	3.1	1.5	0.1	setosa
11	5.4	3.7	1.5	0.2	setosa
12	4.8	3.4	1.6	0.2	setosa

The next step is to obtain summary on the dataset.

The following summary contains values on the criteria, such as

- *minimum* (*Min.*),
- 1st Quartile (1st Qu.) (i.e. 25% of data was taken under the produced value),
- Median (can be named as a middle value of some numeric sequence),
- Mean (average value),
- 3rd Quartile (3rd Qu.) (i.e. 75% of data was taken under the produced value),
- maximum (Max.)

for numeric values and

• number of occurrencies of each value for string values (column 'Species' in iris dataset).

Additionally, the range value can be calculated if needed. However, without any certain question, needed to be solved, these values are useless. According to the dataset data, this dataset has no missing values (no NaN or NULLs).

summary(iris)

```
##
     Sepal.Length
                      Sepal.Width
                                       Petal.Length
                                                        Petal.Width
##
                            :2.000
    Min.
           :4.300
                                              :1.000
                                                               :0.100
                     Min.
                                      Min.
                                                       Min.
    1st Qu.:5.100
                     1st Qu.:2.800
                                      1st Qu.:1.600
                                                       1st Qu.:0.300
   Median :5.800
                     Median :3.000
                                      Median :4.350
                                                       Median :1.300
##
                                              :3.758
##
    Mean
           :5.843
                     Mean
                            :3.057
                                      Mean
                                                       Mean
                                                               :1.199
##
    3rd Qu.:6.400
                     3rd Qu.:3.300
                                      3rd Qu.:5.100
                                                       3rd Qu.:1.800
##
    Max.
           :7.900
                     Max.
                            :4.400
                                      Max.
                                              :6.900
                                                       Max.
                                                               :2.500
##
          Species
##
    setosa
               :50
##
    versicolor:50
##
    virginica:50
##
##
##
```

From the data below it is clear that all the metrics are calculated on all the iris species. It would be more informative, if the metrics were calculated on the certain needed iris type. Assume, we need to get some metrics on the iris setosa. Let's create dataframe with setosa values **only**.

```
# attaching library dplyr to use filter() function
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

# to make it more usable to call the names of the columns making it to lowercase
names(iris) <- tolower(names(iris))
# creating dataframe with setosa rows only
setosa <- filter(iris, species == "setosa")
# show top-6 values of the obtained dataframe ordered by the ascending of the index
head(setosa)</pre>
```

```
##
     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
                                                     0.2 setosa
                                        1.3
## 4
              4.6
                                                     0.2 setosa
                           3.1
                                        1.5
              5.0
## 5
                           3.6
                                        1.4
                                                     0.2 setosa
## 6
              5.4
                           3.9
                                        1.7
                                                     0.4 setosa
```

The next step is obtaining summary metrics on the measurements of different *setosa* irises. According to the data of the column '*species*' there are no other data here, except from *setosa* data. The obtained metrics' data also differ from the previously obtained.

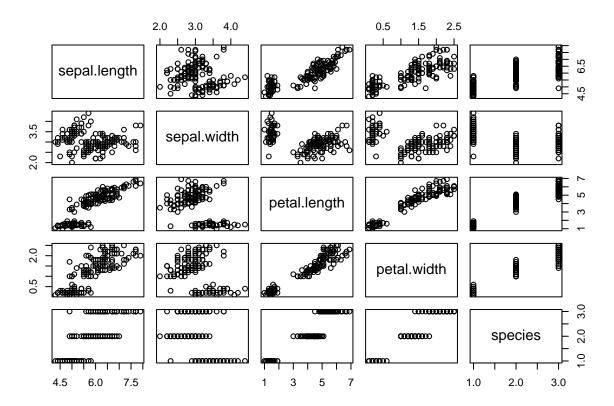
summary(setosa)

```
##
     sepal.length
                      sepal.width
                                       petal.length
                                                        petal.width
##
           :4.300
                     Min.
                            :2.300
                                      Min.
                                             :1.000
                                                       Min.
                                                              :0.100
##
    1st Qu.:4.800
                     1st Qu.:3.200
                                      1st Qu.:1.400
                                                       1st Qu.:0.200
##
   Median :5.000
                     Median :3.400
                                      Median :1.500
                                                       Median :0.200
##
   Mean
           :5.006
                     Mean
                            :3.428
                                             :1.462
                                                              :0.246
                                      Mean
                                                       Mean
##
    3rd Qu.:5.200
                     3rd Qu.:3.675
                                      3rd Qu.:1.575
                                                       3rd Qu.:0.300
##
    Max.
           :5.800
                     Max.
                            :4.400
                                      Max.
                                             :1.900
                                                       Max.
                                                              :0.600
##
          species
##
              :50
    setosa
##
    versicolor: 0
##
    virginica: 0
##
##
##
```

The final step is creating visualizations. Note, that there was no particular aim, set as a goal of investigations. The first visualization is done on the whole iris dataset. From the visualization below some conclusions could be made:

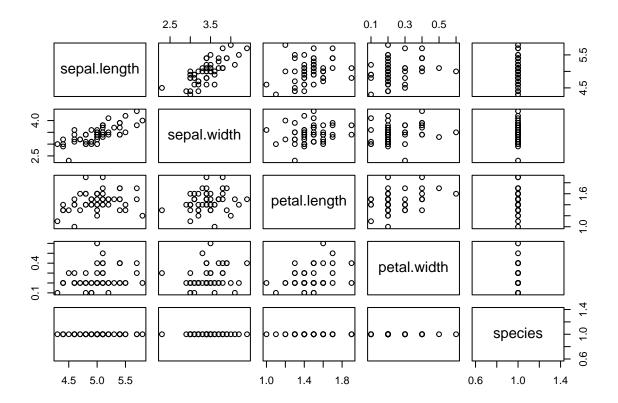
- it was a bad practice to visualize the whole dataset, that contains both numeric and string values. The obtained row and column 'species' is not pretty informative;
- it is useful to assess the whole picture of the dataset, but for more accurate conclusions and results, more detailed visualizations have to be made.

```
plot(iris)
```



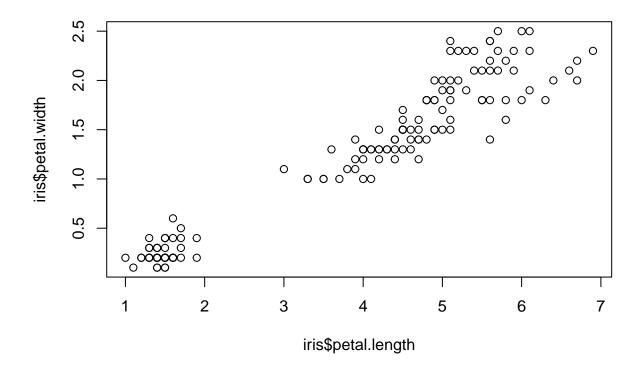
Let's create a visualization of the created *setosa* data **only** dataframe. The obtained visualization allows to make conclusions and analysis on the certain type of iris flowers *setosa*. For this graph comments and observation are the same, as for the previous one.

plot(setosa)



To reveal the dependencies between the metrics of the dataset, we have to build a more detailed visualization. To reveal the relationships between *petal length* and *petal width* let's build a graph. **All the dataset** *iris* will be analyzed. From the obtained visualization it is clear, that in most cases increasing of petal length is connected with the increasing of petal width and vice versa. The other fact is that there is no petal width and length values in the interval from 2 to 3 including these values. There is also one more interesting detail: there is a cluster in the lower left corner. This cluster may contain data on the miniature sized irises. That may be just small ones on any of the represented types or just small type (that may be *setosa*, *versicolor* or *virginica* with equal probabilities).

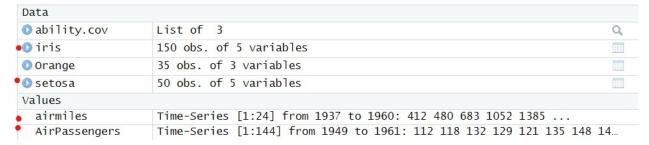
plot(iris\$petal.length, iris\$petal.width)



More investigations can be made with the formulated question to solve.

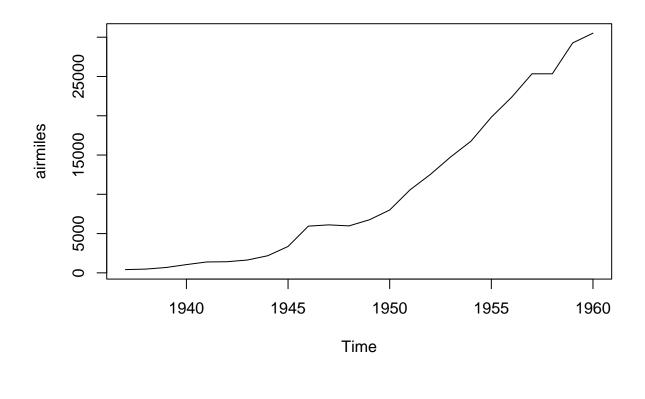
• The second dataset is 'airmiles'. This dataset is represented by the time series of values of flown passenger miles by the USA commercial airlines from 1937 to 1960.

It is important to highlight, that **iris** dataset is represented by the *Data*, whereas **airmiles** and other dataset, that will be considered later, **AirPassengers**, are represented by the *values of time-series*:



Let's visualize dataset 'airmiles'. From the obtained visualization can be made a conclusion, that with the time number of airmiles flown increases. This characteristic increases particularly rapidly from 1947 to 1957. This decade was probably characterised by the technical progress and popularization of air transport.

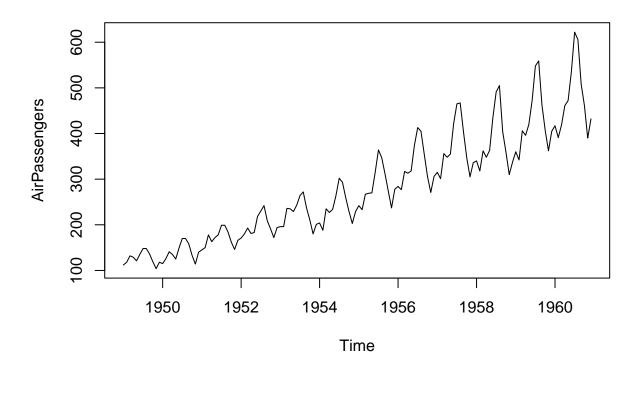
```
library(datasets)
plot(airmiles)
```



• The third dataset is 'AirPassengers'. This dataset, as it was recently mentioned, is represented by the time-series of values of monthly totals of international airline passengers from 1949 to 1960.

Let's visualize dataset 'AirPassengers'. From the obtained visualization it is clear, that it's a representation of the non-linear model. To make an exact conclusion about how does dependent variable (AirPassengers) depends from the independent one (Time) it is needed to approximate existing model to linear one by the means of linear regression. Without this step it can be said that with the time number of air passengers increases.

library(datasets)
plot(AirPassengers)



During this work I've been doing commits on the **linked GitHub repository.** Here is commit history as an evidence of proper work:

