

Data Mining and Machine Learning Lab 2.

In this R lab you will:

Be introduced to Data Exploration in R.

Instructions: Create a file called xxxxxxxx.doc where <xxxxxxx> is your UCD student number. Write your answers in this file and save it to your own computer so you don't lose your answers. Then upload to the moodle before the end of the lab.

At the top of the file, fill in your details below (delete the 'x' where your information goes):

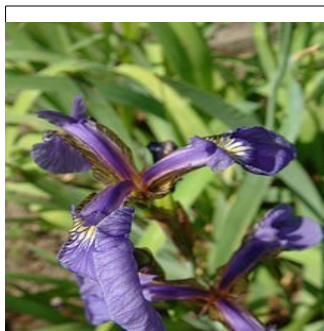
Name: x

BDIC Student Number: x

UCD Student Number: x

The Iris flower data set

In this lab we will use the *Iris flower data set*. The Iris flower data set or Fisher's Iris data set is a multivariate data set introduced by the British statistician and biologist Ronald Fisher in his 1936. The data set consists of 50 samples from each of three species of Iris (*Iris setosa*, *Iris virginica* and *Iris versicolor*). Four features were measured from each sample: the length and the width of the sepals and petals, in centimeters. Based on the combination of these four features, Fisher developed a linear discriminant model to distinguish the species from each other.



Iris setosa



Iris versicolor



Iris virginica

1. Load the iris data set

```
> data(iris)
```

This is your **ABT** – 4 descriptive features (Sepal.Length Sepal.Width Petal.Length Petal.Width) and one target feature (Species)

The function `dim()` returns the dimensions of the data set:

```
> dim(iris)
[1] 150 5
```

Take a look at the attributes of the iris data set:

```
> attributes(iris)
$names
[1] "Sepal.Length" "Sepal.Width"  "Petal.Length" "Petal.Width"  "Species"

$row.names
 [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
[19] 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36
[37] 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
[55] 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
[73] 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
[91] 91 92 93 94 95 96 97 98 99 100 101 102 103 104 105 106 107 108
[109] 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125 126
[127] 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142 143 144
[145] 145 146 147 148 149 150

$class
[1] "data.frame"
```

Look at the first rows of data using:

```
> head(iris)
  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
```

Look at the last rows of data using:

```
> tail(iris)
```

Or, specify exactly which rows you want to look at e.g. the first three rows:

```
> iris[1:3, ]
```

The first 10 values of Sepal.Length can be extracted using either:

```
> iris[1:10, "Sepal.Length"]
```

or

```
> iris$Sepal.Length[1:10]
```

2. Get summary statistics

Summary statistics are a good place to start when looking at a new dataset. The `summary()` function provides the summary of each variable in a `data.frame`.

The syntax and use of the `summary()` function is: `summary(object)`

The function returns a table with a column for each of the variables in a `data.frame`.

For categorical variables (factors, logical, and character variables), the frequency of occurrences is returned. Low frequency levels will be combined in an "other" category if there are too many levels to be displayed.

For numeric variables, the five number summary (median, first quartile, third quartile, min, and max) and the mean are returned.

```
> summary(iris)
 Sepal.Length      Sepal.Width      Petal.Length      Petal.Width
Min.      :4.300    Min.      :2.000    Min.      :1.000    Min.      :0.100
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
Mean   :5.843    Mean   :3.057    Mean   :3.758    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
```

3. The Data Quality Report

We can use the information above to generate the Data Quality Report, but we also need:

- the standard deviation for each feature
- the percentage of instances in the ABT that are missing a value for each feature
- the cardinality of each feature (cardinality measures the number of distinct values present in the ABT for a feature)

Can you get the standard deviation for sepal length?

```
> sd(iris$Sepal.Length)
[1] 0.8280661
```

Question 1:

- What is standard deviation for sepal length? 0.8280661

To get the % missing and the cardinality we can use the Library Hmisc. To install the library type:

```
> install.packages("Hmisc")  
(This will take a few minutes...)
```

Load the Library:

```
> library("Hmisc")
```

Use the function describe() get get a summary of the iris data set which shows if there is any missing data and the number of distinct values for each feature (i.e. the cardinality):

```
> describe(iris)  
iris
```

```
5 Variables      150 Observations  
-----  
Sepal.Length  
  n missing distinct    Info    Mean      Gmd      .05      .10  
150      0        35    0.998    5.843    0.9462    4.600    4.800  
  .25     .50     .75     .90     .95  
5.100    5.800    6.400    6.900    7.255  
  
lowest : 4.3 4.4 4.5 4.6 4.7, highest: 7.3 7.4 7.6 7.7 7.9  
-----  
Sepal.Width  
  n missing distinct    Info    Mean      Gmd      .05      .10  
150      0        23    0.992    3.057    0.4872    2.345    2.500  
  .25     .50     .75     .90     .95  
2.800    3.000    3.300    3.610    3.800  
  
lowest : 2.0 2.2 2.3 2.4 2.5, highest: 3.9 4.0 4.1 4.2 4.4  
-----  
Petal.Length  
  n missing distinct    Info    Mean      Gmd      .05      .10  
150      0        43    0.998    3.758    1.979    1.30    1.40  
  .25     .50     .75     .90     .95  
1.60    4.35    5.10    5.80    6.10  
  
lowest : 1.0 1.1 1.2 1.3 1.4, highest: 6.3 6.4 6.6 6.7 6.9  
-----  
Petal.Width  
  n missing distinct    Info    Mean      Gmd      .05      .10  
150      0        22    0.99    1.199    0.8676    0.2     0.2  
  .25     .50     .75     .90     .95  
0.3     1.3     1.8     2.2     2.3  
  
lowest : 0.1 0.2 0.3 0.4 0.5, highest: 2.1 2.2 2.3 2.4 2.5  
-----  
Species  
  n missing distinct  
150      0        3  
  
Value      setosa versicolor virginica  
Frequency      50      50      50  
Proportion    0.333    0.333    0.333  
-----
```

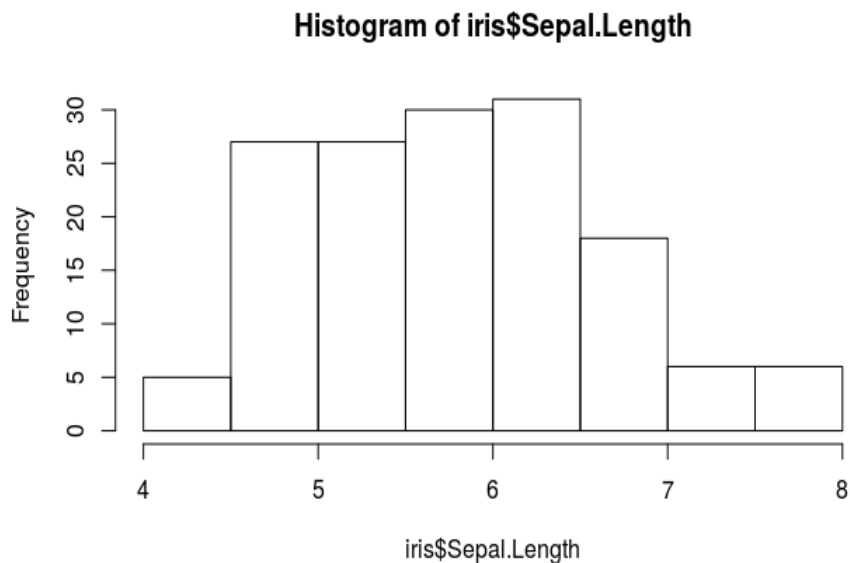
Question 2:

- What is the cardinality i.e. the number of “distinct” values for:
 - Sepal.Length 35
 - Sepal.Width 23
 - Petal.Length 43
 - Petal.Width 22
 - Species 3

4. Histograms

Next plot histograms for your 4 descriptive features. For Sepal length you can do this:

```
> hist(iris$Sepal.Length)
```



The histogram will appear in RStudio. Click on the “export” button and save the histogram and copy it into your .doc file which you will upload to moodle at the end of the lab. Repeat this for the other 3 descriptive features.

5. Bar Plot

Create a bar plot for the target feature (Species):

Species is “Categorical” not numerical i.e. see what you get when you type:

```
> iris$Species
```

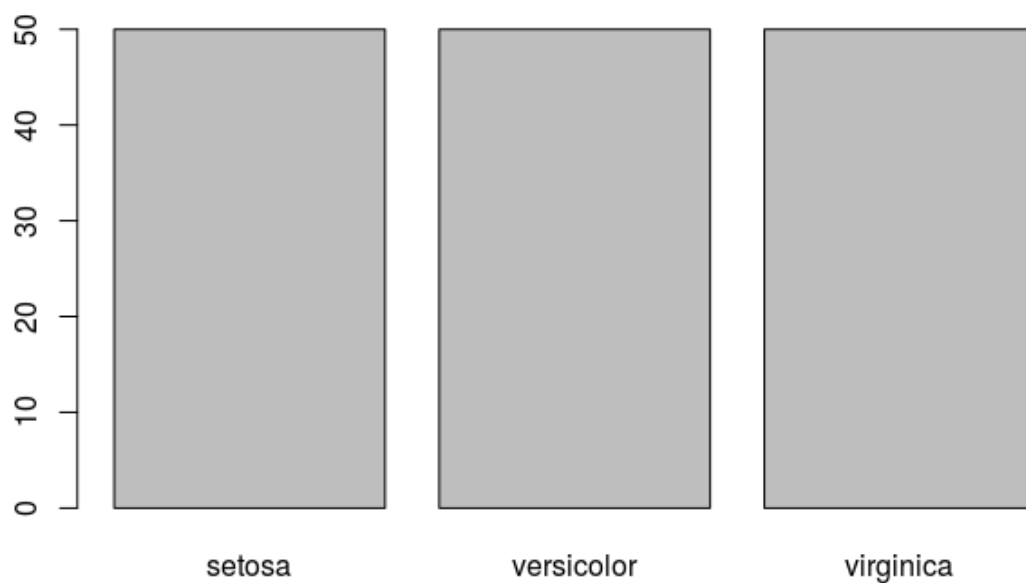
So, before we can plot the data we must create a table:

```
> table(iris$Species)
setosa versicolor virginica
    50         50         50
```

We can then create the bar plot:

```
> barplot(table(iris$Species))
```

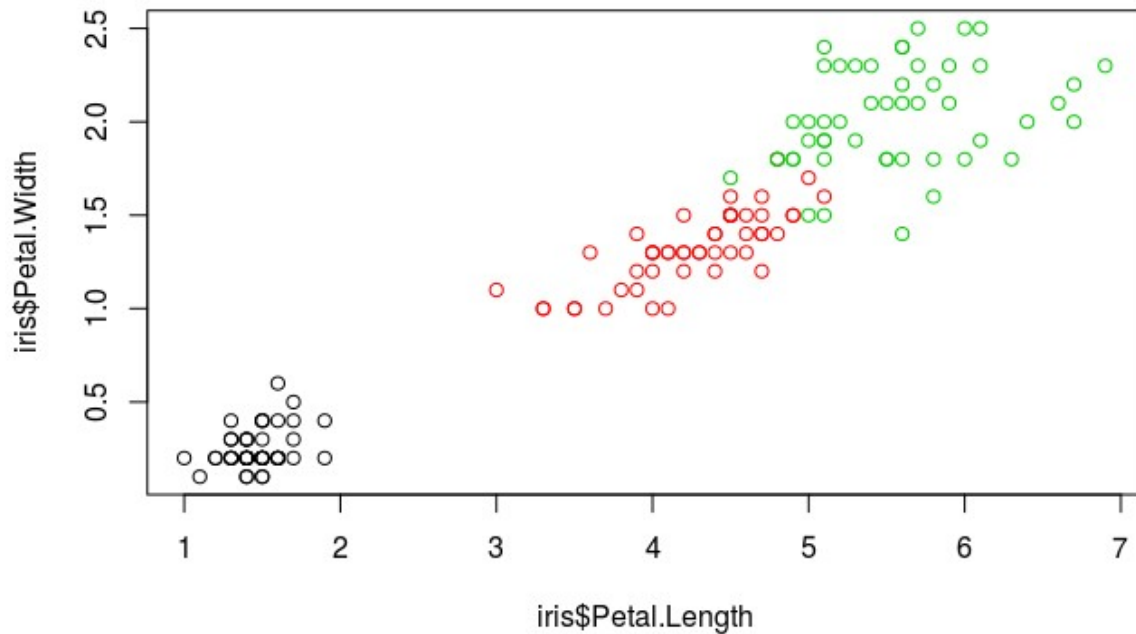
Create the bar plot and export it, save it and copy it into your file for upload.



6. Scatter plots

We can use scatter plots to show the relationship between two numeric variables as follows:

```
> plot(x=iris$Petal.Length, y=iris$Petal.Width, col=iris$Species)
```



Export the plot and save it into your file for upload.

Repeat for Sepal.Length and Sepal.Width

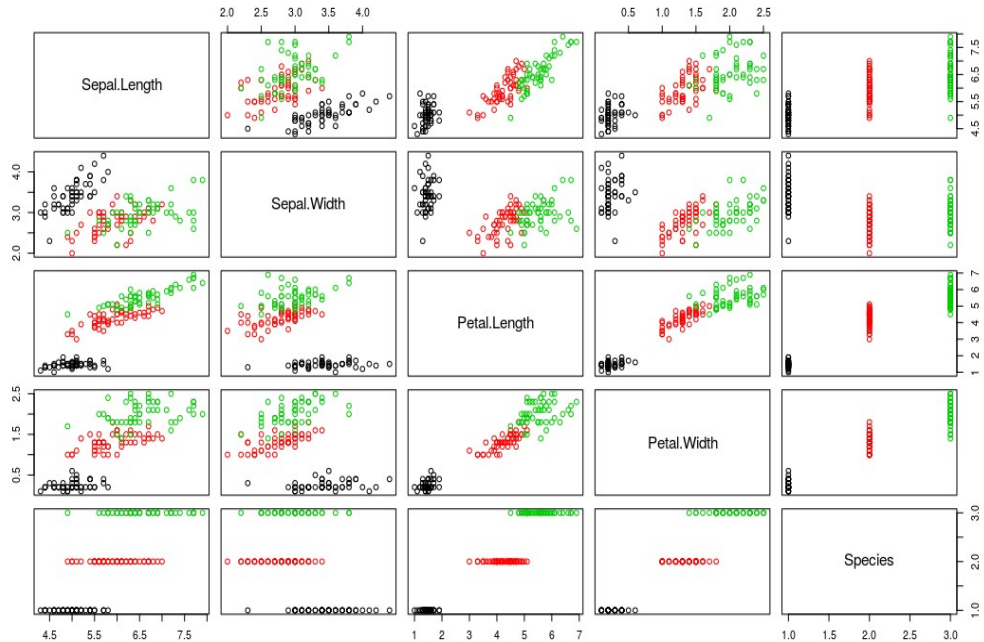
Question 3: What do you notice about the two plots (i.e. does it look like there is a correlation between either pair of features)?

- The plots show that there is a strong correlation between Petal length and width (i.e. as petal length increases so does petal width) but there is not a correlation between Sepal length and width.

7. Scatter plot matrix

A scatter plot matrix shows the relationship between several numeric variables and can be plotted as follows:

```
> pairs(iris, col=iris$Species)
```



8. Correlation

Another useful method to explore the relationships within a dataset is to examine the correlation between the variables.

The syntax and use of the `cor()` function

```
cor(object)
```

Returns a matrix of the correlations.

The object passed to `cor` needs to be a two dimensional object with a type of numeric.

The iris data set is a two dimensional object, though not all the variables are numeric. So we will exclude the non-numeric variable (Species) from our call to `cor()`.

```
> cor(iris[, 1:4])
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width
Sepal.Length	1.0000000	-0.1175698	0.8717538	0.8179411
Sepal.Width	-0.1175698	1.0000000	-0.4284401	-0.3661259
Petal.Length	0.8717538	-0.4284401	1.0000000	0.9628654
Petal.Width	0.8179411	-0.3661259	0.9628654	1.0000000

9. Correlation matrix visualization

Install the package 'corrplot' and include the library:

```
> install.packages('corrplot')
```

```
> library('corrplot')
```

Create the correlation matrix:

```
> correlationMatrix <- cor(iris[, 1:4])
```

Create the plot:

```
> corrplot(correlationMatrix, method = "circle")
```

Export the plot and save it into your file for upload.

Question 4: What does the colour of the circle signify? What does the size of the circle mean? (hint `?corrplot`)

- The sizes of the circles are scaled according to the absolute value of the strength of the correlation to draw attention to those pairs of features with the strongest relationships.
- Blue represents positive correlations and red represents negative or inverse correlations.

