

Assignment A1

Title - Analysis on Iris flower Dataset.

Problem Statement :-

Download the iris flower dataset or any other dataset into a dataframe. Use python / R and perform following :

1. How many features are there and what are their types?
2. Compare & display summary statistics for each features available in dataset (e.g. min, max, mean, std-dev, variance, percentile)
3. Data visualization - create a histogram for each feature in the dataset to illustrate feature distribution.
4. Create a box plot for each feature in the dataset. All of the box plots should be combined into a single plot. compare distributions and find outliers.

Objectives -

- To learn the concept & terminologies in data analytics.
- To learn how to display summary set statistics & charts for each feature.

Outcomes - We will be able to -

- learn the concepts in data analytics.
- learn how to summarize & plot charts.

Theory -

A) Iris flower dataset -

- The dataset is a multivariate dataset introduced by the British statistician & biochemist Ronald Fisher in 1936.

- Dataset consist of 50 samples from each of 3 species of Iris, which are *setosa*, *virginica* & *versicolor*.
- four features measured from each sample are length and width of sepals & petals in mm.

B7 Summary statistics :-

1. Mean :- It identifies the average value of set of values

$$\bar{x} = \frac{\sum x_i}{n}$$
 where x_i = value of attributes.
 n = total no. of items

2. Range - It shows the mathematical model between the lowest & highest values in the dataset. it measures the variability of dataset.

$$\text{Range} = \text{max} - \text{min}$$

3. Standard deviation :- It measures the variability of dataset like range. The smaller standard deviation indicates less variability.

$$\sigma = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n}}$$

4. Variance - It measures the how far the data is spread out

$$\sigma^2 = \frac{\sum (x_i - \bar{x})^2}{n}$$

c) Applications -

1. Histogram -

- It is suitable for visualizing distribution of numeric data over a continuous interval or a certain time period.

- The histogram organises large amount of data & provides a visualization quickly, using a single dimension.

2. Box plot -

- It allows quick graphical examination of one or more dataset. It may seem primitive than a histogram but they do have some advantages.
- They take up space & are particularly useful for comparing distributions between several groups of data.

3. Data visualization

- It quickly creates insightful data visuals.
- They allow anyone to organise & present information quickly.

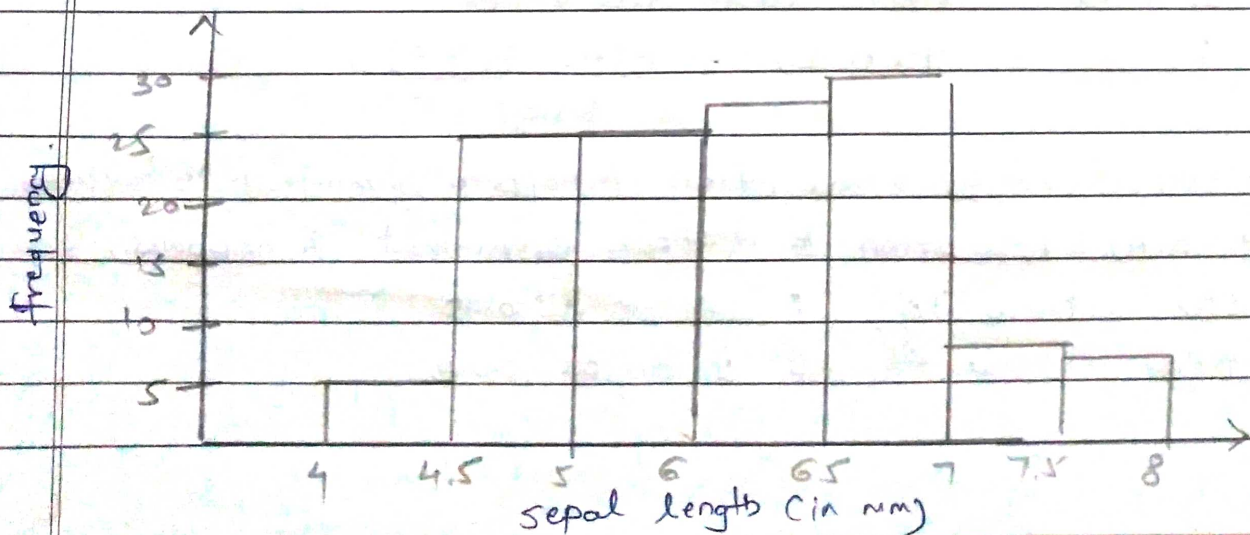
Conclusion -

Thus, we studied about concepts in data analytics & the dataset. we also presented the data in charts & box plots.

Test case -

Input	Output
Column of Sepal length	Mean = 5.843 mm.

Histogram of sepal length.



```
In [3]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [4]: df=pd.read_csv(r"C:\Users\Viraj Shinde\Desktop\LP1\iris.data")
```

```
In [5]: df.head()
```

Out[5]:

	Sepal length	Sepal width	Petal length	Petal width	Species
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

```
In [6]: df.tail()
```

Out[6]:

	Sepal length	Sepal width	Petal length	Petal width	Species
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

```
In [7]: X= df.drop('Species', axis = 1)
```

```
In [8]: df.shape
list(df.columns)
```

Out[8]: ['Sepal length', 'Sepal width', 'Petal length', 'Petal width', 'Species']

```
In [9]: df.dtypes
```

Out[9]: Sepal length float64
Sepal width float64
Petal length float64
Petal width float64
Species object
dtype: object

```
In [10]: df['Sepal length'].describe()
```

```
Out[10]: count      150.000000  
mean         5.843333  
std          0.828066  
min          4.300000  
25%          5.100000  
50%          5.800000  
75%          6.400000  
max          7.900000  
Name: Sepal length, dtype: float64
```

```
In [11]: df['Sepal width'].describe()
```

```
Out[11]: count      150.000000  
mean         3.054000  
std          0.433594  
min          2.000000  
25%          2.800000  
50%          3.000000  
75%          3.300000  
max          4.400000  
Name: Sepal width, dtype: float64
```

```
In [12]: df['Petal length'].describe()
```

```
Out[12]: count      150.000000  
mean         3.758667  
std          1.764420  
min          1.000000  
25%          1.600000  
50%          4.350000  
75%          5.100000  
max          6.900000  
Name: Petal length, dtype: float64
```

```
In [13]: df['Petal width'].describe()
```

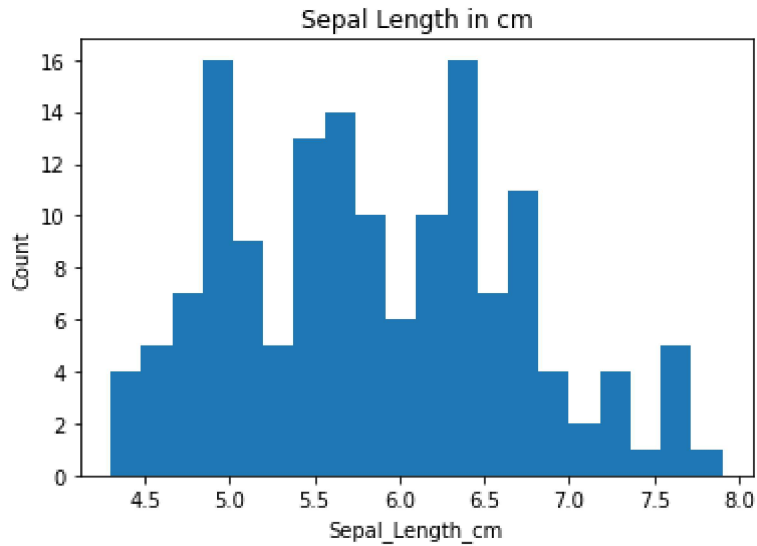
```
Out[13]: count      150.000000  
mean         1.198667  
std          0.763161  
min          0.100000  
25%          0.300000  
50%          1.300000  
75%          1.800000  
max          2.500000  
Name: Petal width, dtype: float64
```

```
In [14]: df['Species'].describe()
```

```
Out[14]: count           150  
unique             3  
top      Iris-versicolor  
freq              50  
Name: Species, dtype: object
```

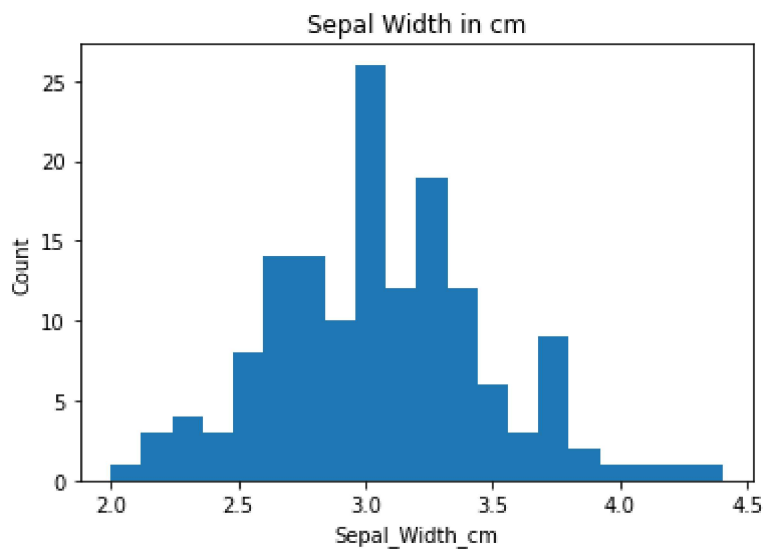
```
In [15]: x = df["Sepal length"]  
plt.hist(x, bins = 20)  
plt.title("Sepal Length in cm")  
plt.xlabel("Sepal_Length_cm")  
plt.ylabel("Count")
```

Out[15]: Text(0, 0.5, 'Count')



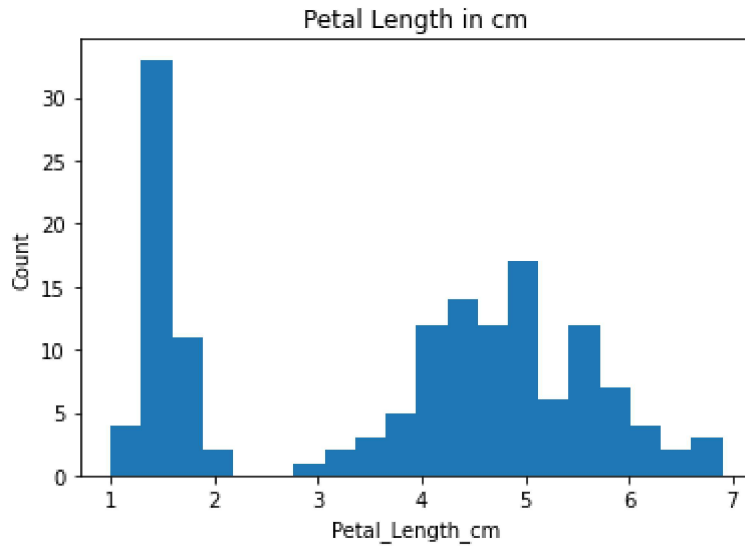
```
In [16]: x = df["Sepal width"]  
plt.hist(x, bins = 20)  
plt.title("Sepal Width in cm")  
plt.xlabel("Sepal_Width_cm")  
plt.ylabel("Count")
```

Out[16]: Text(0, 0.5, 'Count')



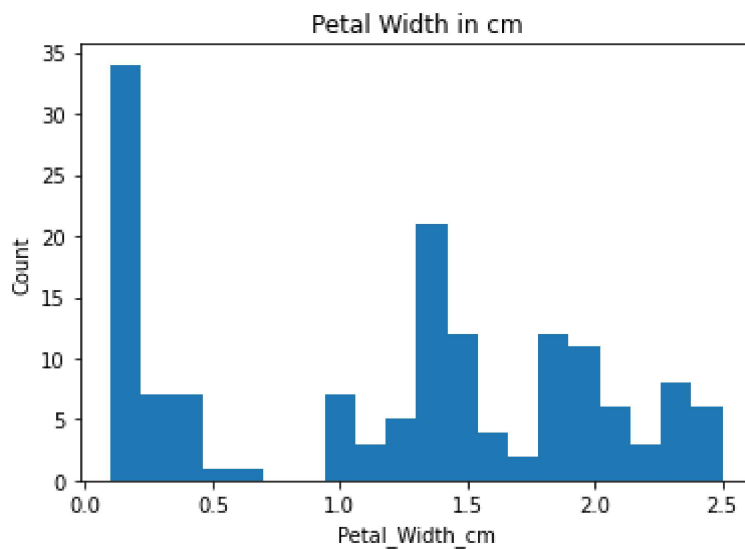
```
In [17]: x = df["Petal length"]
plt.hist(x, bins = 20)
plt.title("Petal Length in cm")
plt.xlabel("Petal_Length_cm")
plt.ylabel("Count")
```

Out[17]: Text(0, 0.5, 'Count')



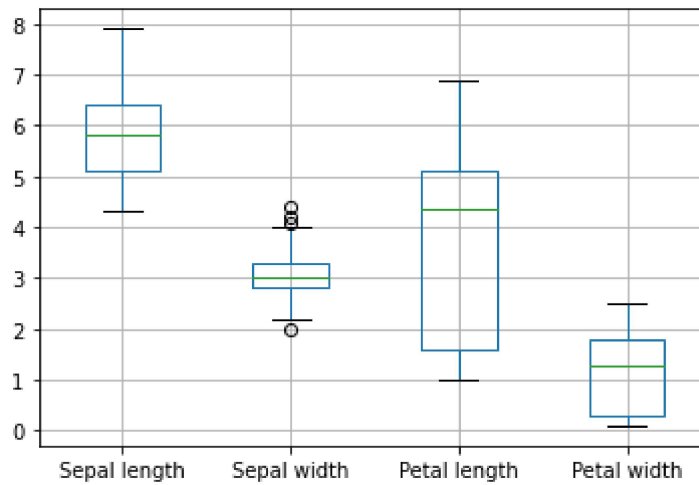
```
In [18]: x = df["Petal width"]
plt.hist(x, bins = 20)
plt.title("Petal Width in cm")
plt.xlabel("Petal_Width_cm")
plt.ylabel("Count")
```

Out[18]: Text(0, 0.5, 'Count')



```
In [19]: X.boxplot()
```

```
Out[19]: <AxesSubplot:>
```



```
In [20]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 150 entries, 0 to 149  
Data columns (total 5 columns):  
#   Column          Non-Null Count  Dtype  
---  ---  
0   Sepal length    150 non-null   float64  
1   Sepal width     150 non-null   float64  
2   Petal length    150 non-null   float64  
3   Petal width     150 non-null   float64  
4   Species         150 non-null   object  
dtypes: float64(4), object(1)  
memory usage: 6.0+ KB
```

```
In [ ]:
```