

Download the Iris flower dataset or any other dataset into a DataFrame. (e.g., <https://archive.ics.uci.edu/ml/datasets/Iris> ). Scan the dataset and give the inference as:

1. List down the features and their types (e.g., numeric, nominal) available in the dataset.
  2. Create a histogram for each feature in the dataset to illustrate the feature distributions.
  3. Create a boxplot for each feature in the dataset.
  4. Compare distributions and identify outliers
- 

```
In [10]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [11]: df = pd.read_csv("iris dataset.csv")
```

```
In [12]: 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

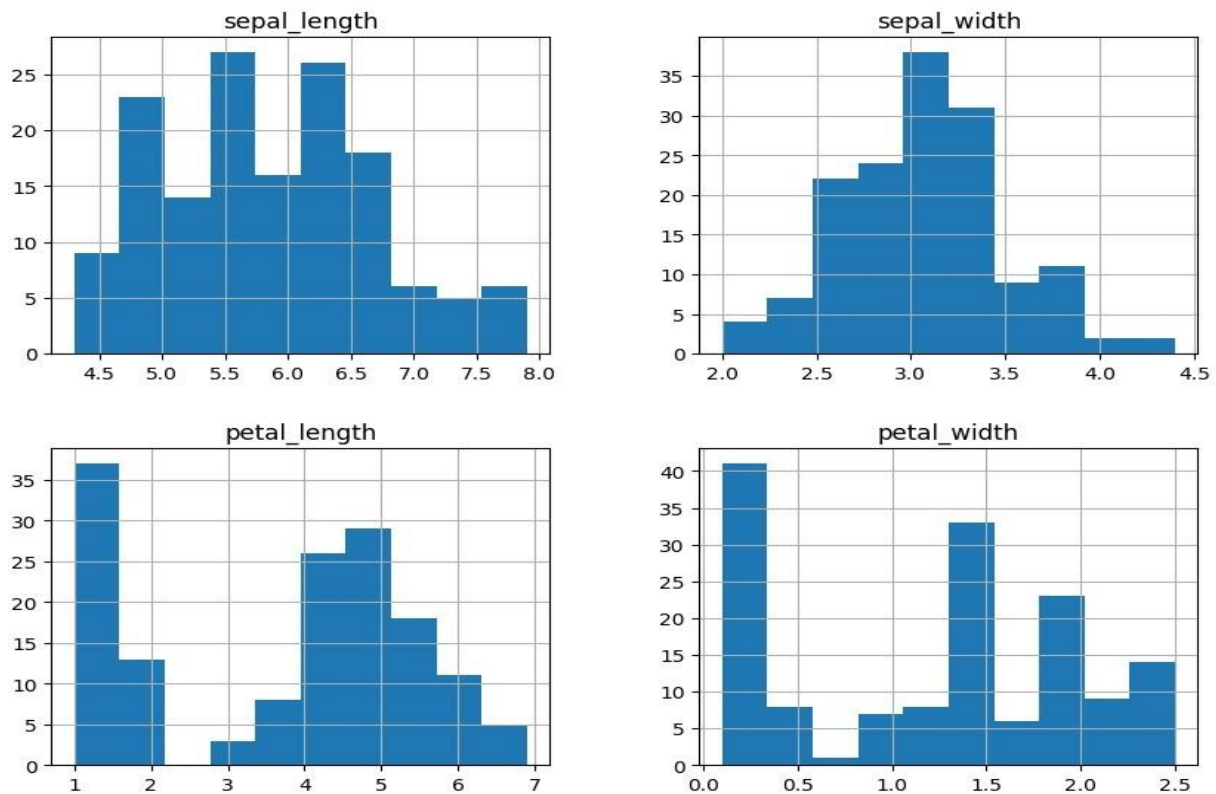
dtypes: float64(4), object(1) memory

usage: 6.0+ KB

## 2) Create a histogram for each feature in the dataset to illustrate the feature distributions

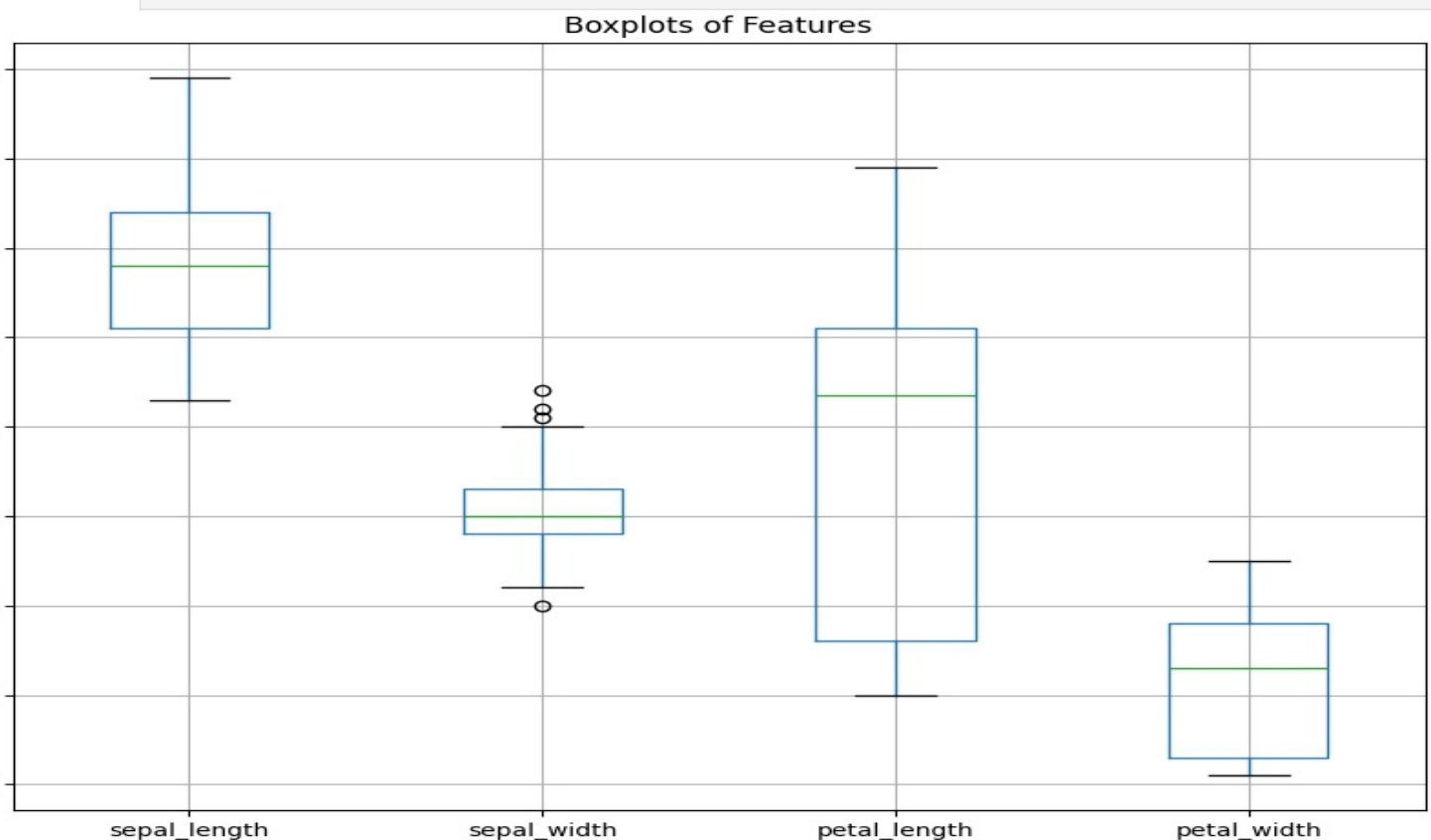
```
In [14]: df.hist(figsize=(10, 8)) plt.suptitle('Histograms of
Features') plt.show()
```

## Histograms of Features



3) Create a boxplot for each feature in the dataset.

```
In [15]: df.boxplot(figsize=(10, 8)) plt.title('Boxplots of
Features') plt.show()
```



If we observe closely for the box 2, interquartile distance is roughly around 0.75 hence the values lying beyond this range of (third quartile + interquartile distance) i.e. roughly around 4.05 will be considered as outliers. Similarly outliers with other boxplots can be found.