

```
In [2]: #import dataset
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [5]: iris = pd.read_csv("Iris1.csv")
```

analyse and visualize data set

```
In [6]: iris
```

```
Out[6]:
```

| | sepalength | sepalwidth | petallength | petalwidth | class |
|-----|------------|------------|-------------|------------|----------------|
| 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 |
| ... | ... | ... | ... | ... | ... |
| 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 |

150 rows × 5 columns

```
In [7]: iris.head()
```

```
Out[7]:
```

| | sepalength | sepalwidth | petallength | petalwidth | class |
|---|------------|------------|-------------|------------|-------------|
| 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 [8]: len(iris['class'])
```

```
Out[8]: 150
```

```
In [10]: iris.columns
```

```
Out[10]: Index(['sepalength', 'sepalwidth', 'petallength', 'petalwidth', 'class'], dtype='object')
```

```
In [13]: for col in iris.columns:
          print(col)
```

```
sepalength
sepalwidth
petallength
petalwidth
class
```

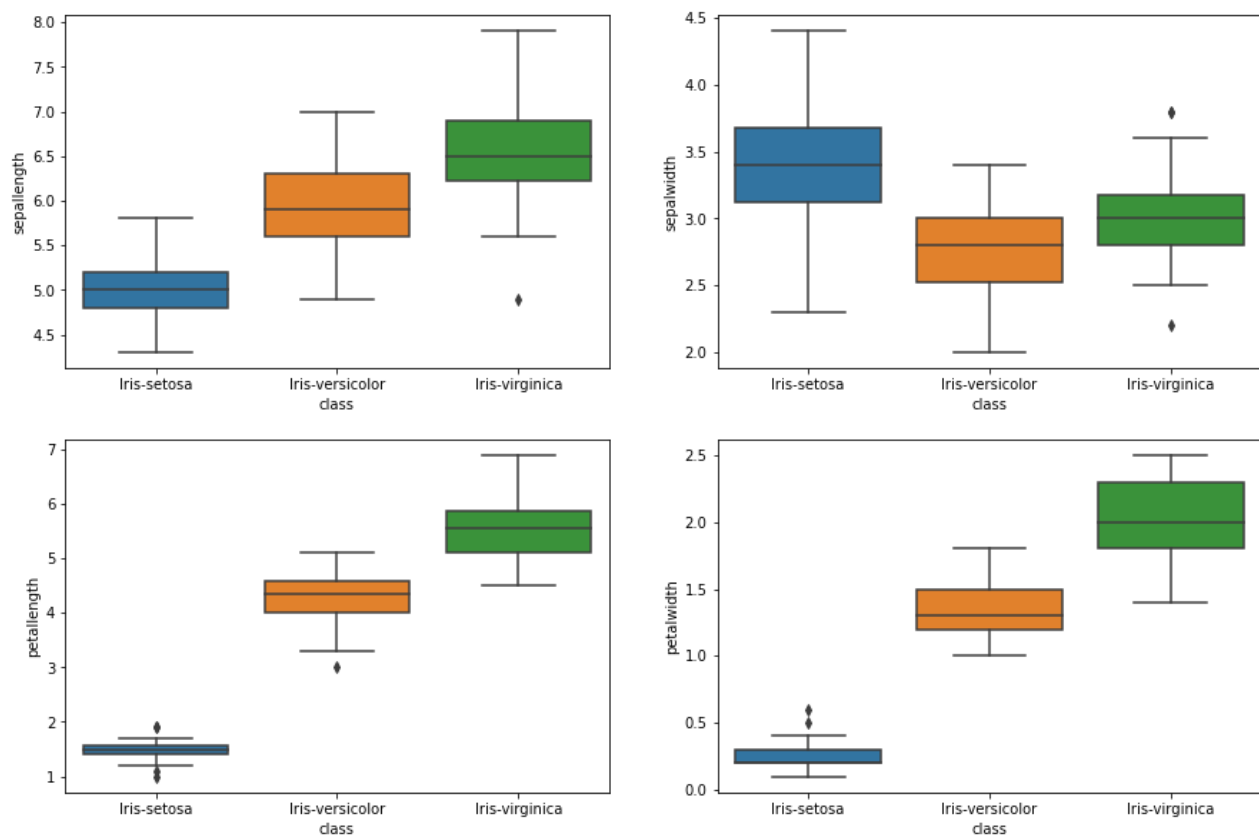
```
In [16]: iris.groupby('class').size()
```

```
Out[16]: class
Iris-setosa      50
Iris-versicolor  50
Iris-virginica   50
dtype: int64
```

```
In [17]: import seaborn as sns
```

```
In [22]: plt.figure(figsize = (15,10))
          plt.subplot(2,2,1)
          sns.boxplot(x = 'class', y = 'sepalength', data = iris)
          plt.subplot(2,2,2)
          sns.boxplot(x = 'class', y = 'sepalwidth', data = iris)
          plt.subplot(2,2,3)
          sns.boxplot(x = 'class', y = 'petallength', data = iris)
          plt.subplot(2,2,4)
          sns.boxplot(x = 'class', y = 'petalwidth', data = iris)
```

```
Out[22]: <AxesSubplot:xlabel='class', ylabel='petalwidth'>
```



```
In [25]: iris.isnull().values.any()
```

```
Out[25]: False
```

```
In [26]: iris.head()
```

```
Out[26]:
```

| | sepal length | sepal width | petal length | petal width | class |
|---|--------------|-------------|--------------|-------------|-------------|
| 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 [27]: x = iris.iloc[ : , :-1].values
         y = iris.iloc[ : , -1 ].values
```

```
In [28]: x
```

```
Out[28]: array([[5.1, 3.5, 1.4, 0.2],
                [4.9, 3. , 1.4, 0.2],
                [4.7, 3.2, 1.3, 0.2],
                [4.6, 3.1, 1.5, 0.2],
                [5. , 3.6, 1.4, 0.2],
                [5.4, 3.9, 1.7, 0.4],
```

```
[4.6, 3.4, 1.4, 0.3],  
[5. , 3.4, 1.5, 0.2],  
[4.4, 2.9, 1.4, 0.2],  
[4.9, 3.1, 1.5, 0.1],  
[5.4, 3.7, 1.5, 0.2],  
[4.8, 3.4, 1.6, 0.2],  
[4.8, 3. , 1.4, 0.1],  
[4.3, 3. , 1.1, 0.1],  
[5.8, 4. , 1.2, 0.2],  
[5.7, 4.4, 1.5, 0.4],  
[5.4, 3.9, 1.3, 0.4],  
[5.1, 3.5, 1.4, 0.3],  
[5.7, 3.8, 1.7, 0.3],  
[5.1, 3.8, 1.5, 0.3],  
[5.4, 3.4, 1.7, 0.2],  
[5.1, 3.7, 1.5, 0.4],  
[4.6, 3.6, 1. , 0.2],  
[5.1, 3.3, 1.7, 0.5],  
[4.8, 3.4, 1.9, 0.2],  
[5. , 3. , 1.6, 0.2],  
[5. , 3.4, 1.6, 0.4],  
[5.2, 3.5, 1.5, 0.2],  
[5.2, 3.4, 1.4, 0.2],  
[4.7, 3.2, 1.6, 0.2],  
[4.8, 3.1, 1.6, 0.2],  
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[5.2, 4.1, 1.5, 0.1],  
[5.5, 4.2, 1.4, 0.2],  
[4.9, 3.1, 1.5, 0.1],  
[5. , 3.2, 1.2, 0.2],  
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[4.4, 3. , 1.3, 0.2],  
[5.1, 3.4, 1.5, 0.2],  
[5. , 3.5, 1.3, 0.3],  
[4.5, 2.3, 1.3, 0.3],  
[4.4, 3.2, 1.3, 0.2],  
[5. , 3.5, 1.6, 0.6],  
[5.1, 3.8, 1.9, 0.4],  
[4.8, 3. , 1.4, 0.3],  
[5.1, 3.8, 1.6, 0.2],  
[4.6, 3.2, 1.4, 0.2],  
[5.3, 3.7, 1.5, 0.2],  
[5. , 3.3, 1.4, 0.2],  
[7. , 3.2, 4.7, 1.4],  
[6.4, 3.2, 4.5, 1.5],  
[6.9, 3.1, 4.9, 1.5],  
[5.5, 2.3, 4. , 1.3],  
[6.5, 2.8, 4.6, 1.5],  
[5.7, 2.8, 4.5, 1.3],  
[6.3, 3.3, 4.7, 1.6],  
[4.9, 2.4, 3.3, 1. ],  
[6.6, 2.9, 4.6, 1.3],  
[5.2, 2.7, 3.9, 1.4],  
[5. , 2. , 3.5, 1. ],  
[5.9, 3. , 4.2, 1.5],  
[6. , 2.2, 4. , 1. ],  
[6.1, 2.9, 4.7, 1.4],  
[5.6, 2.9, 3.6, 1.3],  
[6.7, 3.1, 4.4, 1.4],  
[5.6, 3. , 4.5, 1.5],  
[5.8, 2.7, 4.1, 1. ],  
[6.2, 2.2, 4.5, 1.5],  
[5.6, 2.5, 3.9, 1.1],  
[5.9, 3.2, 4.8, 1.8],
```

```
[6.1, 2.8, 4. , 1.3],  
[6.3, 2.5, 4.9, 1.5],  
[6.1, 2.8, 4.7, 1.2],  
[6.4, 2.9, 4.3, 1.3],  
[6.6, 3. , 4.4, 1.4],  
[6.8, 2.8, 4.8, 1.4],  
[6.7, 3. , 5. , 1.7],  
[6. , 2.9, 4.5, 1.5],  
[5.7, 2.6, 3.5, 1. ],  
[5.5, 2.4, 3.8, 1.1],  
[5.5, 2.4, 3.7, 1. ],  
[5.8, 2.7, 3.9, 1.2],  
[6. , 2.7, 5.1, 1.6],  
[5.4, 3. , 4.5, 1.5],  
[6. , 3.4, 4.5, 1.6],  
[6.7, 3.1, 4.7, 1.5],  
[6.3, 2.3, 4.4, 1.3],  
[5.6, 3. , 4.1, 1.3],  
[5.5, 2.5, 4. , 1.3],  
[5.5, 2.6, 4.4, 1.2],  
[6.1, 3. , 4.6, 1.4],  
[5.8, 2.6, 4. , 1.2],  
[5. , 2.3, 3.3, 1. ],  
[5.6, 2.7, 4.2, 1.3],  
[5.7, 3. , 4.2, 1.2],  
[5.7, 2.9, 4.2, 1.3],  
[6.2, 2.9, 4.3, 1.3],  
[5.1, 2.5, 3. , 1.1],  
[5.7, 2.8, 4.1, 1.3],  
[6.3, 3.3, 6. , 2.5],  
[5.8, 2.7, 5.1, 1.9],  
[7.1, 3. , 5.9, 2.1],  
[6.3, 2.9, 5.6, 1.8],  
[6.5, 3. , 5.8, 2.2],  
[7.6, 3. , 6.6, 2.1],  
[4.9, 2.5, 4.5, 1.7],  
[7.3, 2.9, 6.3, 1.8],  
[6.7, 2.5, 5.8, 1.8],  
[7.2, 3.6, 6.1, 2.5],  
[6.5, 3.2, 5.1, 2. ],  
[6.4, 2.7, 5.3, 1.9],  
[6.8, 3. , 5.5, 2.1],  
[5.7, 2.5, 5. , 2. ],  
[5.8, 2.8, 5.1, 2.4],  
[6.4, 3.2, 5.3, 2.3],  
[6.5, 3. , 5.5, 1.8],  
[7.7, 3.8, 6.7, 2.2],  
[7.7, 2.6, 6.9, 2.3],  
[6. , 2.2, 5. , 1.5],  
[6.9, 3.2, 5.7, 2.3],  
[5.6, 2.8, 4.9, 2. ],  
[7.7, 2.8, 6.7, 2. ],  
[6.3, 2.7, 4.9, 1.8],  
[6.7, 3.3, 5.7, 2.1],  
[7.2, 3.2, 6. , 1.8],  
[6.2, 2.8, 4.8, 1.8],  
[6.1, 3. , 4.9, 1.8],  
[6.4, 2.8, 5.6, 2.1],  
[7.2, 3. , 5.8, 1.6],  
[7.4, 2.8, 6.1, 1.9],  
[7.9, 3.8, 6.4, 2. ],  
[6.4, 2.8, 5.6, 2.2],  
[6.3, 2.8, 5.1, 1.5],  
[6.1, 2.6, 5.6, 1.4],  
[7.7, 3. , 6.1, 2.3],
```

```
[6.3, 3.4, 5.6, 2.4],
[6.4, 3.1, 5.5, 1.8],
[6. , 3. , 4.8, 1.8],
[6.9, 3.1, 5.4, 2.1],
[6.7, 3.1, 5.6, 2.4],
[6.9, 3.1, 5.1, 2.3],
[5.8, 2.7, 5.1, 1.9],
[6.8, 3.2, 5.9, 2.3],
[6.7, 3.3, 5.7, 2.5],
[6.7, 3. , 5.2, 2.3],
[6.3, 2.5, 5. , 1.9],
[6.5, 3. , 5.2, 2. ],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3. , 5.1, 1.8]])
```

In [29]:

 y [illegible]

```
In [32]: #split the dataset
from sklearn.model_selection import train_test_split
array = iris.values
x = array[:,0:4]
y = array[:,4]
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3, random_state=0)
```

```
In [33]: from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
svc = SVC(max_iter=1000, gamma='auto')
svc.fit(x_train, y_train)
y_pred = svc.predict(x_test)
acc_svc = round(accuracy_score(y_pred, y_test) , 2)*100
print("Accuracy : " , acc_svc)
```

Accuracy : 98.0

In []: