

IRIS implementation

```
In [1]: import numpy as np
import pandas as pd      #importing the header files required for our project
import matplotlib.pyplot as plt
```

```
In [3]: data=pd.read_csv("G:/csvfiles/iris.csv") #Loadind the data by using the pandas library
```

```
In [ ]: #understanding the dataa
```

```
In [4]: data.shape
```

```
Out[4]: (150, 5)
```

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

```
Out[5]:
```

| | sepal.length | sepal.width | petal.length | petal.width | variety |
|---|--------------|-------------|--------------|-------------|---------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | Setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | Setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | Setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | Setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | Setosa |

```
In [6]: data.isnull().sum()
```

```
Out[6]: sepal.length    0
sepal.width           0
petal.length          0
petal.width           0
variety               0
dtype: int64
```

```
In [7]: import seaborn as sns
```

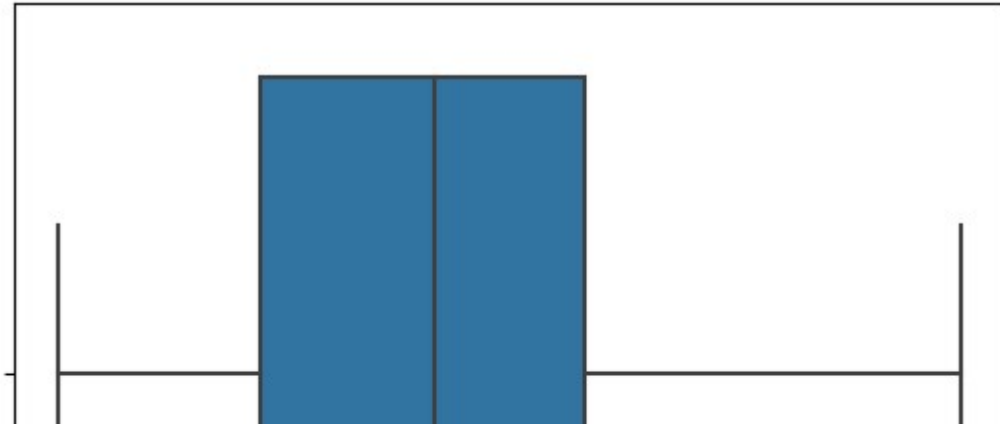
```
In [ ]: #data preprocessing
```

```
In [45]: sns.boxplot(data['sepal.length'])
```

C:\Users\ADMIN\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[45]: <AxesSubplot:xlabel='sepal.length'>
```

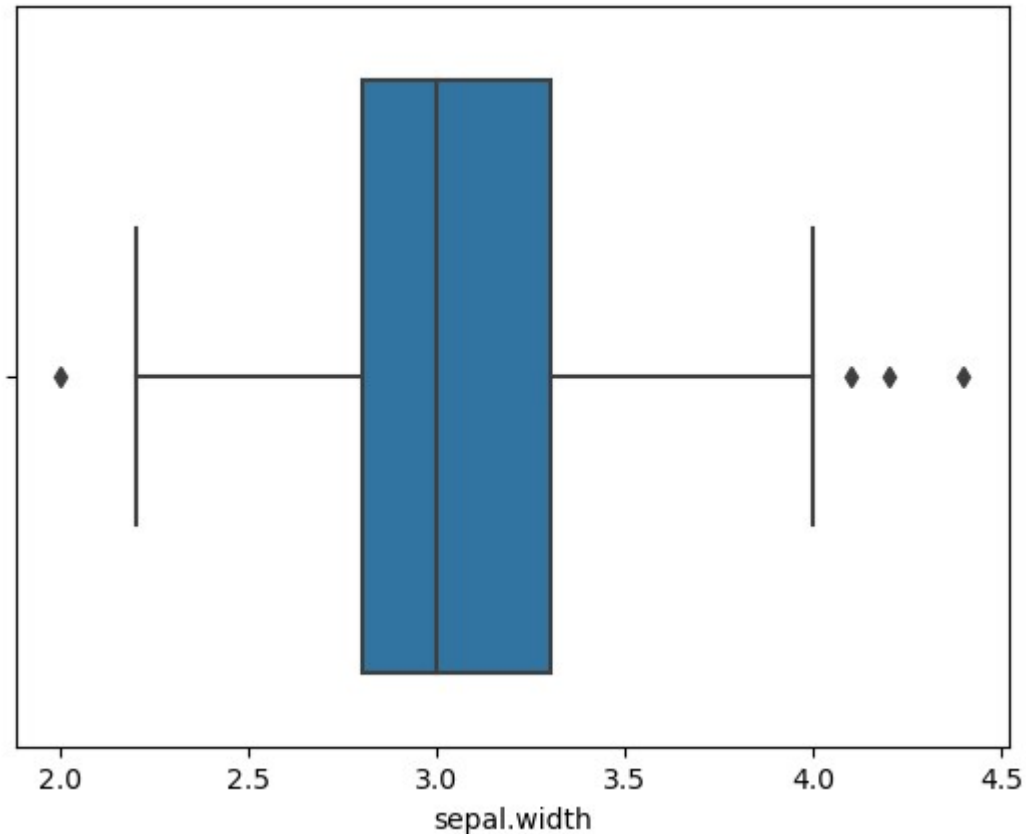


```
In [9]: sns.boxplot(data['sepal.width'])
```

C:\Users\ADMIN\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[9]: <AxesSubplot:xlabel='sepal.width'>
```



```
In [10]: q1=data['sepal.width'].quantile(0.25)
```

```
In [11]: q3=data['sepal.width'].quantile(0.75)
```

```
In [12]: iqr=q3-q1
```

```
In [17]: upper_limit=iqr*1.5+q3  
lower_limit=q1-1.5*iqr
```

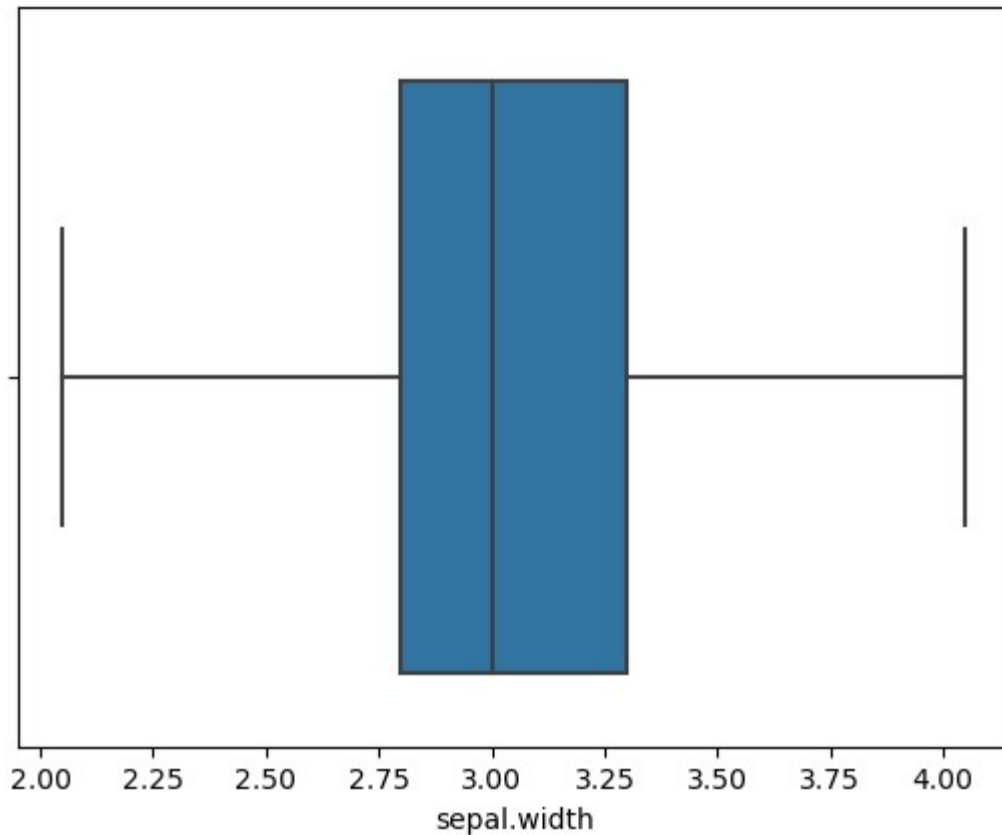
```
In [18]: def Imputation(value):  
        if value > upper_limit:  
            return upper_limit  
        elif value < lower_limit:  
            return lower_limit  
        else:  
            return value
```

```
In [19]: data['sepal.width']=data['sepal.width'].apply(Imputation)
```

```
In [20]: sns.boxplot(data['sepal.width'])
```

C:\Users\ADMIN\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
warnings.warn(

```
Out[20]: <AxesSubplot:xlabel='sepal.width'>
```

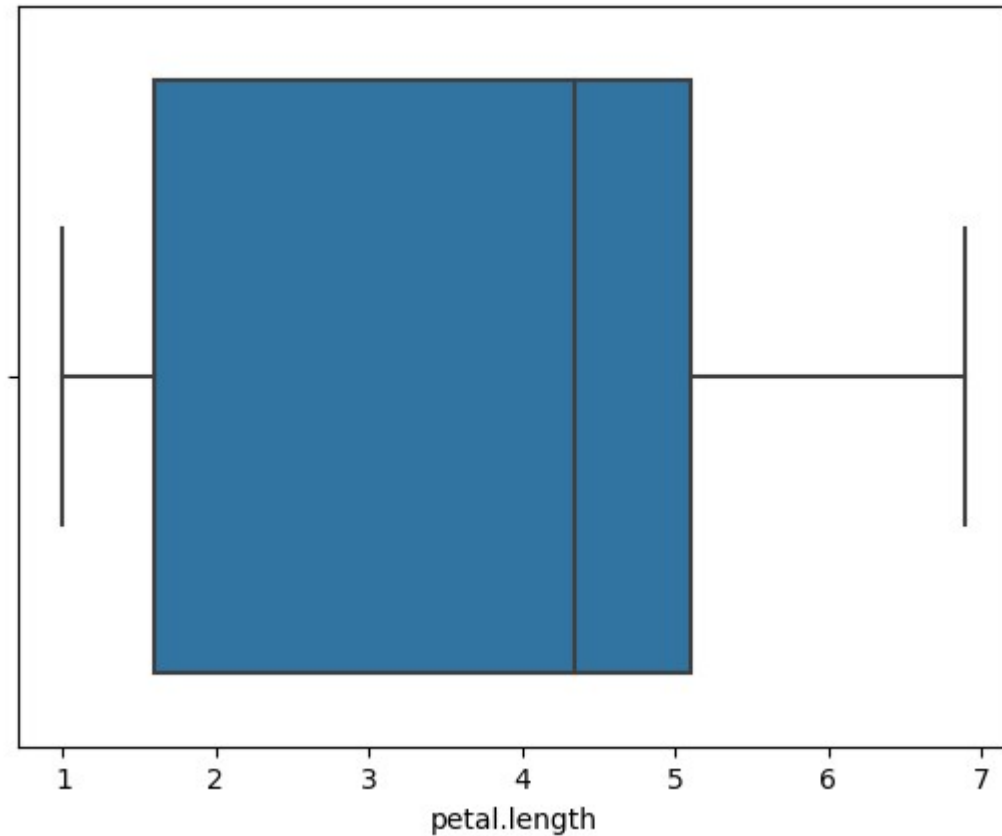


```
In [21]: sns.boxplot(data['petal.length'])
```

C:\Users\ADMIN\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[21]: <AxesSubplot:xlabel='petal.length'>
```

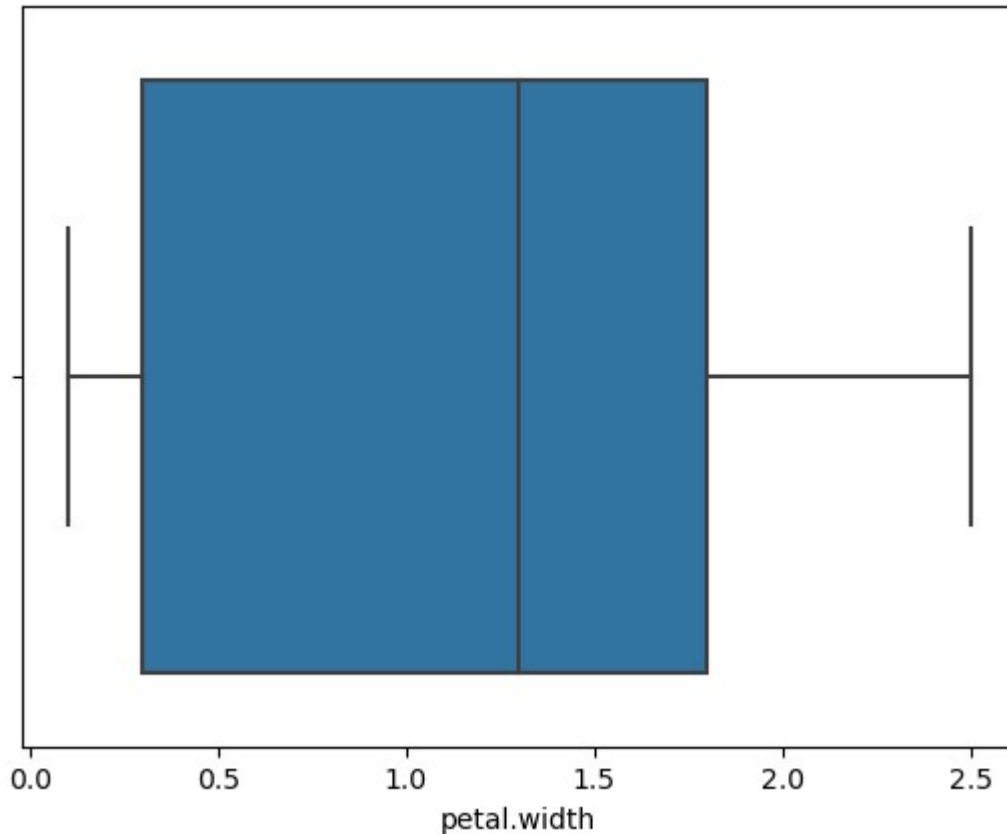


```
In [22]: sns.boxplot(data['petal.width'])
```

C:\Users\ADMIN\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
Out[22]: <AxesSubplot:xlabel='petal.width'>
```



```
In [23]: data['petal.width'].unique()
```

```
Out[23]: array([0.2, 0.4, 0.3, 0.1, 0.5, 0.6, 1.4, 1.5, 1.3, 1.6, 1. , 1.1, 1.8,
                1.2, 1.7, 2.5, 1.9, 2.1, 2.2, 2. , 2.4, 2.3])
```

```
In [25]: data['petal.length'].unique()
```

```
Out[25]: array([1.4, 1.3, 1.5, 1.7, 1.6, 1.1, 1.2, 1. , 1.9, 4.7, 4.5, 4.9, 4. ,
                4.6, 3.3, 3.9, 3.5, 4.2, 3.6, 4.4, 4.1, 4.8, 4.3, 5. , 3.8, 3.7,
                5.1, 3. , 6. , 5.9, 5.6, 5.8, 6.6, 6.3, 6.1, 5.3, 5.5, 6.7, 6.9,
                5.7, 6.4, 5.4, 5.2])
```

```
In [26]: data['sepal.length'].unique()
```

```
Out[26]: array([5.1, 4.9, 4.7, 4.6, 5. , 5.4, 4.4, 4.8, 4.3, 5.8, 5.7, 5.2, 5.5,  
         4.5, 5.3, 7. , 6.4, 6.9, 6.5, 6.3, 6.6, 5.9, 6. , 6.1, 5.6, 6.7,  
         6.2, 6.8, 7.1, 7.6, 7.3, 7.2, 7.7, 7.4, 7.9])
```

```
In [27]: data['sepal.width'].unique()
```

```
Out[27]: array([3.5 , 3. , 3.2 , 3.1 , 3.6 , 3.9 , 3.4 , 2.9 , 3.7 , 4. , 4.05,  
         3.8 , 3.3 , 2.3 , 2.8 , 2.4 , 2.7 , 2.05, 2.2 , 2.5 , 2.6 ])
```

```
In [28]: data.shape
```

```
Out[28]: (150, 5)
```

```
In [29]: from sklearn.tree import DecisionTreeClassifier #importing the model
```

```
In [30]: tree=DecisionTreeClassifier()
```

```
In [32]: from sklearn.model_selection import train_test_split
```

```
In [35]: x=data.iloc[:,0:4]
```

```
In [36]: x.head(5)
```

```
Out[36]:
```

| | sepal.length | sepal.width | petal.length | petal.width |
|---|--------------|-------------|--------------|-------------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 |

```
In [37]: y=data['variety']
```

```
In [38]: x_train,x_test,y_train,y_test=train_test_split(x,y,random_state=42,test_size=0.2) #training the model
```

```
In [39]: tree.fit(x_train,y_train)
```

```
Out[39]: DecisionTreeClassifier()
```

```
In [40]: y_pred=tree.predict(x_test)    #predicting the labels by using model
```

```
In [41]: from sklearn.metrics import accuracy_score
```

```
In [42]: accuracy=(accuracy_score(y_test,y_pred)*100)    #finding the accuracy of our model
```

```
In [43]: print(accuracy)
```

```
100.0
```



```
In [44]: print(y_test,y_pred)
```

```
73    Versicolor
18      Setosa
118   Virginica
78    Versicolor
76    Versicolor
31      Setosa
64    Versicolor
141   Virginica
68    Versicolor
82    Versicolor
110   Virginica
12      Setosa
36      Setosa
9      Setosa
19      Setosa
56    Versicolor
104   Virginica
69    Versicolor
55    Versicolor
132   Virginica
29      Setosa
127   Virginica
26      Setosa
128   Virginica
131   Virginica
145   Virginica
108   Virginica
143   Virginica
45      Setosa
30      Setosa
Name: variety, dtype: object ['Versicolor' 'Setosa' 'Virginica' 'Versicolor' 'Versicolor' 'Setosa'
 'Versicolor' 'Virginica' 'Versicolor' 'Versicolor' 'Virginica' 'Setosa'
 'Setosa' 'Setosa' 'Setosa' 'Versicolor' 'Virginica' 'Versicolor'
 'Versicolor' 'Virginica' 'Setosa' 'Virginica' 'Setosa' 'Virginica'
 'Virginica' 'Virginica' 'Virginica' 'Virginica' 'Setosa' 'Setosa']
```

```
In [ ]:
```