

In [19]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn import preprocessing
from sklearn.model_selection import train_test_split

-----
ModuleNotFoundError                                Traceback (most recent call last)
<ipython-input-19-c21fa305d695> in <module>
      2 import pandas as pd
      3 import matplotlib.pyplot as plt
----> 4 from sklearn.neighbors import KNeighborsClassifier
      5 from sklearn import preprocessing
      6 from sklearn.model_selection import train_test_split
```

ModuleNotFoundError: No module named 'sklearn.neighbors'

In [3]:

```
iris = pd.read_csv("C:/Users/Subham Kumar Singh/Desktop/IRIS.csv")
```

In [7]:

```
iris.tail()
```

Out[7]:

	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 [6]:

```
iris.shape
```

Out[6]:

```
(150, 5)
```

In [8]:

```
iris['species'].value_counts()
```

Out[8]:

```
Iris-setosa      50
Iris-virginica   50
Iris-versicolor  50
Name: species, dtype: int64
```

In [9]:

```
iris.columns
```

Out[9]:

```
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
      'species'],
      dtype='object')
```

In [10]:

```
iris.values
```

Out[10]:

```
array([[5.1, 3.5, 1.4, 0.2, 'Iris-setosa'],
       [4.9, 3.0, 1.4, 0.2, 'Iris-setosa'],
       [4.7, 3.2, 1.3, 0.2, 'Iris-setosa'],
       [4.6, 3.1, 1.5, 0.2, 'Iris-setosa'],
       [5.0, 3.6, 1.4, 0.2, 'Iris-setosa'],
       [5.4, 3.9, 1.7, 0.4, 'Iris-setosa'],
       [4.6, 3.4, 1.4, 0.3, 'Iris-setosa'],
       [5.0, 3.4, 1.5, 0.2, 'Iris-setosa'],
       [4.4, 2.9, 1.4, 0.2, 'Iris-setosa'],
       [4.9, 3.1, 1.5, 0.1, 'Iris-setosa'],
       [5.4, 3.7, 1.5, 0.2, 'Iris-setosa'],
       [4.8, 3.4, 1.6, 0.2, 'Iris-setosa'],
       [4.8, 3.0, 1.4, 0.1, 'Iris-setosa'],
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       [5.7, 4.4, 1.5, 0.4, 'Iris-setosa'],
       [5.4, 3.9, 1.3, 0.4, 'Iris-setosa'],
       [5.1, 3.5, 1.4, 0.3, 'Iris-setosa'],
```

```
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[5.6, 2.7, 4.2, 1.3, 'Iris-versicolor'],
```

```

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[6.5, 3.0, 5.2, 2.0, 'Iris-virginica'],
[6.2, 3.4, 5.4, 2.3, 'Iris-virginica'],
[5.9, 3.0, 5.1, 1.8, 'Iris-virginica']], dtype=object)

```

In [11]:

```
iris.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 [14]:

```
iris.describe(include="all")
```

Out[14]:

	sepal_length	sepal_width	petal_length	petal_width	species
count	150.000000	150.000000	150.000000	150.000000	150
unique	NaN	NaN	NaN	NaN	3
top	NaN	NaN	NaN	NaN	Iris-setosa
freq	NaN	NaN	NaN	NaN	50
mean	5.843333	3.054000	3.758667	1.198667	NaN
std	0.828066	0.433594	1.764420	0.763161	NaN
min	4.300000	2.000000	1.000000	0.100000	NaN
25%	5.100000	2.800000	1.600000	0.300000	NaN
50%	5.800000	3.000000	4.350000	1.300000	NaN
75%	6.400000	3.300000	5.100000	1.800000	NaN
max	7.900000	4.400000	6.900000	2.500000	NaN

In [15]:

```
X=iris.iloc[:, :4]
X.head()
```

Out[15]:

	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 []:

In [16]:

```
Y=iris.iloc[:, -1]
Y.head()
```

Out[16]:

```
0    Iris-setosa
1    Iris-setosa
2    Iris-setosa
3    Iris-setosa
4    Iris-setosa
Name: species, dtype: object
```

In [26]:

```
from sklearn import preprocessing
```

In [27]:

```
X = preprocessing.StandardScaler().fit_transform(X)
X[0:4]
```

Out[27]:

```
array([[ -0.90068117,  1.03205722, -1.3412724 , -1.31297673],
       [-1.14301691, -0.1249576 , -1.3412724 , -1.31297673],
       [-1.38535265,  0.33784833, -1.39813811, -1.31297673],
       [-1.50652052,  0.10644536, -1.2844067 , -1.31297673]])
```

In [28]:

```
from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test=train_test_split( X,Y,test_size=0.3,random_state=1)
Y_test.shape
```

Out[28]:

```
(45,)
```

In [34]:

```
from sklearn.neighbors import KNeighborsClassifier
```

In [35]:

```
knnmodel=KNeighborsClassifier(n_neighbors=3)
knnmodel.fit(X_train,Y_train)
Y_predict1=knnmodel.predict(X_test)
```

In [36]:

```
from sklearn.metrics import accuracy_score
```

In [37]:

```
acc=accuracy_score(Y_test,Y_predict1)
acc
```

Out[37]:

0.9777777777777777

In [41]:

```
from sklearn.metrics import confusion_matrix
```

In [43]:

```
cm=confusion_matrix(Y_test.values,Y_predict1)
cm
```

Out[43]:

```
array([[14,  0,  0],
       [ 0, 18,  0],
       [ 0,  1, 12]], dtype=int64)
```

In [44]:

```
cm1=pd.DataFrame(data=cm,index=['setosa','versicolor','virginica'],columns=['setosa','versicolor','virginica'])
cm1
```

Out[44]:

	setosa	versicolor	virginica
setosa	14	0	0
versicolor	0	18	0
virginica	0	1	12

In [45]:

```
prediction_output=pd.DataFrame(data=[Y_test.values,Y_predict1],index=['Y_test','Y_pedit1'])
```

In [46]:

```
prediction_output.transpose()
```

Out[46]:

	Y_test	Y_pedit1
0	Iris-setosa	Iris-setosa
1	Iris-versicolor	Iris-versicolor
2	Iris-versicolor	Iris-versicolor
3	Iris-setosa	Iris-setosa
4	Iris-virginica	Iris-virginica
5	Iris-versicolor	Iris-versicolor
6	Iris-virginica	Iris-virginica
7	Iris-setosa	Iris-setosa
8	Iris-setosa	Iris-setosa
9	Iris-virginica	Iris-virginica
10	Iris-versicolor	Iris-versicolor
11	Iris-setosa	Iris-setosa
12	Iris-virginica	Iris-virginica
13	Iris-versicolor	Iris-versicolor
14	Iris-versicolor	Iris-versicolor
15	Iris-setosa	Iris-setosa
16	Iris-versicolor	Iris-versicolor
17	Iris-versicolor	Iris-versicolor
18	Iris-setosa	Iris-setosa
19	Iris-setosa	Iris-setosa
20	Iris-versicolor	Iris-versicolor
21	Iris-versicolor	Iris-versicolor
22	Iris-versicolor	Iris-versicolor

```
22 Iris-versicolor Iris-versicolor
23 Y_test Iris-setosa Y_predict1 Iris-setosa
24 Iris-virginica Iris-virginica
25 Iris-versicolor Iris-versicolor
26 Iris-setosa Iris-setosa
27 Iris-setosa Iris-setosa
28 Iris-versicolor Iris-versicolor
29 Iris-virginica Iris-virginica
30 Iris-versicolor Iris-versicolor
31 Iris-virginica Iris-virginica
32 Iris-versicolor Iris-versicolor
33 Iris-virginica Iris-virginica
34 Iris-virginica Iris-virginica
35 Iris-setosa Iris-setosa
36 Iris-versicolor Iris-versicolor
37 Iris-setosa Iris-setosa
38 Iris-versicolor Iris-versicolor
39 Iris-virginica Iris-virginica
40 Iris-virginica Iris-virginica
41 Iris-setosa Iris-setosa
42 Iris-virginica Iris-versicolor
43 Iris-virginica Iris-virginica
44 Iris-versicolor Iris-versicolor
```

```
prediction_output.iloc[0,:].value_counts()
```

```
Iris-versicolor    18
Iris-setosa        14
Iris-virginica     13
Name: Y_test, dtype: int64
```

In [47]:

Out[47]:

In []: