

2.Program to implement k-NN classification using any standard dataset available in the public domain and find the accuracy of the algorithm.

```
#importing the libraries
```

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

```
#read the dataset
```

```
iris=pd.read_csv("iris_csv.csv")
iris.tail()
```

	sepalength	sepalwidth	petallength	petalwidth	class
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

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

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

```
iris.columns
```

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

iris.values

```
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'],
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       [4.8, 3.0, 1.4, 0.1, 'Iris-setosa'],
       [4.3, 3.0, 1.1, 0.1, 'Iris-setosa'],
       [5.8, 4.0, 1.2, 0.2, 'Iris-setosa'],
       [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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       [5.1, 3.8, 1.5, 0.3, 'Iris-setosa'],
       [5.4, 3.4, 1.7, 0.2, 'Iris-setosa'],
       [5.1, 3.7, 1.5, 0.4, 'Iris-setosa'],
       [4.6, 3.6, 1.0, 0.2, 'Iris-setosa'],
       [5.1, 3.3, 1.7, 0.5, 'Iris-setosa'],
       [4.8, 3.4, 1.9, 0.2, 'Iris-setosa'],
       [5.0, 3.0, 1.6, 0.2, 'Iris-setosa'],
       [5.0, 3.4, 1.6, 0.4, 'Iris-setosa'],
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       [5.5, 4.2, 1.4, 0.2, 'Iris-setosa'],
       [4.9, 3.1, 1.5, 0.1, 'Iris-setosa'],
```

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[5.7, 2.5, 5.0, 2.0, 'Iris-virginica'],
[5.8, 2.8, 5.1, 2.4, 'Iris-virginica'],

```
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[6.2, 2.8, 4.8, 1.8, 'Iris-virginica'],  
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[7.2, 3.0, 5.8, 1.6, 'Iris-virginica'],  
[7.4, 2.8, 6.1, 1.9, 'Iris-virginica'],  
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[6.3, 2.8, 5.1, 1.5, 'Iris-virginica'],  
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[6.4, 3.1, 5.5, 1.8, 'Iris-virginica'],  
[6.0, 3.0, 4.8, 1.8, 'Iris-virginica'],  
[6.9, 3.1, 5.4, 2.1, 'Iris-virginica'],  
[6.7, 3.1, 5.6, 2.4, 'Iris-virginica'],  
[6.9, 3.1, 5.1, 2.3, 'Iris-virginica'],  
[5.8, 2.7, 5.1, 1.9, 'Iris-virginica'],  
[6.8, 3.2, 5.9, 2.3, 'Iris-virginica'],  
[6.7, 3.3, 5.7, 2.5, 'Iris-virginica'],  
[6.7, 3.0, 5.2, 2.3, 'Iris-virginica'],  
[6.3, 2.5, 5.0, 1.9, 'Iris-virginica'],  
[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)
```

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

	sepalength	sepalwidth	petallength	petalwidth	class
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

```
x=iris.iloc[:,4]
```

```
x.head()
```

	sepalength	sepalwidth	petallength	petalwidth
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

```
y=iris.iloc[:,-1]
```

```
y.head()
```

```
0  Iris-setosa
```

```
1  Iris-setosa
```

```
2  Iris-setosa
```

```
3  Iris-setosa
```

```
4  Iris-setosa
```

```
Name: class, dtype: object
```

#Data Normalization

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

```
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]])
```

#train test split

```
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
```

```
(45,)
```

#training and predicting

```
knnmodel=KNeighborsClassifier(n_neighbors=3)
```

```
knnmodel.fit(x_train,y_train)
```

```
KNeighborsClassifier(n_neighbors=3)
```

```
KNeighborsClassifier(algorithm='auto',leaf_size=30,metric='minkowski',  
metric_params=None,n_jobs=None,n_neighbors=3,p=2,  
weights='uniform')
```

```
KNeighborsClassifier(n_neighbors=3)
```

```
y_predict1=knnmodel.predict(x_test)
```

```
# Accuracy
```

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

```
acc=accuracy_score(y_test,y_predict1)
```

```
Acc
```

```
0.9777777777777777
```

```
#confusion matrix
```

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

```
cm=confusion_matrix(y_test.values,y_predict1)
```

```
Cm
```

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

```
cm1=pd.DataFrame(data=cm,index=['setosa','versicolor','verginica'],col  
umns=['setosa','versicolor','verginica'])
```

```
cm1
```

	setosa	versicolor	verginica
setosa	14	0	0
versicolor	0	18	0
verginica	0	1	12

```
#output visualization
```

```
prediction_output=pd.DataFrame(data=[y_test.values,y_predict1],inde  
x=['y_test','y_predict1'])
```

```
prediction_output.transpose()
```

```
y_testy_predict1
```

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
23	Iris-setosa	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