

Iris Flower Species Prediction Using KNN

```
In [ ]: #Import Libraries
import numpy as np
import pandas as pd
```

```
In [71]: #Load Dataset
df=pd.read_csv('/home/student/Downloads/Iris.csv')
df
```

```
Out[71]:
```

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|-----|-----|---------------|--------------|---------------|--------------|----------------|
| 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |
| ... | ... | ... | ... | ... | ... | ... |
| 145 | 146 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 147 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 148 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 149 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 150 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

```
In [35]: df.columns
```

```
Out[35]: Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',  
              'Species'],  
              dtype='object')
```

```
In [36]: df.head()
```

```
Out[36]:
```

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|---|----|---------------|--------------|---------------|--------------|-------------|
| 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |

```
In [37]: df.tail()
```

```
Out[37]:
```

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|-----|-----|---------------|--------------|---------------|--------------|----------------|
| 145 | 146 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 147 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 148 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 149 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 150 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

```
In [38]: df.shape
```

```
Out[38]: (150, 6)
```

```
In [39]: #Checking for Null values  
df.isna().sum()
```

```
Out[39]: Id                0  
SepalLengthCm            0  
SepalWidthCm             0  
PetalLengthCm            0  
PetalWidthCm             0  
Species                  0  
dtype: int64
```

```
In [40]: df1=df.drop(['Id'],axis=1)  
df1
```

```
Out[40]:
```

| | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|-----|---------------|--------------|---------------|--------------|----------------|
| 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 |

```
In [72]: #Input Data
x=df1.iloc[:, -1].values
x
```

```
Out[72]: 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],
 [5.2, 3.2, 1.5, 0.2]])
```

```
In [73]: #Output Data
          y=df1.iloc[:, -1].values
          y
```

```
Out[73]: array(['Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',  
                'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',  
                'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',  
                'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa'])
```

```
In [74]: #Split Dataset into Train & Test Datasets  
from sklearn.model_selection import train_test_split  
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.30)
```

```
In [75]: #Feature Scaling- Standard Scaler  
from sklearn.preprocessing import StandardScaler  
scaler = StandardScaler()  
scaler.fit(x_train)  
x_train = scaler.transform(x_train)  
x_test = scaler.transform(x_test)
```

```
In [76]: #KNN model  
from sklearn.neighbors import KNeighborsClassifier  
classifier=KNeighborsClassifier(n_neighbors=7)  
classifier.fit(x_train,y_train)  
y_pred=classifier.predict(x_test)  
y_pred
```

```
Out[76]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica',  
                'Iris-virginica', 'Iris-setosa', 'Iris-virginica',  
                'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa',  
                'Iris-versicolor', 'Iris-setosa', 'Iris-virginica',  
                'Iris-virginica', 'Iris-virginica', 'Iris-versicolor',  
                'Iris-setosa', 'Iris-virginica', 'Iris-setosa', 'Iris-setosa',  
                'Iris-versicolor', 'Iris-virginica', 'Iris-virginica',  
                'Iris-versicolor', 'Iris-setosa', 'Iris-versicolor',  
                'Iris-versicolor', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica',  
                'Iris-virginica', 'Iris-virginica', 'Iris-virginica',  
                'Iris-virginica', 'Iris-virginica', 'Iris-setosa',  
                'Iris-virginica', 'Iris-versicolor', 'Iris-virginica',
```



```

'Iris-versicolor', 'Iris-virginica', 'Iris-virginica',
'Iris-versicolor', 'Iris-setosa', 'Iris-versicolor',
'Iris-versicolor', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica',
'Iris-virginica', 'Iris-virginica', 'Iris-virginica',
'Iris-virginica', 'Iris-virginica', 'Iris-setosa',
'Iris-virginica', 'Iris-versicolor', 'Iris-virginica',
'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
'Iris-versicolor', 'Iris-versicolor', 'Iris-versicolor',
'Iris-virginica'], dtype=object)

```

```

In [69]: print(classifier.predict([[5.7,4.4,1.5,0.4]]))

['Iris-virginica']

```

```

In [79]: #Evaluating Model -Accuracy Score,Confusion Matrix and ConfusionMatrixDisplay
from sklearn.metrics._plot.confusion_matrix import confusion_matrix
from sklearn.metrics import classification_report,accuracy_score,ConfusionMatrixDisplay
labels=[ 'Iris-setosa','Iris-versicolor', 'Iris-virginica']

result=confusion_matrix(y_test,y_pred)
cm=ConfusionMatrixDisplay(result,display_labels=labels)
cm.plot()
score=accuracy_score(y_test,y_pred)
print(result)
print(score)

```

```

[[12  0  0]
 [ 0 13  0]
 [ 0  1 19]]
0.9777777777777777

```

