## Assignment on Classification using KNN.

Build an application to classify an iris flower into its specie using KNN (Iris dataset from Skleam). Display Accuracy score, classification Report & Confusion Matrix).

```
In [1]: import pandas as pd
```

Location of dataset

```
In [2]: url = "https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data"
```

Assign colum names to the dataset

```
In [3]: names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width', 'Class']
```

Read dataset to pandas dataframe

```
In [4]: irisdata = pd.read_csv(url, names=names)
```

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

Out[5]:

|   | 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 [6]: irisdata.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
sepal-length 150 non-null float64
sepal-width 150 non-null float64
petal-length 150 non-null float64
petal-width 150 non-null float64
Class 150 non-null object
dtypes: float64(4), object(1)
memory usage: 5.9+ KB
```

```
In [7]: x=irisdata.iloc[:,0:4]
```

## Assign data from first fifth columns to y variable

```
In [11]: 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)
    print("dimension of x-train:",x_train.shape, "x_test:",x_test.shape)
    y_test
```

dimension of x-train: (105, 4) x\_test: (45, 4)

## Out[11]:

| Class |  |
|-------|--|
|-------|--|

147 Iris-virginica

127 Iris-virginica

86 Iris-versicolor

78 Iris-versicolor

82 Iris-versicolor

60 Iris-versicolor

133 Iris-virginica

1 Iris-setosa

53 Iris-versicolor

61 Iris-versicolor

130 Iris-virginica

25 Iris-setosa

107 Iris-virginica

9 Iris-setosa

74 Iris-versicolor

104 Iris-virginica

115 Iris-virginica

108 Iris-virginica

145 Iris-virginica

68 Iris-versicolor

118 Iris-virginica

46 Iris-setosa

31 Iris-setosa

8 Iris-setosa

90 Iris-versicolor

146 Iris-virginica

114 Iris-virginica

77 Iris-versicolor

81 Iris-versicolor

11 Iris-setosa

76 Iris-versicolor

122 Iris-virginica

84 Iris-versicolor

75 Iris-versicolor

40 Iris-setosa

134 Iris-virginica

predictions = knn.predict(x\_test)

```
Class
           144
                 Iris-virginica
            41
                  Iris-setosa
            43
                  Iris-setosa
                Iris-versicolor
            89
            23
                  Iris-setosa
            54
                Iris-versicolor
             3
                  Iris-setosa
                Iris-versicolor
            48
                  Iris-setosa
In [15]:
          from sklearn.preprocessing import StandardScaler
           scaler = StandardScaler()
           scaler.fit(x_train)
           x_train = scaler.transform(x_train)
           x_test = scaler.transform(x_test)
           from sklearn.neighbors import KNeighborsClassifier
In [16]:
          Initialize the KNN classifier with 3 neighbors
In [17]:
           knn = KNeighborsClassifier(n_neighbors=3)
          Fit the model to the training data
          knn.fit(x_train, y_train.values.ravel())
In [18]:
Out[18]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                       metric_params=None, n_jobs=1, n_neighbors=3, p=2,
                       weights='uniform')
```

In [19]:

| [ 0 2 13]]      | precision | recall | f1-score | support |
|-----------------|-----------|--------|----------|---------|
| Iris-setosa     | 1.00      | 0.92   | 0.96     | 13      |
| Iris-versicolor | 0.84      | 0.94   | 0.89     | 17      |
| Iris-virginica  | 0.93      | 0.87   | 0.90     | 15      |
| avg / total     | 0.92      | 0.91   | 0.91     | 45      |

| In [ ]: |  |  |  |
|---------|--|--|--|
|         |  |  |  |