PRANAVA RAMAN B M S LAB - 09 - SPOT - NAIVE BAYES AND KNN

1. NAIVE BAYES ALGORITHM

```
№ ↑ ↓ ■ … ■
       import numpy as np
       import pandas as pd
       from math import sqrt, exp, pi
       from sklearn.model_selection import train_test_split
       from sklearn import metrics
       data = pd.read_csv('iris.data.csv')
       print('training data set sample')
       print(data.sample(5))
 ... training data set sample
        sepal_length sepal_width petal_length petal_width
                                                                Class name
                          2.8
    133
                 6.3
                                     5.1 1.5 Iris-virginica
                             3.2
    50
                 7.0
                                          4.7
                                                       1.4 Iris-versicolor
    29
                 4.7
                             3.2
                                          1.6
                                                      0.2
                                                               Iris-setosa
    3
                 4.6
                             3.1
                                          1.5
                                                      0.2
                                                                Iris-setosa
    92
                 5.8
                             2.6
                                          4.0
                                                      1.2 Iris-versicolor
                                                                                                          № ↑ ↓ ■ … ■
       def labels_to_num(data):
           # convert strings to numbers
           data.replace(['Iris-versicolor','Iris-setosa','Iris-virginica'], [0,1,2], inplace=True)
       print('\nreplaced labels with numbers')
       labels to num(data)
       print(data.sample(5))
       # split data by class
       def split_by_class(data):
           # split data by class
           classes = dict()
           for row in data:
               if(row[-1] not in classes.keys()):
                  classes[row[-1]] = list()
           classes[row[-1]].append(row)
           return classes
 [4]
                                                                                                                         Python
    replaced labels with numbers
        sepal_length sepal_width petal_length petal_width Class_name
    9
                 4.9
                            3.1
                                     1.5
                                                0.1
                                                                   1
    11
                 4.8
                             3.4
                                          1.6
                                                       0.2
                                                                     1
    89
                 5.5
                             2.5
                                          4.0
                                                       1.3
                                                                    0
    119
                 6.0
                             2.2
                                          5.0
                                                      1.5
    42
                 4.4
                             3.2
                                          1.3
                                                       0.2
                                                                    1
                                                                                                          № ↑ ↓ ■ … ■
      # get mean, std and size
      def get_info(data):
          # get data
          info = [(np.mean(col), np.std(col), len(col)) for col in zip(*data)]
          del info[-1] # remove target coloumn
          return info
      # get mean, std and size by class
      def get_info_by_class(data):
          classData = split_by_class(data)
          info = dict()
          for classVal, rows in classData.items():
              info[classVal] = get_info(rows)
          return info
[10] 		 0.1s
                                                                                                                         Python
```

```
№ ↑ ↓ ■ … •
        def calc_prob(x, mean, stdev):
    exponent = exp(-((x-mean)**2 / (2 * stdev**2 )))
    return (1 / (sqrt(2 * pi) * stdev)) * exponent
         def predict(info, dataset):
             outProbs = list()
             for data in dataset:
                 total_rows = sum([info[label][0][2] for label in info])
                 probs = dict()
                  for classVal, class_info in info.items():
                      probs[classVal] = info[classVal][0][2]/float(total_rows)
for i in range(len(class_info)):
                           mean, stdev, _ = class_info[i]
probs[classVal] *= calc_prob(data[i], mean, stdev)
                  outProbs.append(probs)
             preds = list()
             for prob in outProbs:
                 preds.append(max(prob, key=prob.get))
             return preds
                                                                                                                                                     Python
        # split training and testing data
        X_train, X_test = train_test_split(data,test_size=0.2)
        info = get_info_by_class(X_train.values)
        y_test = X_test.iloc[:,-1].values
        X_{\text{test}} = X_{\text{test.iloc}}[:,0:-1]
        preds = predict(info, X_test.values)
        print('accuracy', metrics.accuracy_score(y_test,preds))
        print('confusion matrix')
        print(metrics.confusion_matrix(y_test, preds))
        print('classification report')
        print(metrics.classification_report(y_test, preds))
[12] 🗸 0.1s
                                                                                                                                                     Python
... accuracy 0.9
```

2. KNN CLASSIFIER

```
№ ↑ ↓ ■ … •
       import numpy as nm
       import matplotlib.pyplot as mtp
       from sklearn.preprocessing import LabelEncoder
       import math
       import pandas as pd
       le = LabelEncoder()
       df= pd.read_csv('iris.data.csv')
       data_set = df.apply(le.fit_transform)
x= data_set.iloc[:, :-1].values
       y= data_set.iloc[:, -1].values
       from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 0.25)
       from sklearn.preprocessing import StandardScaler
       st_x= StandardScaler()
       x_train= st_x.fit_transform(x_train)
       x_test= st_x.transform(x_test)
                                                                                                                                       Python
       from sklearn.neighbors import KNeighborsClassifier
       classifier= KNeighborsClassifier(n_neighbors=3, metric='minkowski', p=2 )
       classifier.fit(x_train, y_train)
... KNeighborsClassifier(n_neighbors=3)
       y_pred= classifier.predict(x_test)
[40] 🗸 0.9s
                                                                                                                                       Python
> ~
       from sklearn.metrics import confusion_matrix
       cm= confusion_matrix(y_test, y_pred)
[42] 🗸 0.6s
                                                                                                                                       Python
[43] 🗸 0.2s
                                                                                                                                       Python
 ... array([[15, 0, 0],
           [ 0, 11, 0],
            [ 0, 3, 9]], dtype=int64)
        from sklearn.metrics import classification_report
        matrix = classification_report(y_test,y_pred,labels=[0,1])
        print('\nClassification report : \n',matrix)
     ✓ 0.1s
[44]
     Classification report :
                                recall f1-score support
                    precision
                0
                         1.00
                                   1.00
                                              1.00
                1
                         0.79
                                   1.00
                                              0.88
                                                           11
                                  1.00
        micro avg
                         0.90
                                              0.95
                                                           26
        macro avg
                         0.89
                                   1.00
                                              0.94
                                                           26
     weighted avg
                         0.91
                                   1.00
                                              0.95
                                                           26
```