

- \* Write a program to implement the naive Bayesian classifier for a simple training dataset stored as a .csv file. Compute the accuracy of the classifier, considering few test dataset.

```
*
import csv, random, math
import statistics as st
def loadcsv(filename):
    line = csv.reader(open(filename, "r"))
    dataset = list(line)
    for i in range(len(dataset)):
        dataset[i] = [float(x) for x in dataset[i]]
    return dataset
```

```
def splitDataset (dataset, splitRatio):
    testsize = int (len (dataset) * splitRatio);
    trainset = list (dataset)
    testset = []
    while len(testset) < testsize:
        index = random.randrange (len (trainset))
        testset.append (trainset.pop(index))
    return (trainset, testset)
```

```
def separateByClass (dataset):
    separated = {}
    for i in range (len (dataset)):
        x = dataset[i]
```

Teacher's Signature \_\_\_\_\_

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if  $x[-1]$  not in Separated:Separated  $[x[-1]] = []$ Separated  $[x[-1]].append(x)$ 

return Separated

def compute\_mean\_std(dataset):

mean\_std = [(st.mean(attribute), st.stdev(attribute))

for attribute in zip(\*dataset)]

del mean\_std[-1]

return mean\_std

def Summarize\_Byclass(dataset):

Separated = separate\_Byclass(dataset),

Summary = {}

for class\_value, instance in Separated.items():

Summary[class\_value] = compute\_mean\_std(instance)

return Summary.

def estimate\_Probability(x, mean, stdev):

exponent = math.exp(-(math.pow(x - mean, 2) / (2 \* math.  
pow(stdev, 2))))

def CalculateClassProbabilities(Summaries, testvector):

p = {}

for class\_value, class\_Summarise in Summaries.items():

p[class\_value] = 1

for i in range(len(class\_Summarises)):

mean, stdev = class\_Summarises[i],

x = testvector[i]

p[class\_value] \*= estimate\_Probability(x, mean, stdev)

return p

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```

def predict(Summaries, testVector):
    all_p = calculateClassProbabilities(Summaries, testVector)
    bestLabel, bestProb = None, -1
    for lbl, p in all_p.items():
        if bestLabel is None or p > bestProb:
            bestProb = p
            bestLabel = lbl
    return bestLabel.

def performClassification(Summaries, testset):
    predictions = []
    for i in range(len(testset)):
        result = predict(Summaries, testset[i])
        predictions.append(result)
    return predictions.

def getAccuracy(testset, predictions):
    correct = 0
    for i in range(len(testset)):
        if testset[i][-1] == predictions[i]:
            correct += 1
    return (correct / float(len(testset))) * 100.0

```

```

dataset = load_csv('C:/Users/Mili/OneDrive/Desktop/Mili/AMT17CS058-
Mihir/diabetes.csv')
print('Pima Indian Diabetes Data Set loaded...')
print('Total instance available:', len(dataset))
print('Total attribute present:', len(dataset[0])-1)
print('First Five instance of data set')
for i in range(5):
    print(i+1, ':', dataset[i])

```

Teacher's Signature \_\_\_\_\_

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Split Ratio = 0.2

trainingset, testset = splitDataset (dataset, splitRatio)

print ("In Dataset is split into training set and testing set").

print ("Training example = {0} In Testing example = {1} : format  
(len(trainingset), len(testset)))

summaries = SummariseByClass (trainingset);

prediction = perform Classification (summaries, testset)

accuracy = get Accuracy (testset, predictions)

print ("In Accuracy of the Naive Bayesian classifier is :  
accuracy)

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## Output

Pima Indian Diabetes Dataset loaded ...

Total instance available : 768

Total Attribute present : 8

First Five instance of dataset :

1: [6.0, 148.0, 72.0, 35.0, 0.0, 33.6, 0.627, 50.0, 1.0]

2: [1.0, 85.0, 66.0, 29.0, 0.0, 26.6, 0.351, 31.0, 0.0]

3: [8.0, 183.0, 64.0, 0.0, 0.0, 23.3, 0.674, 32.0, 1.0]

4: [1.0, 89.0, 66.0, 23.0, 94.0, 28.1, 0.167, 21.0, 0.0]

5: [0.0, 137.0, 42.0, 35.0, 168.0, 45.1, 2.288, 33.0, 1.0]

Dataset is Split into training and testing set

Training example = 615

Testing example = 153

Accuracy of the Naive Bayesian Classifier is

75.16339869281046.