VISVESVARAYA TECHNOLOGICAL UNIVERSITY

"JnanaSangama", Belgaum -590014, Karnataka.



LAB REPORT on

MACHINE LEARNING (20CS6PCMAL)

Submitted by

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in partial fulfillment for the award of the degree of BACHELOR OF ENGINEERING
in
COMPUTER SCIENCE AND ENGINEERING



B.M.S. COLLEGE OF ENGINEERING
(Autonomous Institution under VTU)
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B. M. S. College of Engineering,

Bull Temple Road, Bangalore 560019(Affiliated To Visvesvaraya Technological University, Belgaum)

Department of Computer Science and Engineering



CERTIFICATE

This is to certify that the Lab work entitled "MACHINE LEARNING" carried out by SAURAB V MOTAGI (1BM19CS145), who is bonafide student of B. M. S. College of Engineering. It is in partial fulfillment for the award of Bachelor of Engineering in Computer Science and Engineering of the Visvesvaraya Technological University, Belgaum during the year 2022. The Lab report has been approved as it satisfies the academic requirements in respect of a Machine Learning- (20CS6PCMAL) work prescribed for the said degree.

Name of the Lab-Incharge Designation Department of CSE BMSCE, Bengaluru **Prof.Saritha.A.N**Assistant Professor
Department of CSE
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LAB PROGRAM 1:

Implement and demonstrate the FIND-S algorithm for finding the most specific hypothesis based on a given set of training data samples.

Code:

```
i.
 import
 csv
          def updateHypothesis(x,h):
              if h==[]:
                  return x
              for i in range(0,len(h)):
                  if x[i].upper()!=h[i].upper():
                      h[i] = '?'
              return h
          if __name__ == "__main__":
              data = []
              h = []
              # reading csv file
              with open('data.csv', 'r') as file:
                  reader = csv.reader(file)
                  print("Data: ")
                  for row in reader:
                      data.append(row)
                      print(row)
              if data:
                  for x in data:
                      if x[-1].upper()=="YES":
                          x.pop() # removing last field
```

```
h = updateHypothesis(x,h)
```

print("\nHypothesis: ",h)

ii.

```
In [1]: import pandas as pd
         import numpy as np
In [2]: n=int(input("Enter number of rows:"))
         columns=['Time','Weather','Temperature','humidity','Enjoying?']
         print("Enter the data:\n")
         for i in range(n):
             print("Enter Hypothesis:",i+1,"\n")
             temp=[]
             for x in columns:
                 t=input("Enter value for: "+x+": ")
                 temp.append(t)
             d.append(temp)
        Enter number of rows:3
        Enter the data:
        Enter Hypothesis: 1
        Enter value for: Time: eve
        Enter value for: Weather: sunny
        Enter value for: Temperature: warm
        Enter value for: humidity: mild
        Enter value for: Enjoying?: yes
        Enter Hypothesis: 2
        Enter value for: Time: eve
        Enter value for: Weather: rainy
        Enter value for: Temperature: cold
        Enter value for: humidity: less
        Enter value for: Enjoying?: yes
        Enter Hypothesis: 3
        Enter value for: Time: eve
        Enter value for: Weather: sunny
        Enter value for: Temperature: warm
        Enter value for: humidity: mild
        Enter value for: Enjoying?: no
In [ ]: for x in d:
             print(x)
```

```
Enter value for: Time: eve
            Enter value for: Weather: rainy
            Enter value for: Temperature; cold
Enter value for: humidity: less
            Enter value for: Enjoying?: yes
            Enter Hypothesis: 3
            Enter value for: Time: eve
Enter value for: Weather: sunny
            Enter value for: Temperature: warm
            Enter value for: humidity: mild
            Enter value for: Enjoying?: no
In [ ]: for x in d:
                 print(x)
             hypo=[]
            nypo=||
for i in range(len(d[0])):
    hypo.append("?")
for i in range(len(d)):
    if d[i][len(d[0])-1]=='yes':
                       hypo=d[i]
In [ ]: for i in range(len(d)):
                  if d[i][len(d[0])-1]=='yes':
                      for j in range(len(d[0])):
    if(d[i][j]!=hypo[j]):
        hypo[j]="?"
In [ ]: print(hypo)
```

1	sky	air temp	humidity	wind	water	forecast	enjoy sport
2	sunny	warm	normal	strong	warm	same	yes
3	sunny	warm	high	strong	warm	same	yes
4	rainy	cold	high	strong	warm	change	no
5	sunny	warm	high	strong	cool	change	yes

LAB PROGRAM 2:

For a given set of training data examples stored in a .CSV file, implement and demonstrate the Candidate-Elimination algorithm to output a description of the set of all hypotheses consistent with the training examples.

CODE:

```
import numpy as np
import pandas as pd
data = pd.read csv("testdemo.csv")
concepts = np.array(data.iloc[:,0:-1])
print("\nInstances are:\n",concepts)
target = np.array(data.iloc[:,-1])
print("\nTarget Values are: ",target)
def learn(concepts, target):
    specific h = concepts[0].copy()
   print("\nSpecific Boundary: ", specific h)
    general h = [["?" for i in range(len(specific h))] for i in
range(len(specific h))]
   print("\nGeneric Boundary: ",general h)
    for i, h in enumerate(concepts):
        print("\nInstance", i+1 , "is ", h)
        if target[i] == "yes":
            for x in range(len(specific h)):
                if h[x]!= specific h[x]:
                    specific h[x] ='?'
                    general h[x][x] = "?"
        if target[i] == "no":
            for x in range(len(specific h)):
                if h[x]!= specific h[x]:
                    general h[x][x] = specific h[x]
                else:
                    general h[x][x] = '?'
        print("Specific Boundary = ", specific h)
        print("Generic Boundary = ", general h)
        print("\n")
    indices = [i for i, val in enumerate(general h) if val == ['?', '?', '?',
'?', '?', '?']]
    for i in indices:
```

```
general_h.remove(['?', '?', '?', '?', '?', '?'])
return specific_h, general_h

s_final, g_final = learn(concepts, target)

print(" The Final Specific_h : ", s_final, sep="\n")
print("The Final General_h : ", g_final, sep="\n")
```

1	sky	airtemp	humidity	wind	water	forecast	enjoysport	
2	sunny	warm	normal	strong	warm	same	yes	
3	sunny	warm	high	strong	warm	same	yes	
4	rainy	cold	high	strong	warm	change	no	
5	sunny	warm	high	strong	cool	change	yes	

```
In [1]: import numpy as np
           import pandas as pd
           data = pd.read_csv("testdemo.csv")
           concepts = np.array(data.iloc[:,0:-1])
           print("\nInstances are:\n",concepts)
           target = np.array(data.iloc[:,-1])
           print("\nTarget Values are: ",target)
          Instances are:
           Instances are:
[['sunny' warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]
         Target Values are: ['yes' 'yes' 'no' 'yes']
In [1]: def learn(concepts, target):
               specific_h = concepts[0].copy()
                print("\nSpecific Boundary: ", specific_h)
                general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
               print("\nGeneric Boundary: ",general_h)
                for i, h in enumerate(concepts):
                   print("\nInstance", i+1 , "is ", h)
                    if target[i] == "yes":
                        for x in range(len(specific_h)):
                            if h[x]!= specific_h[x]:
                                  specific_h[x] ='?
                                 general_h[x][x] ='?'
                    if target[i] == "no":
                         for x in range(len(specific_h)):
                             if h[x]!= specific_h[x]:
                                  general\_h[x][x] = specific\_h[x]
                              else:
                                  general_h[x][x] = '?'
                    print("Specific Boundary = ", specific_h)
print("Generic Boundary = ", general_h)
                    print("\n")
                indices = [i for i, val in enumerate(general_h) if val == ['?', '?', '?', '?', '?', '?']]
```

```
In [1]: def learn(concepts, target):
             specific_h = concepts[0].copy()
             print("\nSpecific Boundary: ", specific_h)
             general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
             print("\nGeneric Boundary: ",general_h)
             for i, h in enumerate(concepts):
                 print("\nInstance", i+1 , "is ", h)
                 if target[i] == "yes":
                     for x in range(len(specific_h)):
                         if h[x]!= specific_h[x]:
                            specific_h[x] ='?'
                             general_h[x][x] = ?
                 if target[i] == "no":
                     for x in range(len(specific_h)):
                         if h[x]!= specific_h[x]:
                            general_h[x][x] = specific_h[x]
                         else:
                             general_h[x][x] = '?'
                 print("Specific Boundary = ", specific_h)
                 print("Generic Boundary = ", general_h)
                 print("\n")
             indices = [i for i, val in enumerate(general_h) if val == ['?', '?', '?', '?', '?', '?']]
             for i in indices:
                 general_h.remove(['?', '?', '?', '?', '?', '?'])
             return specific_h, general_h
         s_final, g_final = learn(concepts, target)
         print(" The Final Specific_h : ", s_final, sep="\n")
         print("The Final General_h : ", g_final, sep="\n")
```

LAB PROGRAM 3:

Write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample.

Code:

```
import
math
         import csv
         def load_csv(filename):
             lines=csv.reader(open(filename, "r"));
             dataset = list(lines)
             headers = dataset.pop(0)
             return dataset, headers
         class Node:
             def __init__(self,attribute):
                 self.attribute=attribute
                 self.children=[]
                 self.answer=""
         def subtables(data,col,delete):
             dic={}
             coldata=[row[col] for row in data]
             attr=list(set(coldata))
             counts=[0]*len(attr)
             r=len(data)
             c=len(data[0])
             for x in range(len(attr)):
                 for y in range(r):
                     if data[y][col]==attr[x]:
                         counts[x]+=1
             for x in range(len(attr)):
                 dic[attr[x]]=[[0 for i in range(c)] for j in range(counts[x])]
                 pos=0
```

```
for y in range(r):
            if data[y][col]==attr[x]:
                if delete:
                    del data[y][col]
                dic[attr[x]][pos]=data[y]
                pos+=1
    return attr,dic
def entropy(S):
    attr=list(set(S))
    if len(attr)==1:
        return 0
    counts=[0,0]
    for i in range(2):
        counts[i]=sum([1 for x in S if attr[i]==x])/(len(S)*1.0)
    sums=0
    for cnt in counts:
        sums+=-1*cnt*math.log(cnt,2)
    return sums
def compute_gain(data,col):
    attr,dic = subtables(data,col,delete=False)
    total_size=len(data)
    entropies=[0]*len(attr)
    ratio=[0]*len(attr)
    total_entropy=entropy([row[-1] for row in data])
    for x in range(len(attr)):
        ratio[x]=len(dic[attr[x]])/(total_size*1.0)
        entropies[x]=entropy([row[-1] for row in dic[attr[x]]])
        total_entropy-=ratio[x]*entropies[x]
    return total_entropy
def build_tree(data,features):
    lastcol=[row[-1] for row in data]
    if(len(set(lastcol)))==1:
        node=Node("")
        node.answer=lastcol[0]
        return node
```

```
n=len(data[0])-1
    gains=[0]*n
    for col in range(n):
        gains[col]=compute_gain(data,col)
    split=gains.index(max(gains))
    node=Node(features[split])
    fea = features[:split]+features[split+1:]
    attr,dic=subtables(data,split,delete=True)
    for x in range(len(attr)):
        child=build_tree(dic[attr[x]],fea)
        node.children.append((attr[x],child))
    return node
def print_tree(node,level):
    if node.answer!="":
        print(" "*level, node.answer)
        return
    print(" "*level, node.attribute)
    for value, n in node.children:
        print(" "*(level+1), value)
        print_tree(n,level+2)
def classify(node,x_test,features):
    if node.answer!="":
        print(node.answer)
        return
    pos=features.index(node.attribute)
    for value, n in node.children:
        if x_test[pos]==value:
            classify(n,x_test,features)
'''Main program'''
dataset,features=load_csv("id3.csv")
node1=build_tree(dataset,features)
print("The decision tree for the dataset using ID3 algorithm is")
print_tree(node1,0)
testdata,features=load csv("id3 test 1.csv")
```

```
for xtest in testdata:
    print("The test instance:",xtest)
    print("The label for test instance:")
    classify(node1,xtest,features)
```

```
In [1]: import math
         import csv
In [2]: def load_csv(filename):
             lines=csv.reader(open(filename, "r"));
             dataset = list(lines)
             headers = dataset.pop(0)
             return dataset, headers
         class Node:
             def __init__ (self,attribute):
                 self.attribute=attribute
                 self.children=[]
                 self.answer=""
In [3]: def subtables(data,col,delete):
             coldata=[row[col] for row in data]
             attr=list(set(coldata))
             counts=[0]*len(attr)
             r=len(data)
             c=len(data[0])
             for x in range(len(attr)):
                 for y in range(r):
                     if data[y][col]==attr[x]:
                        counts[x]+=1
             for x in range(len(attr)):
                 dic[attr[x]]=[[0 for i in range(c)] for j in range(counts[x])]
                 for y in range(r):
                     if data[y][col]==attr[x]:
                         if delete:
                            del data[y][col]
                         dic[attr[x]][pos]=data[y]
             return attr,dic
```

```
In [4]: def entropy(S):
            attr=list(set(S))
             if len(attr)==1:
                 return 0
             counts=[0,0]
             for i in range(2):
                counts[i]=sum([1 for x in S if attr[i]==x])/(len(S)*1.0)
             for cnt in counts:
                sums+=-1*cnt*math.log(cnt,2)
             return sums
In [5]: def compute_gain(data,col):
             attr,dic = subtables(data,col,delete=False)
             total_size=len(data)
             entropies=[0]*len(attr)
             ratio=[0]*len(attr)
             total_entropy=entropy([row[-1] for row in data])
             for x in range(len(attr)):
                ratio[x]=len(dic[attr[x]])/(total_size*1.0)
                 entropies[x]=entropy([row[-1] for row in dic[attr[x]]])
                 total_entropy==ratio[x]*entropies[x]
             return total_entropy
In [6]: def build_tree(data,features):
            lastcol=[row[-1] for row in data]
             if(len(set(lastcol)))==1:
                 node=Node("")
                 node.answer=lastcol[0]
                 return node
             n=len(data[0])-1
             gains=[0]*n
             for col in range(n):
                 gains[col]=compute_gain(data,col)
             split=gains.index(max(gains))
             node=Node(features[split])
             fea = features[:split]+features[split+1:]
```

```
for col in range(n):
                 gains[col]=compute_gain(data,col)
             split=gains.index(max(gains))
             node=Node(features[split])
             fea = features[:split]+features[split+1:]
             attr,dic=subtables(data,split,delete=True)
             for x in range(len(attr)):
                 child=build_tree(dic[attr[x]],fea)
                 node.children.append((attr[x],child))
             return node
In [7]: def print_tree(node,level):
             if node.answer!="":
                 print(" "*level, node.answer)
                 return
             print(" "*level, node.attribute)
             for value, n in node.children:
                 print(" "*(level+1),value)
                 print_tree(n,level+2)
In [8]: def classify(node,x_test,features):
             if node.answer!="":
                 print(node.answer)
                 return
             pos=features.index(node.attribute)
             for value, n in node children:
                 if x_test[pos]==value:
                     classify(n,x_test,features)
In [9]: ''Main program'''
         dataset, features=load_csv("id3.csv")
         nodel=build_tree(dataset,features)
         print("The decision tree for the dataset using ID3 algorithm is")
         print_tree(node1,0)
         testdata, features=load_csv("id3_test_1.csv")
         for xtest in testdata:
            print("The test instance:",xtest)
             print("The label for test instance:",end=" ")
             classifv(node1.xtest.features)
```

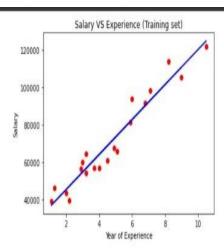
```
In [9]: '''Main program'''
         dataset,features=load_csv("id3.csv")
         node1=build_tree(dataset,features)
         print("The decision tree for the dataset using ID3 algorithm is")
         print_tree(node1,0)
         testdata,features=load_csv("id3_test_1.csv")
         for xtest in testdata:
            print("The test instance:",xtest)
            print("The label for test instance:",end=" ")
            classify(node1,xtest,features)
        The decision tree for the dataset using ID3 algorithm is
         Outlook
           overcast
             yes
           sunny
             Humidity
              high
                 no
               normal
                 yes
           rain
             Wind
              strong
                 no
               weak
                 yes
        The test instance: ['rain', 'cool', 'normal', 'strong']
        The label for test instance: no
        The test instance: ['sunny', 'mild', 'normal', 'strong']
        The label for test instance: yes
In [ ]:
```

	Outlook	Temperature	Humidity	Wind	Answer
2	sunny	hot	high	weak	no
3	sunny	hot	high	strong	no
	overcast	hot	high	weak	yes
5	rain	mild	high	weak	yes
6	rain	cool	normal	weak	yes
7	rain	cool	normal	strong	no
8	overcast	cool	normal	strong	yes
9	sunny	mild	high	weak	no
0	sunny	cool	normal	weak	yes
1	rain	mild	normal	weak	yes
2	sunny	mild	normal	strong	yes
3	overcast	mild	high	strong	yes
4	overcast	hot	normal	weak	yes
5	rain	mild	high	strong	no

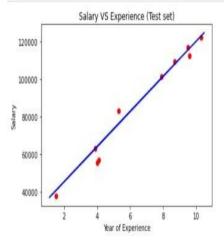
LAB PROGRAM 4:

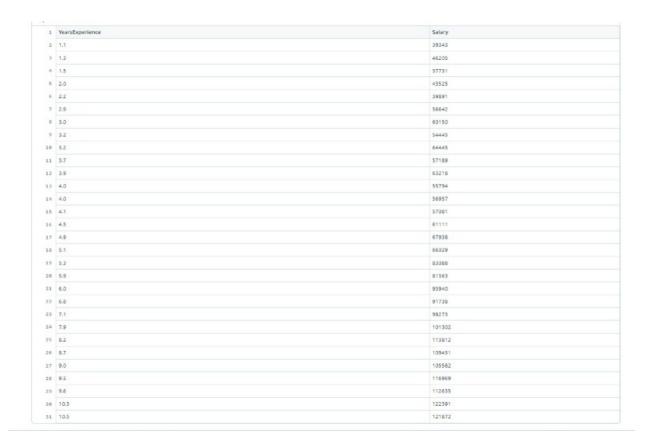
Implement the Linear Regression algorithm in order to fit data points. Select appropriate data set for your experiment and draw graphs.

```
import numpy as np
          import matplotlib.pyplot as plt
          import pandas as pd
          dataset = pd.read_csv('salary_data.csv')
          X = dataset.iloc[:, :-1].values
          y = dataset.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=1/3, random_state=0)
In [11]: # Fitting Simple Linear Regression to the Training set
          from sklearn.linear_model import LinearRegression
          regressor = LinearRegression()
          regressor.fit(X_train, y_train)
         LinearRegression()
In [12]: # Predicting the Test set results
          y_pred = regressor.predict(X_test)
In [13]:
          # Visualizing the Training set results
          viz_train = plt
          viz_train.scatter(X_train, y_train, color='red')
          viz_train.plot(X_train, regressor.predict(X_train), color='blue')
          viz_train.title('Salary VS Experience (Training set)')
          viz_train.xlabel('Year of Experience')
          viz_train.ylabel('Salary')
          viz_train.show()
```



In [14]: # Visualizing the Test set results viz_test = plt viz_test.scatter(X_test, y_test, color='red') viz_test.plot(X_train, regressor.predict(X_train), color='blue') viz_test.title('Salary VS Experience (Test set)') viz_test.xlabel('Year of Experience') viz_test.ylabel('Salary') viz_test.show()





PROGRAM 5:

Write a program to implement the naïve Bayesian classifier for a sample training data set stored as a .CSV file. Compute the accuracy of the classifier, considering few test data sets

Code:

import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
df = pd.read_csv("pima_indian.csv")

```
feature_col_names = ['num_preg', 'glucose_conc', 'diastolic_bp', 'thickness', 'insulin', 'bmi',
'diab_pred', 'age']
predicted_class_names = ['diabetes']
X = df[feature_col_names].values # these are factors for the prediction
y = df[predicted_class_names].values # this is what we want to predict
#splitting the dataset into train and test data
xtrain,xtest,ytrain,ytest=train_test_split(X,y,test_size=0.33)
print ('\n the total number of Training Data:',ytrain.shape)
print ('\n the total number of Test Data :',ytest.shape)
# Training Naive Bayes (NB) classifier on training data.clf = GaussianNB().fit(xtrain,ytrain.ravel())
predicted = clf.predict(xtest)
predictTestData= clf.predict([[6,148,72,35,0,33.6,0.627,50]])
#printing Confusion matrix, accuracy, Precision and Recall
print('\n Confusion matrix')
print(metrics.confusion matrix(ytest,predicted))
print('\n Accuracy of the classifier is',metrics.accuracy_score(ytest,predicted))
print('\n The value of Precision', metrics.precision_score(ytest,predicted))
print('\n The value of Recall', metrics.recall score(ytest,predicted))
print("Predicted Value for individual Test Data:", predictTestData)
```

```
In [18]: import pandas as pd
        from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn import metrics
In [24]:
       df = pd.read_csv("pima_indian.csv")
col_names = ['num_preg', 'glucose_conc', 'diastolic_bp', 'thickness', 'insulin', 'bmi', 'diab_pred', 'age']
predicted_class = ['diabetes']
        df
Out[24]: num_preg glucose_conc diastolic_bp thickness insulin bmi diab_pred age diabetes
        0
              6
                                  72
                                         35
       1 1 85 66 29 0 26.6 0.351 31 0
        2 8 183 64 0 0 23.3 0.672 32
       3 1 89 66 23 94 28.1 0.167 21 0
               0 137 40 35 168 43.1
                                                       2.288 33
       763
            10 101 76 48 180 32.9
                                                       0.171 63
       764 2 122 70 27 0 36.8 0.340 27 0
       765
              5
                        121
                             72 23 112 26.2
                                                       0.245 30
             1 126 60 0 0 30.1 0.349 47 1
       766
       767
            1 93
                             70 31 0 30.4
                                                       0.315 23
      768 rows × 9 columns
In [25]: X = df[col_names].values
        y = df[predicted_class].values
In [26]:
        print(df.head)
        xtrain,xtest,ytrain,ytest-train_test_split(X,y,test_size-0.4)
        print ('\n the total number of Training Data :',ytrain.shape)
```

```
In [25]: X = df[col_names].values
         y - df[predicted_class].values
In [26]: print(df.head)
         xtrain,xtest,ytrain,ytest-train_test_split(X,y,test_size-0.4)
         print ('\n the total number of Training Data :',ytrain.shape)
print ('\n the total number of Test Data :',ytest.shape)
                                         cbound method NDFrame.head of
                              148
85
                                                                 0 23.3
                               183
                                                   8
23
35
...
48
27
23
8
31
                                           66
48
         4
                                                             168 43.1
                   0
                             137
                                                               188 32.9
                   10
                                           70
72
60
         764
                               122
                                                                 0 36.8
                                                              112 26.2
0 30.1
         765
                               121
         766
                              126
                                                               0 30.4
         767
                   1
                               93
             diab_pred age diabetes
0.627 50 1
                 0.351 31
                0.672 32
                0.167 21
                 2.288 33
                                  1
               0.171 63
         763
               0.340 27
0.245 30
         764
         765
         766
                 0.349 47
         767
               0.315 23
         [768 rows x 9 columns]>
         the total number of Training Data : (460, 1)
         the total number of Test Data : (308, 1)
In [27]: clf = GaussianNB().fit(xtrain,ytrain.ravel())
         predicted = clf.predict(xtest)
         predictTestData= clf.predict([[6,148,72,35,8,33.6,8.627,58]])
In [28]:
         print('\n Confusion matrix')
         print/metrics.confusion matrix(vtest.predicted))
```

```
[768 rows x 9 columns]>
           the total number of Training Data : (460, 1)
           the total number of Test Data : (308, 1)
In [27]:
clf = GaussianNB().fit(xtrain,ytrain.ravel())
           predicted = clf.predict(xtest)
           predictTestData= clf.predict([[6,148,72,35,0,33.6,0.027,50]])
In [28]: print('\n Confusion matrix')
           print(metrics.confusion_matrix(ytest,predicted))
           print('\n Accuracy of the classifier is',metrics.accuracy_score(ytest,predicted))
           print('\n The value of Precision', metrics.precision_score(ytest,predicted))
           print('\n The value of Recall', metrics.recall_score(ytest,predicted))
           print("Predicted Value for individual Test Data:", predictTestData)
          Confusion matrix
          [[177 22]
[ 45 64]]
           Accuracy of the classifier is 0.7824675324675324
           The value of Precision 0.7441860465116279
           The value of Recall 0.5871559633027523
          Predicted Value for individual Test Data: [1]
```