VISVESVARAYA TECHNOLOGICAL UNIVERSITY

"JnanaSangama", Belgaum -590014, Karnataka.



LAB REPORT on

MACHINE LEARNING LAB

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
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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 **LIKITHA B (1BM19CS079)**, 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 **Course Title - (Course code)** work prescribed for the said degree.

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WEEK 1 FIND S ALGORITHM

```
import pandas as pd
import numpy as np
#to read the data in the csv file
print("USN:1BM19CS095")
data = pd.read csv(r"C:\Users\admin\Downloads\data.csv")
print(data,"\n")
#making an array of all the attributes
d = np.array(data)[:,:-1]
print("The attributes are: ",d)
#segragating the target that has positive and negative examples
target = np.array(data)[:,-1]
print("The target is: ",target)
#training function to implement find-s algorithm
def train(c,t):
    for i, val in enumerate(t):
        if val == "Yes":
            specific hypothesis = c[i].copy()
```

break

OUTPUT

```
USN: 1BM19CS095
     Time Weather Temperature Company Humidity
                                            Wind Goes
0 Morning Sunny
                      Warm Yes
                                      Mild Strong Yes
                       Cold
                                      Mild Normal
1 Evening Rainy
                               No
2 Morning Sunny Moderate Yes Normal Normal Yes
                             Yes
3 Evening Sunny
                       Cold
                                      High Strong Yes
The attributes are: [['Morning' 'Sunny' 'Warm' 'Yes' 'Mild' 'Strong']
['Evening' 'Rainy' 'Cold' 'No' 'Mild' 'Normal']
['Morning' 'Sunny' 'Moderate' 'Yes' 'Normal' 'Normal']
 ['Evening' 'Sunny' 'Cold' 'Yes' 'High' 'Strong']]
The target is: ['Yes' 'No' 'Yes' 'Yes']
n The final hypothesis is: ['?' 'Sunny' '?' 'Yes' '?' '?']
```

WEEK 2

CANDIDATE ELIMINATION ALGORITHM

```
Import
numpy
as np
    import pandas as pd

data = pd.read_csv(r'C:\Users\admin\Downloads\enjoysport.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("\nInitialization of specific_h and genearal_h")
    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":
            print("Instance is Positive ")
            for x in range(len(specific_h)):
                if h[x]!= specific_h[x]:
                    specific_h[x] ='?'
                    general_h[x][x] = '?'
        if target[i] == "no":
            print("Instance is Negative ")
            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 Bundary after ", i+1, "Instance is ", specific_h)
        print("Generic Boundary after ", i+1, "Instance is ", 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("Final Specific_h: ", s_final, sep="\n")
print("Final General_h: ", g_final, sep="\n")
```

OUTPUT

```
Instances are:
       Instances are:

[['sunny' 'warm' 'high' 'strong' 'warm' 'same']

['rainy' 'cold' 'high' 'strong' 'warm' 'change']

['rainy' 'cold' 'high' 'strong' 'warm' 'change']

['sunny' 'warm' 'high' 'strong' 'cool' 'change']
   Target Values are: ['yes' 'yes' 'no' 'yes']
Initialization of specific_h and genearal_h
Specific Boundary: ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
   Generic Goundary: [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?']]
   Instance I is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
    Specific Bundary after 1 Instance is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
Generic Boundary after 1 Instance is [['?', '2', '2', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?']]
     Instance 2 Is ['sunny' 'warm' 'high' 'strong' 'warm' 'same']
      Instance is Positive
    Specific Bundary after 2 Instance is ['sunny' 'ware' '?' 'strong' 'ware' 'same']

Generic Boundary after 2 Instance is [['?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?']]
     Instance 3 is ['rainy' 'cold' 'high' 'strong' 'ware' 'change']
    Instance is Negative
Specific Bundary after 3 Instance is ['sunny' 'warm' '?' 'strong' 'warm' 'same']
Generic Boundary after 3 Instance is [['sunny', '2', '2', '2', '2', '2'], ['2', 'warm', '2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'],
     Instance 4 is ['sunny' 'warm' 'high' 'strong' 'cool' 'change']
     Instance is Positive
     Specific Bundary after 4 Instance is ['sunny' 'warm' '?' 'strong' '?' '?']

Generic Boundary after 4 Instance is [['sunny', '2', '2', '2', '2', '2'], ['2', 'warm', '2', '2', '2', '2', '2'], ['2', '3', '2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2', '2'], ['2', '2'], ['2', '2', '2'], ['2', '2'], ['2', '2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2', '2'], ['2'
   Final Specific_h:
!'sumnv' 'warm' '?' 'strong' '?' '?']
     ['sunny "', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]
```

WEEK 3 DECISION TREE USING ID3 ALGORITHM

```
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(r"C:\Users\admin\Downloads\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(r"C:\Users\admin\Downloads\id3_test.csv")
for xtest in testdata:
    print("The test instance:",xtest)
    print("The label for test instance:")
```

OUTPUT

```
The decision tree for the dataset using ID3 algorithm is
Outlook
  overcast
    yes
  rain
    Wind
       strong
        no
      weak
        yes
  sunny
    Humidity
       high
         no
       normal
        yes
The test instance: ['rain', 'cool', 'normal', 'strong']
The label for test instance:
The test instance: ['sunny', 'mild', 'normal', 'strong']
The label for test instance:
yes
```

WEEK 4

LINEAR REGRESSION

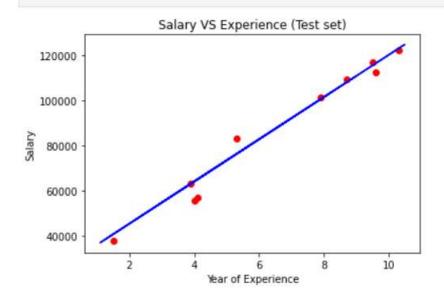
```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

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)
```

```
# Fitting Simple Linear Regression to the Training set
from sklearn.linear model import LinearRegression
regressor = LinearRegression()
regressor.fit(X train, y train)
# Predicting the Test set results
y pred = regressor.predict(X test)
# 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()
# 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()
```

OUTPUT





LAB 5 NAÏVE BAYES NETWORK

```
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(r"C:\Users\admin\Downloads\data5.csv")
col names = ['num preg', 'glucose conc', 'diastolic bp', 'thickness',
'insulin', 'bmi', 'diab pred', 'age']
predicted class = ['diabetes']
X = df[col names].values
y = df[predicted_class].values
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)
clf = GaussianNB().fit(xtrain,ytrain.ravel())
predicted = clf.predict(xtest)
predictTestData= clf.predict([[6,148,72,35,0,33.6,0.627,50]])
```

```
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)
                           num_preg glucose_conc diastolic_bp thickness insulin bmi \
 148
                                             0 33.6
         6
                             72
                                  35
 1
         1
                   85
                             66
                                     29
                                            0 26.6
                                            0 23.3
 2
         8
                  183
                             64
                                     .0
                                    23
 3
         1
                  89
                            66
                                           94 28.1
        0
 4
                  137
                            40
                                    35 168 43.1
                                          180 32.9
 763
         10
                  101
                             76
                                     48
                            70
                                    27
 764
         2
                  122
                                            0 36.8
                                    23
                                          112 26.2
 765
                             72
         5
                  121
                                            0 30.1
 766
                            60
                                      0
                  126
         1
 767
                  93
                            79
                                   31
                                           0 30.4
    diab_pred age diabetes
 0
       0.627
             50
       0.351 31
 1
       0.672 32
 2
                     1
       0.167 21
 4
       2.288 33
                    1
 763
       0.171
             63
 764
       0.340
             27
                     9
       0.245 30
 765
                     9
 766
       0.349 47
                     1
       0.315 23
 [768 rows x 9 columns]>
 the total number of Training Data : (460, 1)
  the total number of Test Data : (308, 1)
   Confusion matrix
  [[176 29]
   [ 40 63]]
   Accuracy of the classifier is 0.775974025974026
   The value of Precision 0.6847826086956522
   The value of Recall 0.6116504854368932
  Predicted Value for individual Test Data: [1]
```