VISVESVARAYA TECHNOLOGICAL UNIVERSITY

"JnanaSangama", Belgaum -590014, Karnataka.



LAB REPORT on

MACHINE LEARNING

Submitted by

Sneha Srivastava (1BM19CS158)

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 Lab" carried out by **SNEHA SRIVASTAVA** (1BM19CS158), who is a 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 with respect to **Machine Learning - (20CS6PCMAL)** work prescribed for the said degree.

Saritha A.N. Assistant Professor Department of CSE BMSCE, Bengaluru **Dr. Jyothi S Nayak**Professor and Head
Department of CSE
BMSCE, Bengaluru

Index Sheet

Sl. No.	Experiment Title	Page No.
1)	Find-S	4-5
2)	Candidate Elimination	6-7
3)	Decision tree based on ID3	8-10
4)	Naive Bayesian Classifier	11-12
5)	Linear Regression	13-14

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

```
import csv
a = []
with open('/kaggle/input/dataset/data.csv','r') as csvfile:
   for row in csv.reader(csvfile):
      a.append(row)
      print(a)
   print("\n The total number of training instances are : ",len(a))
num attribute = len(a[0])-1
print("\n The initial hypothesis is : ")
hypothesis = ['0']*num attribute
print(hypothesis)
for i in range(0, len(a)):
   if a[i][num attribute] == 'yes':
      for j in range(0, num_attribute):
          if hypothesis[j] == '0' or hypothesis[j] == a[i][j]:
             hypothesis[j] = a[i][j]
          else:
             hypothesis[j] = '?'
   print("\n The hypothesis for the training instance {} is :\n" .format(i+1),hypothesis)
print("\n The Maximally specific hypothesis for the training instances is :")
print(hypothesis)
```

```
The total number of training instances are : 5

The initial hypothesis is :
['0', '0', '0', '0', '0', '0']

The hypothesis for the training instance 1 is :
['0', '0', '0', '0', '0', '0']

The hypothesis for the training instance 2 is :
['sunny', 'warm', 'normal', 'strong', 'warm', 'same']

The hypothesis for the training instance 3 is :
['sunny', 'warm', '?', 'strong', 'warm', 'same']

The hypothesis for the training instance 4 is :
['sunny', 'warm', '?', 'strong', 'warm', 'same']

The hypothesis for the training instance 5 is :
['sunny', 'warm', '?', 'strong', '?', '?']

The Maximally specific hypothesis for the training instances is :
['sunny', 'warm', '?', 'strong', '?', '?']
```

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.

```
import numpy as np
import pandas as pd
data = pd.read csv('/kaggle/input/dataset/data.csv')
concepts = np.array(data.iloc[:,0:-1])
print(concepts)
target = np.array(data.iloc[:,-1])
print(target)
def learn(concepts, target):
  specific h = concepts[0].copy()
  print("Initialization of specific h and general h")
  print(specific h)
  general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
  print(general h)
  for i, h in enumerate(concepts):
     print("For Loop Starts")
     if target[i] == "yes":
        print("If 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("If instance is Negative ")
        for x in range(len(specific h)):
           if h[x]!= specific h[x]:
             general h[x][x] = \text{specific } h[x]
           else:
             general h[x][x] = '?'
     print("Steps of Candidate Elimination Algorithm",i+1)
     print(specific h)
     print(general h)
     print("\n")
     print("\n")
  indices = [i \text{ for } i, \text{ val in enumerate}(\text{general } h) \text{ if } \text{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")
```

```
[['sumy' 'sum' 'normal' 'strong' 'warm' 'same']
['rainy' 'cola' 'high' 'strong' 'warm' 'same']
['rainy' 'cola' 'high' 'strong' 'warm' 'same']
['sumy' 'warm' 'high' 'strong' 'warm' 'same']
['sumy' 'warm' 'normal' 'strong' 'warm' 'same']

For Long Starts
I' instence is Registion

For Long Starts

For Long Starts
```

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.

```
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)
"Main program"
dataset,features=load csv("/kaggle/input/train/ids_train.csv")
node1=build tree(dataset,features)
print("The decision tree for the dataset using ID3 algorithm is :\n")
print tree(node1,0)
```

Yes └─ High No └─ Overcast

Yes

```
The decision tree for the dataset using ID3 algorithm is:

Outlook

Rain

Wind

Weak

Yes

Strong

No

Sunny

Humidity

Normal
```

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

```
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("/kaggle/input/diabetes/diabetes.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
y = df[predicted class names].values
print(df.head)
xtrain,xtest,ytrain,ytest=train test split(X,y,test size=0.40)
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)
```

```
[145 rows x 9 columns]>

The total number of Training Data : (87, 1)

The total number of Test Data : (58, 1)

Confusion matrix

[[31 7]

[10 10]]

Accuracy of the classifier is 0.7068965517241379

The value of Precision 0.5882352941176471

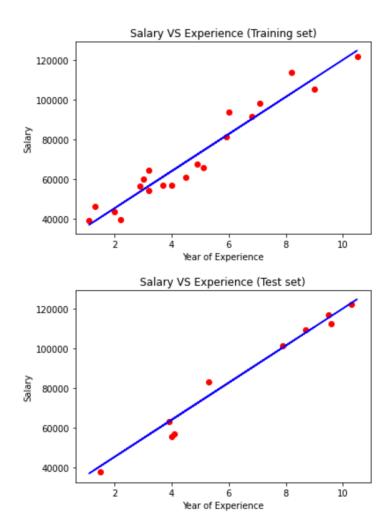
The value of Recall 0.5

Predicted Value for individual Test Data: [1]
```

5. 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
# Importing the dataset
dataset = pd.read csv('/kaggle/input/years-of-experience-and-salary/Years Experience and Salary.csv')
X = dataset.iloc[:, :-1].values #get a copy of dataset exclude last column
y = dataset.iloc[:, 1].values #get array of dataset in column 1st
# Splitting the dataset into the Training set and Test set
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()
```



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