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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)
BENGALURU-560019

May-2022 to July-2022

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CERTIFICATE

This is to certify that the Lab work entitled "Machine Learning" carried out by MITHUN R K (1BM19CS087), 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.

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Course Outcome

CO1	Ability to apply the different learning algorithms.
CO2	Ability to analyze the learning techniques for given dataset.
CO3	Ability to design a model using machine learning to solve a problem.
CO4	Ability to conduct practical experiments to solve problems using appropriate machine learning techniques

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

```
import numpy as np
import pandas as pd
data = pd.read_csv("mydata.csv")
print(data,"\n")
d = np.array(data)[:,:-1]
print("\n The attributes are: ",d)
target = np.array(data)[:,-1]
print("\n The target is: ",target)
def findS(c,t):
  for i, val in enumerate(t):
    if val == "Yes":
      specific_hypothesis = c[i].copy()
  for i, val in enumerate(c):
    if t[i] == "Yes":
      for x in range(len(specific_hypothesis)):
        if val[x] != specific hypothesis[x]:
           specific_hypothesis[x] = '?'
         else:
           pass
  return specific_hypothesis
print("\n The final hypothesis is:",findS(d,target))
```

Dataset:

```
Time Weather Temperature Company Humidity
                                            Wind Goes
0 Morning Sunny
                   Warm
                            Yes
                                     Mild
                                            Strong Yes
1 Evening Rainy
                                     Mild
                                            Normal No
                  Cold
                        No No
2 Morning Sunny
                 Moderate
                           Yes
                                    Normal
                                            Normal Yes
3 Evening Sunny
                  Cold
                            Yes
                                    High
                                             Strong Yes
```

Output:

The final hypothesis is: ['?' 'Sunny' '?' 'Yes' '?' '?']

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.DataFrame(data=pd.read_csv('data.csv'))
print(data)
concepts=np.array(data.iloc[:,0:-1])
print("The attributes are : ",concepts)
target=np.array(data.iloc[:,-1])
print ("\n The target is =",target)
def learn(concepts, target):
  specific_h=concepts[0].copy()
  print("\n Initialization of specfic_h and generalization")
  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]=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 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")
```

Dataset:

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

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):
  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:",end=" ")
classify(node1,xtest,features)
```

Dataset:

Outlook	Temperature	Humidity	Wind
rain	cool	normal	strong
sunny	mild	normal	strong

```
The decision tree for the dataset using ID3 algorithm is
Outlook
   rain
     Wind
       weak
         yes
       strong
         no
   overcast
    yes
   sunny
    Humidity
       high
         no
       normal
        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
```

4. 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

```
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("diabetes.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:', vtrain.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:", predictTestDat
```

Dataset:

	num_preg	glucose_conc	diastolic_bp	thickness	insulin	bmi	diab_pred	age	diabetes
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1
140	3	128	78	0	0	21.1	0.268	55	0
141	5	106	82	30	0	39.5	0.286	38	0
142	2	108	52	26	63	32.5	0.318	22	0
143	10	108	66	0	0	32.4	0.272	42	1
144	4	154	62	31	284	32.8	0.237	23	0

Confusion matrix
[[32 10]
[9 7]]

Accuracy of the classifier is 0.6724137931034483

The value of Precision 0.4117647058823529

The value of Recall 0.4375
Predicted Value for individual Test Data: [1]

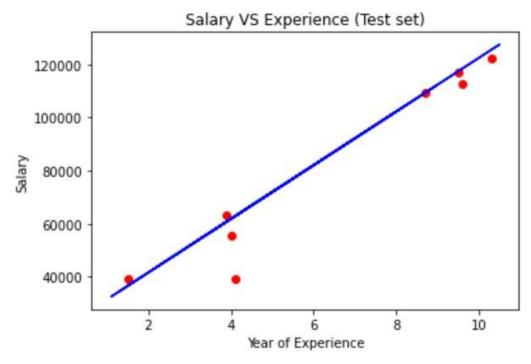
5. Implement the Linear Regression algorithm in order to fit data points. Select appropriate data set for your experiment and draw graphs

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
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=
# 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()
regressor.score(X_train,y_train)
print(regressor.score(X test,y test))
```

Dataset:

	Dataset.						
1	YearsExperience	Salary					
2	1.1	39343					
3	1.3	46205					
4	1.5	37731					
5	2.0	43525					
6	2.2	39891					
7	2.9	56642					
8	3.0	60150					
9	3.2	54445					
10	3.2	64445					
11	3.7	57189					
12	3.9	63218					
13	4.0	55794					
14	4.0	56957					
15	4.1	57081					
16	4.5	61111					
17	4.9	67938					
18	5.1	66029					
19	5.3	83088					
20	5.9	81363					
21	6.0	93940					
22	6.8	91738					
23	7.1	98273					





0.9251138619118122