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LAB REPORT on

MACHINE LEARNING

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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CERTIFICATE

This is to certify that the Lab work entitled "Machine Learning Lab" carried out by **S R Pooja** (1BM19CS135), 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.

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

CODE:

```
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.

CODE:

```
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] = 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")

```
[['suny,' 'ware' 'normal' 'strong' 'ware' 'saee']
['suny,' 'ware' 'high' 'strong' 'ware' 'saee']
['suny,' 'ware' 'high' 'strong' 'col' 'change']]
['yes' 'yes' 'no' 'yes']
['yes' '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.

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):
  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]
          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)
  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)
```

OUTPUT:

The decision tree for the dataset using ID3 algorithm is :

```
Outlook

Rain

Wind

Weak

Yes

Strong

No

Sunny

Humidity

Normal

Yes

High

No

Overcast

Yes
```

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.

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("/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.

```
CODE:
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()
```

