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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 **Shweta Patil(1BM19CS156)**, 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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Experiment Title	Page No.
Find-S	4-5
Candidate Elimination	6-7
Decision tree based on ID3	8-10
Naive Bayesian Classifier	11-12
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)
```

4

OUTPUT:

```
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']

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', '?', '?']
```

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] = 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(['?', '?', '?', '?', '?', '?'])
                                                                                                                    6
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")
```

```
| Total content | Total conten
```

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.ans wer=""
```

```
defsubtables(data,col,delete):
dic={}
coldata=[row[col] for row indata]
attr=list(set(coldata))
counts=[0]*len(attr)
r=len(data)
c=len(data[0])
forxinrange(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)]forjinrange(counts[x])]
pos=0
foryinrange(r):
if data[y][col]==attr[x]:
if delete:
del data[y][col]
dic[attr[x]][pos]=data[y]
pos+=1
returnattr,dic
defentropy(S):
attr=list(set(S))
iflen(attr)==1:
return0
counts=[0,0]
foriinrange(2):
counts[i]=sum([1forxinSifattr[i]==x])/(len(S)*1.0)
sums=0
forentineounts:
sums+=-1*cnt*math.log(cnt,2)
returnsums
defcompute_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]forrowindata])
forxinrange(len(attr)):
ratio[x]=len(dic[attr[x]])/(total_size*1.0)
entropies[x]=entropy([row[-1]forrowindic[attr[x]]])
total_entropy-=ratio[x]*entropies[x]
returntotal_entropy
defbuild_tree(data,features):
lastcol=[row[-1] for row in data]
```

if(len(set(lastcol)))==1:

8

```
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)
forx 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:
                                                                                                             9
   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
```

10

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')
                                                                            <u>11</u>
  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)
  OUTPUT:
     [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]
```

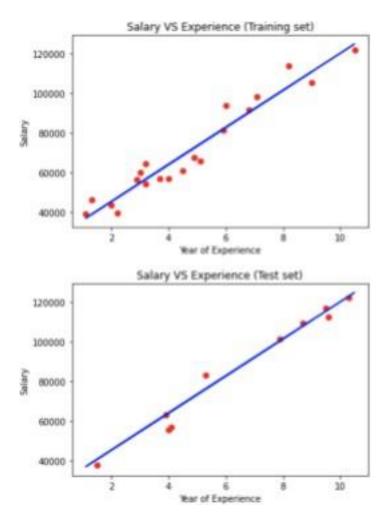
12

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()
```



OUTPUT: 15