

# VISVESVARAYA TECHNOLOGICAL UNIVERSITY

“JnanaSangama”, Belgaum -590014, Karnataka.



## 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**



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**Department of Computer Science and Engineering**



**CERTIFICATE**

This is to certify that the Lab work entitled “Machine Learning Lab” carried out by **Roshni Dash (1BM19CS133)**, 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)
```

## 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', '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 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 :

```

[ ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
  ['sunny' 'warm' 'high' 'strong' 'warm' 'same']
  ['rainy' 'cold' 'high' 'strong' 'warm' 'change']
  ['sunny' 'warm' 'high' 'strong' 'cool' 'change']]
['yes' 'yes' 'no' 'yes']
Initialization of specific_h and general_h
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]]
For Loop Starts
If instance is Positive
Steps of Candidate Elimination Algorithm 1
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]]

For Loop Starts
If instance is Positive
Steps of Candidate Elimination Algorithm 2
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]]

For Loop Starts
If instance is Negative
Steps of Candidate Elimination Algorithm 3
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', 'same']]]

For Loop Starts
If instance is Positive
Steps of Candidate Elimination Algorithm 4
['sunny' 'warm' '?' 'strong' '?' '?']
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]]

Final Specific_h:
['sunny' 'warm' '?' 'strong' '?' '?']
Final_g:

```

```

For Loop Starts
If instance is Positive
Steps of Candidate Elimination Algorithm 2
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

For Loop Starts
If instance is Negative
Steps of Candidate Elimination Algorithm 3
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', 'same']]

For Loop Starts
If instance is Positive
Steps of Candidate Elimination Algorithm 4
['sunny' 'warm' '?' 'strong' '?' '?']
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

Final Specific:h:
['sunny' 'warm' '?' 'strong' '?' '?']
Final General:h:
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]

```

+ Code

+ Markdown

**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):
    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)
```

**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)
```

### **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]
```

**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()

```

## OUTPUT:



