

# 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**



**B.M.S. COLLEGE OF ENGINEERING**

(Autonomous Institution under VTU)

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**B. M. S. College of Engineering,**  
**Bull Temple Road, Bangalore 560019**  
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**Department of Computer Science and Engineering**



**CERTIFICATE**

This is to certify that the Lab work entitled “**Machine Learning**” carried out by **Mohammed Ibrahim Rahil S (1BM19CS090)**, 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 in respect of a **Machine Learning - (20CS6PCMAL)** work prescribed for the said degree.

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### Course Outcome

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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.

```
In [64]: import pandas as pd
import numpy as np
```

```
In [65]: data = pd.read_csv("data.csv")
```

```
In [66]: data
```

```
Out[66]:
```

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

```
In [67]: last_column = data.iloc[:, -1]
required_index = 0
```

```
In [68]: for i, n in enumerate(last_column):
if n == "Yes":
required_index = i;
break
```

```
In [69]: row_length = len(data.index)
col_length = len(data.columns)
f_p_hypothesis = data.iloc[required_index].values.tolist()[:-1]
for i in range(required_index+1, row_length):
if data.iloc[i, col_length-1] == "Yes":
for j in range(0, col_length-1):
if data.iloc[i, j] != "?" and data.iloc[i, j] != f_p_hypothesis[j]:
f_p_hypothesis[j] = "?"
else:
pass
```

```
In [70]: f_p_hypothesis
```

```
Out[70]: ['Sunny', '?', '?', '?']
```

```
In [16]: n = int(input("Enter no of samples "))
data = list()
outcomes = list()
for i in range(0, n):
lis = input("Enter features ").split(" ")
outc = input("Enter outcome ")
data.append(lis)
outcomes.append(outc)
row_length = len(data)
col_length = len(data[0])
for i in range(0, row_length):
if outcomes[i] == "Yes":
hypothesis = data[i]
index = i
break
for i in range(index+1, row_length):
if outcomes[i] == "Yes":
for j in range(0, col_length):
if data[i][j] != '?' and data[i][j] != hypothesis[j]:
hypothesis[j] = "?"

Enter no of samples 4
Enter features Sunny Warm Mild Strong
Enter outcome Yes
Enter features Rainy Cold Mild Normal
Enter outcome No
Enter features Sunny Moderate Normal Normal
Enter outcome Yes
Enter features Sunny Cold High Strong
Enter outcome Yes
```

```
In [17]: hypothesis
```

```
Out[17]: ['Sunny', '?', '?', '?']
```

```
In [15]:
```

```
Out[15]: ['Sunny', 'Warm', 'Mild', '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.**

```
In [9]: import numpy as np
import pandas as pd

data = pd.read_csv('data.csv')
concepts = np.array(data.iloc[:, 0:-1])
print("Concepts : ")
print(concepts)
target = np.array(data.iloc[:, -1])
print("Target : ", target)

def learn(concepts, target):
    specific_h = concepts[0].copy()
    print("Specific Hypothesis : ", specific_h)
    general_h = [["?" for i in range(len(specific_h))]
                  for i in range(len(specific_h))]
    print("General Hypothesis : ", general_h, "\n")
    for i, h in enumerate(concepts):
        if target[i] == "yes":
            print(i+1, "\n 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(i+1, "\n 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("Specific Hypothesis : ", specific_h)
            print("General Hypothesis : ", general_h, end="\n\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")
```

```

Concepts :
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]
Target : ['yes' 'yes' 'no' 'yes']
Specific Hypothesis : ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
General Hypothesis : [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?']]

1 )
  If instance is Positive
Specific Hypothesis : ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
General Hypothesis : [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?']]

2 )
  If instance is Positive
Specific Hypothesis : ['sunny' 'warm' '?' 'strong' 'warm' 'same']
General Hypothesis : [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?']]

3 )
  If instance is Negative
Specific Hypothesis : ['sunny' 'warm' '?' 'strong' 'warm' 'same']
General Hypothesis : [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?'], ['?', '?', '?', '?', '?', 'same']]

4 )
  If instance is Positive
Specific Hypothesis : ['sunny' 'warm' '?' 'strong' '?' '?']
General Hypothesis : [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'],
['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],
['?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

Final Specific_h:
['sunny' 'warm' '?' 'strong' '?' '?']
Final General_h:
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]

```

In [ ]:

**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.**

```
In [7]: 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
```

```
In [8]: 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)
```

```
In [9]: 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)
```



```
In [7]: 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
```

```
In [8]: 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)
```

```
In [9]: 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)
```

```
In [10]: '''Main Program'''
        dataset, features = load_csv("data3.csv")
        model = build_tree(dataset, features)

        print("The decision tree for the dataset using ID3 algorithm is")
        print_tree(model, 0)
        testdata, features = load_csv("data3_test.csv")
        for xtest in testdata:
            print("The test instance: ", xtest)
            print("The label for test instance: ", end = " ")
            classify(model, xtest, features)
```

The decision tree for the dataset using ID3 algorithm is

```
Outlook
  overcast
    yes
  sunny
    Humidity
      normal
        yes
      high
        no
  rain
    Wind
      weak
        yes
      strong
        no
```

```
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.

```
In [1]: import pandas as pd
from sklearn import tree
from sklearn.preprocessing import LabelEncoder
from sklearn.naive_bayes import GaussianNB

data = pd.read_csv('tennisdata.csv')
print("The first 5 values of data is :\n",data.head())
```

```
The first 5 values of data is :
  Outlook Temperature Humidity Windy PlayTennis
0   Sunny           Hot     High   False        No
1   Sunny           Hot     High    True        No
2  Overcast          Hot     High   False        Yes
3   Rainy           Mild     High   False        Yes
4   Rainy           Cool    Normal   False        Yes
```

```
In [2]: X = data.iloc[:, :-1]
print("\nThe First 5 values of train data is\n",X.head())
```

```
The First 5 values of train data is
  Outlook Temperature Humidity Windy
0   Sunny           Hot     High   False
1   Sunny           Hot     High    True
2  Overcast          Hot     High   False
3   Rainy           Mild     High   False
4   Rainy           Cool    Normal   False
```

```
In [3]: y = data.iloc[:, -1]
print("\nThe first 5 values of Train output is\n",y.head())
```

```
The first 5 values of Train output is
0    No
1    No
2    Yes
3    Yes
4    Yes
Name: PlayTennis, dtype: object
```

```
In [4]: le_outlook = LabelEncoder()
X.Outlook = le_outlook.fit_transform(X.Outlook)

le_Temperature = LabelEncoder()
X.Temperature = le_Temperature.fit_transform(X.Temperature)

le_Humidity = LabelEncoder()
X.Humidity = le_Humidity.fit_transform(X.Humidity)

le_Windy = LabelEncoder()
X.Windy = le_Windy.fit_transform(X.Windy)

print("\nNow the Train data is :\n",X.head())
```

```
Now the Train data is :
  Outlook Temperature Humidity Windy
0        2           1         0      0
1        2           1         0      1
2        0           1         0      0
3        1           2         0      0
4        1           0         1      0
```

```
In [5]: le_PlayTennis = LabelEncoder()
y = le_PlayTennis.fit_transform(y)
print("\nNow the Train output is\n",y)
```

```
Now the Train output is
[0 0 1 1 1 0 1 0 1 1 1 1 0]
```

```
In [6]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.20)

classifier = GaussianNB()
classifier.fit(X_train,y_train)

from sklearn.metrics import accuracy_score
print("Accuracy is:",accuracy_score(classifier.predict(X_test),y_test))
```

```
Accuracy is: 0.6666666666666666
```

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

## Linear Regression

```
In [2]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [3]: dataset = pd.read_csv('salaryData.csv')
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 1].values
```

## Splitting the dataset

```
In [4]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=1/3, random_state=0)
regressor = LinearRegression()
regressor.fit(X_train, y_train)
```

```
Out[4]: LinearRegression()
```

## Training

```
In [5]: y_pred = regressor.predict(X_test)
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()
```



## Testing

```
In [6]: 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()
```

