

VISVESVARAYA TECHNOLOGICAL UNIVERSITY

“JnanaSangama”, Belgaum -590014, Karnataka.



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

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CERTIFICATE

This is to certify that the Lab work entitled “**MACHINE LEARNING**” carried out by **HARSHIL DOSHI B (1BM19CS057)**, 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

CO	Ability to apply the different learning algorithms.
1	Ability to analyse the learning techniques for given dataset.
CO	Ability to design a model using machine learning to solve a problem.
2	Ability to conduct practical experiments to solve problems using appropriate machine learning techniques.
CO	
3	
CO4	

1. Find S Algorithm

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

```
In [1]: import numpy as np
import pandas as pd
```

```
In [2]: from google.colab import drive
drive.mount("/content/drive")
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
In [3]: data = pd.read_csv("/content/drive/MyDrive/finddata.csv")
print(data, "\n")
```

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

```
In [4]: d = np.array(data)[:,-1]
print("\n The attributes are: ", d)
target = np.array(data)[:,-1]
print("\n The target is: ", target)
```

The attributes are: [['Morning' 'Sunny' 'Warm' 'Yes' 'Mild' 'Strong']
['Evening' 'Rainy' 'Cold' 'No' 'Mild' 'Normal']
['Morning' 'Sunny' 'Moderate' 'Yes' 'Normal' 'Normal']
['Evening' 'Sunny' 'Cold' 'Yes' 'High' 'Strong']]

The target is: ['Yes' 'No' 'Yes' 'Yes']

```
In [5]: def findS(c,t):
    for i, val in enumerate(t):
        if val == "Yes":
            specific_hypothesis = c[i].copy()
            break
```

The attributes are: [['Morning' 'Sunny' 'Warm' 'Yes' 'Mild' 'Strong']
['Evening' 'Rainy' 'Cold' 'No' 'Mild' 'Normal']
['Morning' 'Sunny' 'Moderate' 'Yes' 'Normal' 'Normal']
['Evening' 'Sunny' 'Cold' 'Yes' 'High' 'Strong']]

The target is: ['Yes' 'No' 'Yes' 'Yes']

```
In [5]: def findS(c,t):
    for i, val in enumerate(t):
        if val == "Yes":
            specific_hypothesis = c[i].copy()
            break

    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
```

```
In [6]: print("\n The final hypothesis is:", findS(d, target))
```

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

2. Candidate Elimination Algorithm

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 [14]: import numpy as np
import pandas as pd

#to read the data in the csv file
data = pd.DataFrame(data=pd.read_csv('enjoysport.csv'))
print(data,"\n")

#making an array of all the attributes
concepts = np.array(data.iloc[:,0:-1])
print("The attributes are: ",concepts)

#segregating the target that has positive and negative examples
target = np.array(data.iloc[:, -1])
print("\n The target is: ",target)

#training function to implement candidate_elimination algorithm
def learn(concepts, target):
    specific_h = concepts[0].copy()
    print("Initialization of specific hypothesis and general hypothesis:")
    print("S0: ",specific_h)
    general_h = [["?" for i in range(len(specific_h))] for i in range(len(specific_h))]
    print("G0: ",general_h)
    for i, h in enumerate(concepts):
        if target[i] == "yes":
            for x in range(len(specific_h)):
                if h[x] != specific_h[x]:
                    specific_h[x] = "?"
                    general_h[x][x] = "?"
        if target[i] == "no":
            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("The steps of Candidate Elimination Algorithm ", i+1)
    print("F"S{i+1}: " ,specific_h)
    actual_general_h = [i for i in general_h if i != ["?" for i in range(len(specific_h))]]
    print("F"G{i+1}: " ,actual_general_h)
    print(general_h)
    actual_general_h = [i for i in general_h if i != ["?" for i in range(len(specific_h))]]

1 sunny warm high strong warm same yes
2 rainy cold high strong warm change no
3 sunny warm high strong cool change yes

The attributes are: [['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
['sunny' 'warm' 'high' 'strong' 'warm' 'same']
['rainy' 'cold' 'high' 'strong' 'warm' 'change']
['sunny' 'warm' 'high' 'strong' 'cool' 'change']]

The target is: ['yes' 'yes' 'no' 'yes']

In [15]: s_final, g_final = learn(concepts, target)
print("The final Specific hypothesis is: ", s_final, sep="\n")
print("The final General Hypothesis is: ", g_final, sep="\n")

Initialization of specific hypothesis and general hypothesis:
S0: ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
G0: [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
The steps of Candidate Elimination Algorithm 1
S1: ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
G1: []
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
The steps of Candidate Elimination Algorithm 2
S2: ['sunny' 'warm' '?' 'strong' 'warm' 'same']
G2: []
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
The steps of Candidate Elimination Algorithm 3
S3: ['sunny' 'warm' '?' 'strong' 'warm' 'same']
G3: [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', 'same']]
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
The steps of Candidate Elimination Algorithm 4
S4: ['sunny' 'warm' '?' 'strong' '?' '?']
G4: [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
The final Specific hypothesis is:
['sunny' 'warm' '?' 'strong' '?' '?']
The final General Hypothesis is:
[['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]
```

3. ID3 Algorithm

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 [32]: import pandas as pd
import numpy as np
from sklearn.datasets import load_iris
data = load_iris()

In [33]: df = pd.DataFrame(data.data, columns = data.feature_names)

In [34]: df
Out[34]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2
...
145	6.7	3.0	5.2	2.3
146	6.3	2.5	5.0	1.9
147	6.5	3.0	5.2	2.0
148	6.2	3.4	5.4	2.3
149	5.9	3.0	5.1	1.8

150 rows x 4 columns

```
In [20]: df['Species'] = data.target
#replace this with the actual names

target = np.unique(data.target)
print(target)

target_names = np.unique(data.target_names)
print(target_names)
targets = dict(zip(target, target_names))
print(targets)
df['Species'] = df['Species'].replace(targets)

[0 1 2]
['setosa' 'versicolor' 'virginica']
{0: 'setosa', 1: 'versicolor', 2: 'virginica'}

In [21]: x = df.drop(columns="Species")
y = df["Species"]

In [22]: feature_names = x.columns
labels = y.unique()

In [36]: from sklearn.model_selection import train_test_split
x_train, test_x, y_train, test_lab = train_test_split(x,y,test_size = 0.4,random_state = 42)
# print(x_train)
# print(test_x)
# print(y_train)
# print(test_lab)
```

```
In [24]: from sklearn.tree import DecisionTreeClassifier
clf = DecisionTreeClassifier(max_depth=3, random_state=42, criterion='entropy')
```

```
Out[25]: DecTision+reeciassi-fier ( c rTterion= 'entropy' , cax_depth=3 random_state=42)
```

```
In [35]: test_pred = clf.predict(test_x)
         test_pred
```

```
Out[35]: array([ 'versicolor', 'setosa', 'virginica', 'versicolor', 'versicolor',
        'setosa', 'versicolor', 'virginica', 'versicolor', 'versicolor',
        'virginica', 'setosa', 'virginica', 'virginica', 'virginica',
        'virginica', 'virginica', 'setosa', 'setosa', 'setosa', 'setosa',
        'versicolor', 'setosa', 'setosa', 'virginica', 'versicolor',
        'setosa', 'setosa', 'setosa', 'virginica', 'versicolor',
        'versicolor', 'setosa', 'setosa', 'versicolor', 'versicolor',
        'setosa', 'setosa', * versicolor'], dtype=object)
```

```
ii-t wtpiotlib, pypJot as pJt
```

```
confusion_matrix = metrics.confusion_matrix(y_test, y_pred)
```

In [28]: confusion matrix

```
out[zsj]: array([ 23,  6,  0],
```

```
In [28a]: conf-usion_matrix
```

```
Out 2B§: array([ ]23, B, 0,§
```

```
[ 0, 1, 17]], dtype=int64)
```

```
In [29]: matrix_df = pd.DataFrame(confusionmatrix)
```

```
plt.figure(figsize=(10,7))
```

```
ax.set_title('Confusion Matrix - Decision Tree')
ax.set_xlabel("Predicted label", fontsize=15)
ax.set_xticklabels([''] + labels)
ax.set_ylabel("True Label", fontsize=15)
ax.set_yticklabels(list(labels), rotation = B)
```

Confusion Mabix - decision Tree

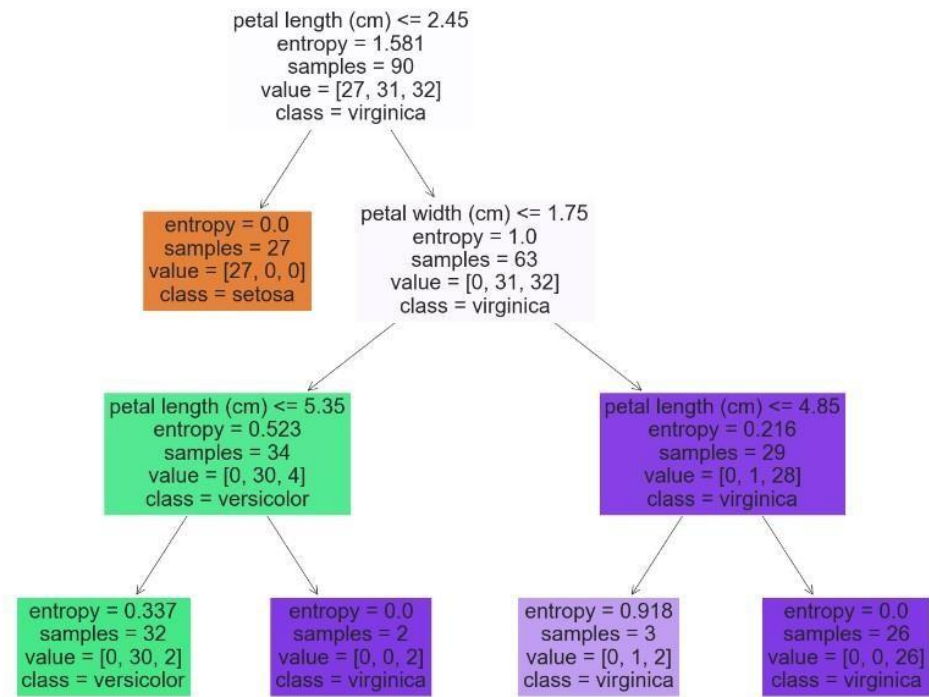
Actual label	Predicted label
0	0
1	1
2	2
3	3
4	4
5	5
6	6
7	7
8	8
9	9

<Figure size 720x504 with 0 Axes>

```
fig = plt.figure(figsize=(25,20))
```

```
feature_names=data.feature_names,
class_names=data.target_names,
```

```
In [31]: from sklearn import tree
fig = plt.figure(figsize=(25,20))
_ = tree.plot_tree(clf,
                  feature_names=data.feature_names,
                  class_names=data.target_names,
                  filled=True)
```



In []:

4. Naïve Bayes

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 [10]: import pandas as pd
data = pd.read_csv('PlayTennis.csv')
data
```

```
Out[10]:
```

	PlayTennis	Outlook	Temperature	Humidity	Wind
0	No	Sunny	Hot	High	Weak
1	No	Sunny	Hot	High	Strong
2	Yes	Overcast	Hot	High	Weak
3	Yes	Rain	Mild	High	Weak
4	Yes	Rain	Cool	Normal	Weak
5	No	Rain	Cool	Normal	Strong
6	Yes	Overcast	Cool	Normal	Strong
7	No	Sunny	Mild	High	Weak
8	Yes	Sunny	Cool	Normal	Weak
9	Yes	Rain	Mild	Normal	Weak
10	Yes	Sunny	Mild	Normal	Strong
11	Yes	Overcast	Mild	High	Strong
12	Yes	Overcast	Hot	Normal	Weak
13	No	Rain	Mild	High	Strong

```
In [11]: y = list(data['PlayTennis'].values)
X = data.iloc[:,1:].values

print(f'Target Values: {y}')
print(f'Features: \n{X}')

Target Values: ['No', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No']
Features:
[['Sunny' 'Hot' 'High' 'Weak']
 ['Sunny' 'Hot' 'High' 'Strong']
 ['Overcast' 'Hot' 'High' 'Weak']
 ['Rain' 'Mild' 'High' 'Weak']
 ['Rain' 'Cool' 'Normal' 'Weak']
 ['Rain' 'Cool' 'Normal' 'Strong']
 ['Overcast' 'Cool' 'Normal' 'Strong']
 ['Sunny' 'Mild' 'High' 'Weak']
 ['Sunny' 'Cool' 'Normal' 'Weak']
 ['Rain' 'Mild' 'Normal' 'Weak']
 ['Sunny' 'Mild' 'Normal' 'Strong']
 ['Overcast' 'Mild' 'High' 'Strong']
 ['Overcast' 'Hot' 'Normal' 'Weak']
 ['Rain' 'Mild' 'High' 'Strong']]
```

```
['Sunny' 'Mild' 'Normal' 'Strong']  
  
['Rain' 'Mild' 'High' 'Strong']
```

```
self.x, self.y = x, y  
  
self.dim = len(self.x[c])  
  
self.output_dom = {}  
  
for i in range(len(self.x)):   
    for j in range(self.dim):  
  
        self.attrs[j].append(self.x[i][j])  
        if not self.y[i] in self.output_dom.keys():  
            self.output_dom[self.y[i]] = 1  
  
        self.output_dom[self.y[i]] += 1  
        self.data.append([self.x[i], self.y[i]])  
        print(self.attrs)  
        print(self.output_dom)  
#         print(self.data)  
  
prob = self.output_dom[y]/self.N
```

```
On .1 : ,nbcs = NaiveBayesClassifier(X_train, y_train,  
                                     total_cases = len(y_val))
```

```
predictions.append(predict)  
  
if y_val[i] == predict:
```

```
['Sunny', 'Overcast'], ['Hot'], ['High'], ['Weak', 'Strong']]  
['Sunny', 'Overcast', 'Rain'], ['Hot', 'Mild'], ['High', 'Weak', 'Strong']]  
[['Sunny', 'Overcast', 'Rain'], ['Hot', 'Mild', 'Cool'], ['High', 'Normal'], ['Weak', 'Strong']]  
[['Sunny', 'Overcast', 'Rain'], ['Hot', 'Mild', 'Cool'], ['High', 'Normal'], ['Weak', 'Strong']]  
[['Sunny', 'Overcast', 'Rain'], ['Hot', 'Mild', 'Cool'], ['High', 'Normal'], ['Weak', 'Strong']]  
[['Sunny', 'Overcast', 'Rain'], ['Hot', 'Mild', 'Cool'], ['High', 'Normal'], ['Weak', 'Strong']]  
Predicted values: ['No', 'Yes', 'No', 'Yes', 'Yes', 'No']  
Actual values: ['Yes', 'Yes', 'Yes', 'Yes', 'Yes', 'No']
```

5. Linear Regression

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

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
In [2]: dataset = pd.read_csv('salary_data.csv')
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 1].values
```

```
In [3]: 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)
```

```
In [4]: # Fitting Simple Linear Regression to the Training set
from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)
```

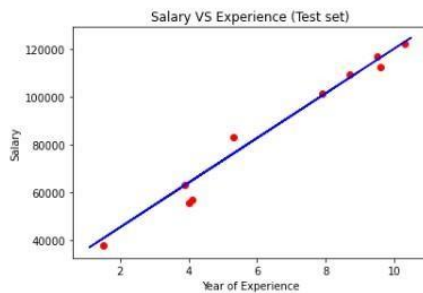
```
Out[4]: LinearRegression()
```

```
In [5]: # Predicting the Test set results
y_pred = regressor.predict(X_test)
```

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



```
In [7]: # 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()
```



```
In [ ]:
```