### 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



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(Autonomous Institution under VTU)

# B. M. S. College of Engineering,

**Bull Temple Road, Bangalore 560019** 

(Affiliated To Visvesvaraya Technological University, Belgaum)

### **Department of Computer Science and Engineering**



#### **CERTIFICATE**

This is to certify that the Lab work entitled "Machine Learning" carried out by HRITIK SINGH (1BM19CS063), 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**

CO1	Ability to apply the different learning algorithms.
CO2	Ability to analyze the learning techniques for given dataset
CO3	Ability to design a model using machine learning to solve a problem.

CO4	Ability to conduct practical experiments to solve problems using appropriate machine learning Techniques.
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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 [14]: import numpy as np
         import pandas as pd
 In [15]: data = pd.read csv("finddata.csv")
         print(data,"\n")
               Time Weather Temperature Company Humidity Goes
         0 Morning Sunny
                                Warm Yes
                                                Mild Yes
         1 Evening Rainy Cold
2 Morning Sunny Moderate
                                         No
                                                Mild
                                                     No
                                        Yes Normal Yes
         3 Evening
                    Sunny Cold Yes High Yes
 In [19]: 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']
          ['Evening' 'Rainy' 'Cold' 'No' 'Mild']
          ['Morning' 'Sunny' 'Moderate' 'Yes' 'Normal']
          ['Evening' 'Sunny' 'Cold' 'Yes' 'High']]
          The target is: ['Yes' 'No' 'Yes' 'Yes']
In [17]: 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 [18]: print("\n The final hypothesis is:",findS(d,target))
           The final hypothesis is: ['?' 'Sunny' '?' 'Yes' '?']
In [ ]:
```

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 [4]: 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("\n 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):
             if target[i] == "yes":
                 for x in range(len(specific h)):
                     if h[x]!= specific h[x]:
                         specific h[x] ='?'
                         general h[x][x] = '?'
                    # print(specific h)
             if target[i] == "no":
                 for x in range(len(specific h)):
                     if h[x]!= specific h[x]:
```

```
print(specific_h)
            print(general_h)
    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)
   #obtaining the final hypothesis
print("\nFinal Specific_h:", s_final, sep="\n")
print("\nFinal General_h:", g_final, sep="\n")
           sky temp humidity
                                               wind water forcast enjoysport
                               normal strong warm
   0 sunny warm
                                                                      same
   1 sunny warm
                                   high strong warm
                                                                        same
                                                                                            ves
                                   high strong warm change
      rainy cold
                                                                                             no
                                   high strong cool change
    3 sunny warm
                                                                                           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']
   Initialization of specific_h and general_h
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'])
     Steps of Candidate Elimination Algorithm 1
    ['sunny' 'warm' 'normal' 'strong' <sup>'</sup>warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?',
'?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
 Steps of Candidate Elimination Algorithm 1
['sunny' 'warm' 'normal' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?']]
Steps of Candidate Elimination Algorithm 2
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?']]
Steps of Candidate Elimination Algorithm 3
['sunny' 'warm' '?' 'strong' 'warm' 'same']
[['sunny', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?', '?', '?', '?', '?']
Steps of Candidate Elimination Algorithm 4
['sunny' 'warm' '?' 'strong' '?' '?']
[['sunny', '?', '?', '?', '?', '?'], ['?', '?', '?', '?'], ['?', '?', '?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?'], ['?', '?', '?']]
Final Specific_h:
['sunny' 'warm' '?' 'strong' '?' '?']
Final General_h:
[['sunny', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?']]
```

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 [24]: import pandas as pd
                     import math
                     import numpy as np
        In [34]: data = pd.read_csv("data.csv")
                     features = [feat for feat in data]
features.remove("answer")
In [37]: class Node:
               def __init__(self):
    self.children = []
    self.value = ""
                    self.isLeaf = False
                    self.pred = ""
In [38]: def entropy(examples):
               pos = 0.0
               neg = 0.0
                    , row in examples.iterrows():
                    if row["answer"] == "yes":
                        pos += 1
                    else:
                       neg += 1
               if pos == 0.0 or neg == 0.0:
                    return 0.0
               else:
                    p = pos / (pos + neg)
                    n = neg / (pos + neg)
return -(p * math.log(p, 2) + n * math.log(n, 2))
In [39]: def info_gain(examples, attr):
               uniq = np.unique(examples[attr])
#print ("\n",uniq)
               gain = entropy(examples)
#print ("\n",gain)
               for u in uniq:
                    subdata = examples[examples[attr] == u]
#print ("\n", subdata)
                    sub_e = entropy(subdata)
                    gain -= (float(len(subdata)) / float(len(examples))) * sub_e
                    #print ("\n",gain)
               return gain
```

```
In [40]: def ID3(examples, attrs):
             root = Node()
             max_gain = 0
             max_feat = ""
             for feature in attrs:
                 #print ("\n",examples)
                  gain = info_gain(examples, feature)
                  if gain > max_gain:
                      max_gain = gain
                      max_feat = feature
             root.value = max_feat
              #print ("\nMax feature attr", max_feat)
             uniq = np.unique(examples[max_feat])
              #print ("\n",unig)
             for u in uniq:
                  #print ("\n",u)
                  subdata = examples[examples[max_feat] == u]
#print ("\n", subdata)
                  if entropy(subdata) == 0.0:
                      newNode = Node()
                      newNode.isLeaf = True
                      newNode.value = u
                      newNode.pred = np.unique(subdata["answer"])
                      root.children.append(newNode)
                 else:
                      dummyNode = Node()
                      dummyNode.value = u
                      new_attrs = attrs.copy()
                      new_attrs.remove(max_feat)
                      child = ID3(subdata, new_attrs)
                      dummyNode.children.append(child)
                      root.children.append(dummyNode)
             return root
  In [41]: def printTree(root: Node, depth=0):
                 for i in range(depth):
    print("\t", end="")
                 print(root.value, end="")
                 if root.isLeaf:
                     print(" -> ", root.pred)
                 print()
                 for child in root.children:
                     printTree(child, depth + 1)
  In [42]: root = ID3(data, features)
            printTree(root)
            outlook
                     overcast -> ['yes']
                     rain
                              wind
                                       strong -> ['no']
                                       weak -> ['yes']
                     sunny
                              humidity
                                       high -> ['no']
                                       normal -> ['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 numpy as np
             import pandas as pd
    In [2]: data = pd.read_csv('/content/dataset.csv')
             data.head()
    Out[2]:
                PlayTennis Outlook Temperature Humidity
                                                      Wind
                      No
                            Sunny
                                         Hot
                                                High
                                                      Weak
             1
                      No
                                         Hot
                                                High Strong
                            Sunny
                      Yes Overcast
                                         Hot
                                                High
                                                      Weak
                      Yes
                             Rain
                                         Mild
                                                High
                                                      Weak
                      Yes
                                                      Weak
                             Rain
                                        Cool
                                               Normal
    In [3]: y = list(data['PlayTennis'].values)
             X = data.iloc[:,1:].values
             print(f'Target Values: {y}')
             print(f'Features: \n{X}')
In [4]: y \text{ train} = y[:8]
         y_val = y[8:]
         X \text{ train} = X[:8]
         X \text{ val} = X[8:]
         print(f"Number of instances in training set: {len(X train)}")
         print(f"Number of instances in testing set: {len(X val)}")
         Number of instances in training set: 8
         Number of instances in testing set: 6
```

```
In [5]: class NaiveBayesClassifier:
              def __init__(self, X, y):
                   self.X, self.y = X, y
                   self.N = len(self.X)
                   self.dim = len(self.X[0])
                   self.attrs = [[] for _ in range(self.dim)]
                   self.output_dom = {}
                  self.data = []
                   for i in range(len(self.X)):
                       for j in range(self.dim):
                           if not self.X[i][j] in self.attrs[j]:
                               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
                       else:
                           self.output_dom[self.y[i]] += 1
                       self.data.append([self.X[i], self.y[i]])
              def classify(self, entry):
                  solve = None
                   max_arg = -1
                   for y in self.output_dom.keys():
                       prob = self.output_dom[y]/self.N
                       for i in range(self.dim):
                           cases = [x \text{ for } x \text{ in self.data if } x[0][i] == entry[i] \text{ and } x[1] == y]
                           n = len(cases)
                           prob *= n/self.N
                       if prob > max_arg:
                           max_arg = prob
                           solve = y
                   return solve
In [6]: nbc = NaiveBayesClassifier(X_train, y_train)
         total_cases = len(y_val)
         good = 0
         bad = 0
         predictions = []
         for i in range(total cases):
             predict = nbc.classify(X val[i])
             predictions.append(predict)
             if y_val[i] == predict:
                  good += 1
             else:
                  bad += 1
         print('Predicted values:', predictions)
         print('Actual values:', y_val)
         print('Total number of testing instances in the dataset:', total cases)
         print('Number of correct predictions:', good)
         print('Number of wrong predictions:', bad)
         print()
         print('Accuracy of Bayes Classifier:', good/total_cases)
         Predicted values: ['No', 'Yes', 'No', 'Yes', 'Yes', 'No']
Actual values: ['Yes', 'Yes', 'Yes', 'Yes', 'No']
         Total number of testing instances in the dataset: 6
         Number of correct predictions: 4
         Number of wrong predictions: 2
         Accuracy of Bayes Classifier: 0.6666666666666666
```

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

```
In [17]: import numpy as np
             import matplotlib.pyplot as plt
              import pandas as pd
             from sklearn.metrics import r2_score
     In [9]: dataset = pd.read_csv('salary_dataset.csv')
             X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 1].values
     In [10]: 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 [11]: # Fitting Simple Linear Regression to the Training set
             from sklearn.linear_model import LinearRegression
             regressor = LinearRegression()
             regressor.fit(X_train, y_train)
    Out[11]: LinearRegression()
    In [15]: # Predicting the Test set results
             y_pred = regressor.predict(X_test)
             y_pred
    Out[15]: array([ 40835.10590871, 123079.39940819, 65134.55626083, 63265.36777221, 115602.64545369, 108125.8914992, 116537.23969801, 64199.96201652, 76349.68719258, 100649.1375447 ])
    In [18]: r2_score(y_test,y_pred)
    Out[18]: 0.9749154407708353
Aur[ +0] . 0.71+7+7+01100777
In [19]: # 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()
                                       Salary VS Experience (Training set)
                  120000
                  100000
                   80000
                   60000
                   40000
                                                                                     10
```

Year of Experience

```
In [14]: # 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()
```

