



## Machine Learning

### Experiment - 1

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

#### Program:-

```
In[4]: import
```

```
In[7]: import pandas as Pd
import numpy as np
```

```
In[8]: data = pd.read_csv('172.16.17.3 \All data \ Enjoy
Sport.csv')
```

Print (data)

	Sky	Air Temp	Humidity	Wind	Water	Forecast	Enjoy Sport
0	Sunny	warm	Normal	strong	warm	same	1
1	Sunny	warm	high	strong	warm	same	1
2	Rainy	cold	high	strong	warm	change	0
3	Sunny	warm	high	strong	warm	change	1

```
In[9]: d = np.array(data)[:, :-1]
```

```
Print ("The attributes are: ", d)
```

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']

```
In[10]: target = np.array(data)[:, -1]
        print("The target is :", target)
        the target is: [1 0 1]
```

```
In[11]: def train(C, t):
        specific_hypothesis = [None] * len(C[0])

        for i in range(len(C)):
            if t[i] == 1:
                for j in range(len(C[i])):
                    if specific_hypothesis[j] is None:
                        specific_hypothesis[j] = C[i][j]
                    else specific_hypothesis[j] != C[i][j]:
                        specific_hypothesis[j] = '?'

        return specific_hypothesis
```

```
In[12]: print("The final hypothesis is :", train(d, target))
        The final hypothesis is: ['sunny', 'warm', '?', 'strong',
```

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EXPERIMENT-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.

Program:-

```
import numpy as np
import pandas as pd

data = pd.read_csv (Path + 'enjoy sport .csv ')
concepts = np. array (data .iloc [:, 0:-1])
Print ("In Instances are : \n", concepts)
target = np. array (data .iloc [:, -1])
Print ("In Target values are :", target)
def learn (concepts, target):
    specific_h = concepts [0]. copy ()
    Print ("In Initialization of specific_h and general_h")
    Print ("In Specific Boundary :", specific_h)
    general_h = [[ "?" for i in range (len (specific_h))] ]
    for i in range (len (specific_h))
        Print ("In Generic Boundary :", general_h)
    for i, h in enumerate (concepts):
        Print ("In Instance ", i+1, "is", h)
        if target [i] == "yes":
            Print ("Instance is positive ")
            for x in range (len (specific_h)):
```





```
if h[x] != Specific_h[x]:
    Specific_h[x] = '?'
    general_h[x][x] = '?'
target[i] = "no":
Print("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 Boundary after ", i+1, "Instance is", Specific_h)
Print("Generic Boundary after ", i+1, "Instance is", general_h)
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, q_final = learn(concepts, targets)
Print("Final Specific_h: ", S_final, sep = "\n")
Print("Final General_h: ", q_final, sep = "\n")
```



Initialization of Specific-h and general-h

Specific Boundary: ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']

Generic Boundary: [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

Instance 1 is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same'] instance is positive.

Specific Boundary after 1 instance is ['sunny' 'warm' 'normal' 'strong' 'warm' 'same']

Generic Boundary after 1 instance is [['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

$$G_1 = G$$

$$S_1 = ['sunny', 'warm', 'normal', 'strong', 'warm', 'same']$$



For Instance 2: Positive output.

$G_2 = G$

$S_2 = ['sunny', 'warm', '?', 'strong', 'warm', 'same']$

For Instance 3:  $< 'rainy', 'cold', 'high', 'strong', 'warm', 'change' >$   
and negative output.

$G_3 = [['sunny', '?', '?', '?', '?', '?'], ['?', 'warm', '?', '?', '?', '?'],$   
 $['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'],$   
 $['?', '?', '?', '?', '?', '?', 'same']]$

$S_3 = S_2$

For Instance 4:  $< 'sunny', 'warm', 'high', 'strong', 'cold', 'change' >$   
and positive output.

$G_4 = G_3$

$S_4 = ['sunny', 'warm', '?', 'strong', '?', '?']$

Print ("Final specific - h: ",  $S_{-final}$ , Sep = "\n")

Print ("Final General - h: ",  $g_{-final}$ , Sep = "\n").

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

Final General - h: [['sunny', '?', '?', '?', '?', '?'],

['?', 'warm', '?', '?', '?', '?']]



EXPERIMENT-3

Write a program to demonstrate the working of Decision tree regression. Use appropriate dataset for decision tree regression.

```
import pandas as pd
from pandas import DataFrame
df_tennis = DataFrame
df_tennis = DataFrame.from_csv(Path)
Print ("In Given play Tennis Data set : \n\n" df_tennis)
```

Given play Tennis Data set :

	PlayTennis	outlook	Temperature	Humidity	wind
0	NO	Sunny	Hot	High	
1	No	Sunny	Hot	High	
2	Yes	Overcast	Hot	High	
3	Yes	Rain	Mild	High	
4	Yes	Rain	Cool	Normal	
5	NO	Rain	Cool	Normal	
6	Yes	Overcast	Cool	High	
7	No	Sunny	Mild	High	
8	Yes	Sunny	Cool	Normal	
9	Yes	Rain	Mild	Normal	
10	Yes	Sunny	Mild	Normal	
11	Yes	Overcast	Mild	High	
12	Yes	Overcast	Hot	Normal	
13	NO	Rain	Mild	High	

```
df_tennis.keys()[0]
```

```
'play Tennis'
```



```
def entropy ( probs ) :  
    import math  
    return sum ( [ - prob * math . log ( Prob , 2 ) for prob in probs ] )
```

```
def entropy_of_list ( a_list ) :  
    from collections import Counter  
    cnt = Counter ( x for x in a_list )  
    num_instances = len ( a_list ) * 1.0  
    Print ( "Number of Instances of the current sub  
            class is {} : ".format ( num ) )  
    probs = [ x / num_instances for x in cnt.values () ]  
    Print ( "Classes : ", min ( cnt ) )  
    Print ( "Probabilities of class {} is {} : ".format  
            ( max ( cnt ) , max ( probs ) ) )  
  
    return entropy ( probs )
```

```
Print ( "INPUT DATA SET FOR ENTROPY CALCULATION :  
       ", df_tennis [ 'play Tennis' ] )
```

```
total_entropy = entropy_of_list ( df_tennis [ 'play  
                                           Tennis' ] )
```

```
Print ( "Total Entropy of Play Tennis Data set : ",  
        total_entropy )
```



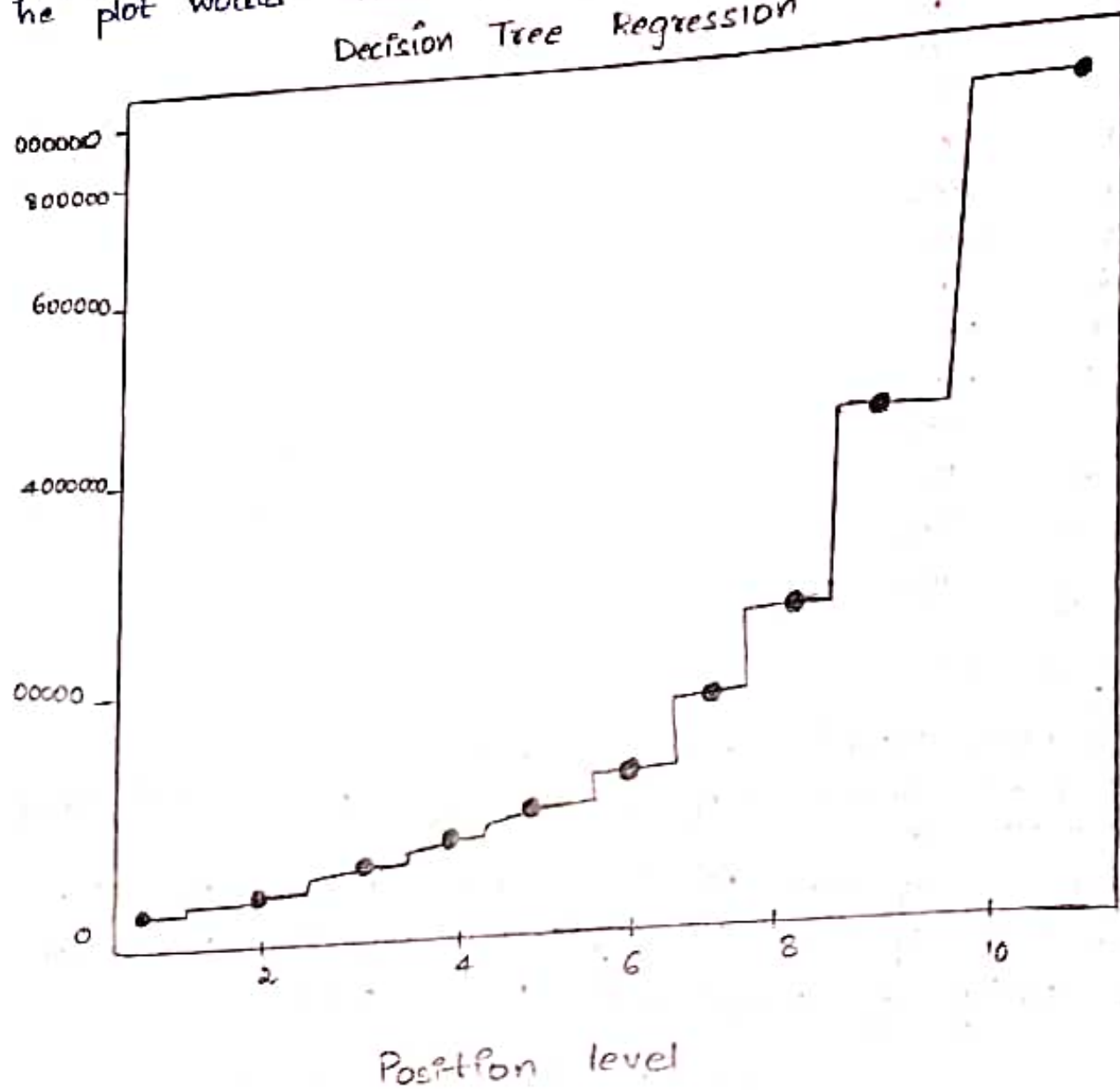


INPUT	DATASET	FOR ENTROPY CALCULATION:
0	NO	
1	Yes	
2	Yes	
3	Yes	
4	Yes	
5	NO	
6	Yes	
7	NO	
8	Yes	
9	Yes	
10	Yes	
11	Yes	
12	Yes	
13	NO	

Name: play Tennis, d-type: object  
Number of instances of the current class is 14.0  
class: NO Yes  
Probabilities of class No is 0.35714285714285715  
Probabilities of class Yes is 0.6428571428571429  
Total Entropy of play Tennis data set:  
 $0.9402857586706309$

the plot would look like this:

Decision Tree Regression



EXPERIMENT - 4

Write a program to demonstrate the working of Decision tree regressor. Use appropriate dataset for Random Forest classification decision tree regressor.

```
import pandas as pd
import matplotlib.pyplot as plt
dataset = pd.read_csv("position-salaries.csv")
X = dataset.iloc[:, 1:2] values
Y = dataset.iloc[:, 2] values
regressor = DecisionTreeRegressor(random_state = 0)
regressor.fit(X, Y)
Y_Pred = regressor.predict([6.5])
X_grid = np.arange(min(X), max(X), 0.01)
X_grid = X_grid.reshape((len(X_grid), 1))
plt.scatter(X, Y, color = 'red')
plt.plot(X_grid, regressor.predict(X_grid), color = 'blue')
plt.title('Decision Tree Regression')
plt.xlabel('Position level')
plt.ylabel('Salary')
plt.show()
```



EXPERIMENT - 5

Write a program to demonstrate the Working of Random Forest classifier. Use appropriate dataset for Random Forest classifier.

```
import pandas as pd
import numpy as np
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix,
Precision_Score, recall_Score, ConfusionMatrixDisplay
from sklearn.model_selection import RandomizedSearchCV,
train_test_split
from scipy.stats import randint
from sklearn.tree import export_graphviz
from IPython.display import Image
import graphviz

bank_data ['default'] = bank_data ['default'].map({ 'no':0,
                                                    'yes':1, 'unknown':0 })

bank_data ['y'] = bank_data ['y'].map({ 'no':0, 'yes':1 })

X = bank_data .drop ('y', axis = 1)
y = bank_data ['y']
X_train, X_test, y_train, y_test = train_test_split (X, y,
                                                    test_size = 0.2)

rf = RandomForestClassifier()
rf.fit (X_train, y_train)
```



```
y_Pred = rf.predict(x_test)
accuracy = accuracy_score(y_test, y_Pred)
Print ("Accuracy:", accuracy)
```

Output:-

Accuracy : 0.888

EXPERIMENT - 6

Write a program to demonstrate the working of logistic Regression classifier. Use appropriate dataset for logistic Regression.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot
as plt
dataset = pd.read_csv("user_data.csv")
x = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:, 4].values
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y,
                                                    test_size = 0.25, random_state = 0)
from sklearn.preprocessing import StandardScaler
sc_x = StandardScaler()
x_train = sc_x.fit_transform(x_train)
x_test = sc_x.transform(x_test)
Print(x_train[0:10, :])
```

Output:

```
[[0.58164944
-0.88670699]
[-0.60673761
1.46173768]
[-0.01254409 -0.5677824]
```





[ - 0.60673761

1.89663484]

[ 1.37390747

- 1.40858358]

[ 1.47293972

0.99784738]

[ 0.08648817

- 0.79972756]

[ - 0.01254409

- 0.24885782]

[ - 0.21060859 - 0.5677284]

[ - 0.21060859

- 0.19087153]