

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

```
import random
```

```
import csv
```

```
attributes = [['sunny', 'rainy'], ['warm', 'cold'], ['normal', 'high'], ['strong', 'weak'], ['warm', 'cool'], ['same', 'change']]
```

```
print(attributes)
```

```
num_attributes = len(attributes)
```

```
print(num_attributes)
```

```
print("\n the most general hypothesis: ['?', '?', '?', '?', '?', '?']")
```

```
print("\n the most specific hypothesis: ['0', '0', '0', '0', '0', '0']")
```

```
h = []
```

```
print("\n the given training dataset")
```

```
with open('C:\Users\ADMIN\Desktop\ALGO\data.csv', 'r') as csvfile:
```

```
    reader = csv.reader(csvfile)
```

```
    for row in reader:
```

```
        h.append(row)
```

```
    print(row)
```

```
print("\n the initial value of hypothesis:")
```

```
h = ['0'] * num_attributes
```

```
print(h)
```

```
for j in range(0, num_attributes):
```

```
a[i] = a[i][i]
```

```
for i is range(1, len(a));
```

```
if (a[i][sum-attributes] == 'yes');
```

```
for j is range(sum-attributes):
```

```
if (b[j] == '0' or b[j] == a[i][j]):
```

```
b[j] = a[i][j]
```

```
else:
```

```
b[j] = '1'
```

```
Print ("10 for training examples: {0} the  
hypothesis") format (i+1, b)
```

Dataset:

Sl. No	sky	Air temp	Humidity	wind	water	Forecast	Enjoy Sport
1	Sunny	warm	normal	strong	warm	same	Yes
2	Sunny	warm	high	strong	warm	same	Yes
3	Rainy	cold	high	strong	warm	change	No
4	Sunny	warm	high	strong	cool	change	Yes

output:-

['Sunny', 'Rainy'], ['warm', 'cold'], ['normal', 'high'],  
 ['strong', 'weak'], ['warm', 'cool'], ['same', 'change']

b.

the most general hypothesis: ['?', '?', '?', '?', '?', '?']

the most specific hypothesis: ['0', '0', '0', '0', '0', '0']

The given training dataset

['sky', 'Air temp', 'Humidity', 'wind', 'water', 'forecast',  
 'Enjoy sport']

['Sunny', 'warm', 'normal', 'strong', 'warm', 'same', 'Yes']

['Sunny', 'warm', 'high', 'strong', 'warm', 'same', 'Yes']

['Rainy', 'cold', 'high', 'strong', 'warm', 'change', 'Yes']

The initial hypothesis is:

['0', '0', '0', '0', '0', '0']

for training example:  $\{0\}$  the hypotheses  $\mathcal{H}$   
 $['sunni', 'wani', 'q', 'strong', '?', '?']$



2. For a given set of set of training data example 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 example.

```

import csv
with open('c:\\users\\Admin\\Desktop\\algorithms\\algorithms.csv')
as f:
    csv_file = csv.reader(f)
    data = list(csv_file)
    print(data)
    s = data[1][-1]
    g = ['?'] for i in range(len(s)) for j in range(
        len(s))
    for i in data:
        if i[-1] == "yes":
            for j in range(len(s)):
                if i[j] != s[j]:
                    s[j] = '?'
                    g[j][j] = '?'
        elif i[-1] == "No":
            for j in range(len(s)):
                if i[j] == s[j]:
                    g[j][j] = s[j]
            else:
                g[j][j] = "?"
    print("10 steps of candidate elimination algorithm: data.index(i)+1)

```

```
print(s)
```

```
print(g)
```

```
gh = []
```

```
for i in g:
```

```
    for j in i:
```

```
        if 'j' != 'i':
```

```
            gh.append(i)
```

```
        break
```

```
print(f"\n Final specific hypothesis: {s}, {g}")
```

```
print(f"\n Final general hypothesis: {s} {gh}")
```

Dataset

Sunny	warm	normal	strong	warm	same	Yes
Sunny	warm	high	strong	warm	same	Yes
Rainy	cold	high	strong	warm	change	No
Sunny	warm	high	strong	warm	change	Yes

Output:

steps of candidate elimination algorithm 1

['sunny', 'warm', '?', 'strong', '?', '?']

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

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

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

steps of candidate elimination algorithm 2.

['sunny', 'warm', '?', 'strong', '?', '?']

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

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

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

steps of candidate elimination algorithm 3.

['sunny', 'warm', '?', 'strong', '?', '?']

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

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

steps of candidate elimination algorithm &

[<sup>1</sup>'empty', <sup>1</sup>'weak', <sup>1</sup>'strong', <sup>1</sup>'?']

[<sup>1</sup>'empty', <sup>1</sup>'?', <sup>1</sup>'?', <sup>1</sup>'?'], [<sup>2</sup>'weak', <sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?'],

[<sup>1</sup>'?', <sup>1</sup>'?', <sup>1</sup>'?', <sup>1</sup>'?'], [<sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?'], [<sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?'],

[<sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?', <sup>2</sup>'?']

final specific hypothesis

[<sup>1</sup>'empty', <sup>1</sup>'weak', <sup>1</sup>'strong', <sup>1</sup>'?']

final general hypothesis

[<sup>1</sup>'empty', <sup>1</sup>'?', <sup>1</sup>'?', <sup>1</sup>'?'], [<sup>1</sup>'weak', <sup>1</sup>'?', <sup>1</sup>'?', <sup>1</sup>'?']



5. 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.

```
import pandas as pd.
```

```
from pandas import DataFrame.
```

```
df_tennis = pd.read_csv('c:\Users\Adwin\Desktop\Albin\play_tennis.csv')
```

```
attribute_names = list(df_tennis.columns)
```

```
attribute_names.remove('play Tennis').
```

```
print(attribute_names).
```

```
def entropy_of_list(lst):
```

```
    from collections import Counter.
```

```
    count = Counter(x for x in lst)
```

```
    num_instances = len(lst) * 1.
```

```
    probs = [x/num_instances for x in count.values]
```

```
    return entropy(probs).
```

```
def entropy(probs):
```

```
    import math
```

```
    return sum([-prob * math.log(prob, 2) for prob in probs]).
```

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

```
def information_gain(df, split_attribute_name, target_attribute_name, tree=0):
```

```
    df_split = df.groupby(split_attribute_name)
```

```
    nobs = len(df.index) * 1.
```

```

df_agg_ent = df_split_agg({target_attribute_name:
    [entropy_of_list, lambda x: len(x)/nobs]})
df_agg_ent.columns = ['entropy', 'prop observations']
new_entropy = sum(df_agg_ent['entropy'] * df_agg_ent['prop observations'])
old_entropy = entropy_of_list(df[target_attribute_name])
print(split_attribute_name, 'Eq 1', old_entropy - new_entropy)
return old_entropy - new_entropy
def ids(df, target_attribute_name, attribute_names, default_class = None):
    from collections import Counter
    count = Counter(x for x in df[target_attribute_name])
    if len(count) == 1:
        return None
    elif df.empty or not attribute_names:
        return default_class
    else:
        default_class = max(count.keys())
        gain = [information_gain(df, attr, target_attribute_name) for attr in attribute_names]
        print(gain)
        index_of_max = gain.index(max(gain))
        best_attr = attribute_names[index_of_max]
        tree = {best_attr: {}}
        remaining_attribute_names = [i for i in attribute_names

```

name if i) = best\_attr]

for attr\_val, data\_subset is df.groupby.  
(best\_attr):

subtree = ID3(data\_subset, target, attribute-  
name, remaining\_attribute\_names-default-  
class).

tree[best\_attr][attr\_val] = subtree

return.

from pprint import pprint.

tree = ID3(df\_tennis, 'playTennis', attribute-  
names).

print("\n\n The resultant decision tree is :  
") pprint(tree)



Expt. No.....

output

['outlook', 'temperature', 'humidity', 'wind']

outlook 10 = 0.2467498197744391

temperature 14 = 0.029222565638954649.

humidity 19 = 0.15183550136234136.

wind 19 = 0.04812703040826927.

temperature 14 = 0.01927309402197489.

humidity 19 = 0.0199730400197489.

wind 19 = 0.9209505944546686.

temperature 14 = 0.570989505944546686.

humidity 19 = 0.9709505944546686.

wind 19 = 0.01997309400197489.

The result of decision tree is

{ 'outlook': { 'overcast': 'yes',

'rain': { 'wind': { 'strong': 'no', 'weak': 'yes' },

'sunny': { 'humidity': { 'high': 'no', 'normal':

'yes' } }