

3. Write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate dataset 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('lab3.csv')
```

```
attribute_name1 = list(df_tennis.columns)
attribute_name1.remove('Play Tennis')
print(attribute_name1)
```

```
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, trace = 0):
    df_split = df.groupby(split_attribute_name)
    nob = len(df.index) * 1
    df_agg_ent = df_split.agg({'target_attribute_name':
                               [entropy_of_list, lambda x: len(x)/nob]})
    df_agg_ent.columns = ['Entropy', 'propobservations']
    new_entropy = sum(df_agg_ent['Entropy'] * df_agg_ent
                       ['propobservations'])
    old_entropy = entropy_of_list(df[target_attribute_name])
    print(split_attribute_name, 'IG: ', old_entropy -
                                                new_entropy)

    return old_entropy - new_entropy
```

```
def id3(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 next(iter(count))
    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
```



```
index_of_max = gain.index(max(gain))
belt_attr = attribute_names[index_of_max]
```

```
tree = {belt_attr: {}}
```

```
remaining_attribute_names = [i for i in attribute_names
                              if i != belt_attr]
```

```
for attr_val, data_subset in df.groupby(belt_attr):
    subtree = id3(data_subset, target_attribute_name,
                  remaining_attribute_names, default_class)
    tree[belt_attr][attr_val] = subtree
return tree
```

```
from pprint import pprint
tree = id3(df_tennis, 'Play Tennis', attribute_names)
print("\n The Resultant Decision Tree is: \n")
pprint(tree)
```



Output:-

['Outlook', 'Temperature', 'Humidity', 'Wind']

Outlook IG: 0.2467498197744391

Temperature IG: 0.0299222565658954647

Humidity IG: 0.15183550136234136

Wind IG: 0.04812703040826927

Temperature IG: 0.01997309402197489

Humidity IG: 0.01997309402197489

Wind IG: 0.9709505944546686

Temperature IG: 0.5709505944546686

Humidity IG: 0.9709505944546686

Wind IG: 0.01997309402197489

The Resultant Decision Tree is:

{ 'Outlook': { 'Overcast': 'Yes',

'Rain': { 'Wind': { 'Strong': 'No', 'Weak': 'Yes' } },

'Sunny': { 'Humidity': { 'High': 'No', 'Normal': 'Yes' } } }