# 4) 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

#### **Understanding the ID3 Decision Tree Algorithm**

**Think of the ID3 algorithm as a game of 20 Questions.** It's like trying to guess something by asking yes/no questions. The goal is to build a tree of questions that helps us make the best guess.

#### The Data

Imagine we have a list of objects (like fruits) and for each object, we have some information (attributes) about them. For example, for fruits, the attributes could be color, size, and taste. Each fruit is either "good" or "bad."

### **Building the Tree**

We start with a big question at the top of the tree, like "Is it red?" This is our first question.

- If the answer is "Yes," we ask another question, like "Is it small?" If "No," we might guess
   "Apple."
- If the answer is "No" to the first question, we might guess "Banana."

We keep asking questions and making guesses until we're sure about the answer (like guessing the fruit correctly).

## Testing with a New Object

Now, imagine you have a new fruit that you've never seen before. You don't know what it is, but you have some information about it (its color, size, and taste).

- You start at the top of your tree with the first question, like "Is it red?" If it's "Yes," you move to the next question, and so on.
- You follow the tree until you reach a guess, like "Apple."

## **Understanding the Program**

The program reads a spreadsheet with the information about fruits and their "good" or "bad" classification.

It uses the ID3 algorithm to build the tree of questions and guesses.

 Then, it defines a new, unknown fruit and uses the tree to guess what it is (whether it's "good" or "bad").

#### The Result

The program tells you what it guessed for the new fruit based on the questions and answers in the tree. This is like a computer making a guess about something it's never seen before.

**In a nutshell:** The program uses a tree of questions to guess what something is, just like playing 20 Questions. It uses information about known things to make the best guess about something new.

```
import pandas as pd
import numpy as np
import pprint
eps = np.finfo(float).eps
from numpy import log2 as log
def find_entropy(df):
   Class = df.keys()[-1]
   entropy = 0
    values = df[Class].unique()
    for value in values:
        fraction = df[Class].value counts()[value] / len(df[Class])
        entropy += -fraction * np.log2(fraction)
    return entropy
def find_entropy_attribute(df, attribute):
   Class = df.keys()[-1]
    target_variables = df[Class].unique()
    variables = df[attribute].unique()
    entropy2 = 0
    for variable in variables:
        entropy = 0
        for target variable in target variables:
            num = len(df[attribute][df[attribute] == variable][df[Class] == target_variable]
            den = len(df[attribute][df[attribute] == variable])
            fraction = num / (den + eps)
            entropy += -fraction * log(fraction + eps)
        fraction2 = den / len(df)
        entropy2 += -fraction2 * entropy
    return abs(entropy2)
def find_winner(df):
   Entropy_att = []
    IG = []
    for key in df.keys()[:-1]:
        Entropy_att.append(find_entropy_attribute(df, key))
        IG.append(find_entropy(df) - find_entropy_attribute(df, key))
    return df.keys()[:-1][np.argmax(IG)]
def get_subtable(df, node, value):
```

```
return df[df[node] == value].reset_index(drop=True)
def buildTree(df, tree=None):
   Class = df.keys()[-1]
   node = find_winner(df)
   attValue = np.unique(df[node])
   if tree is None:
        tree = {}
        tree[node] = \{\}
    for value in attValue:
        subtable = get_subtable(df, node, value)
        clValue, counts = np.unique(subtable[Class], return_counts=True)
        if len(counts) == 1:
           tree[node][value] = clValue[0]
           tree[node][value] = buildTree(subtable)
    return tree
df = pd.read_csv('playtennis.csv') # Make sure 'playtennis.csv' contains your dataset
print("\nGiven Play Tennis Data Set:\n\n", df)
tree = buildTree(df)
print('The resultant decision tree is:')
pprint.pprint(tree)
test = {'Outlook': 'Sunny', 'Temperature': 'Hot', 'Humidity': 'High', 'Wind': 'Weak'}
def func(test, tree, default=None):
   attribute = next(iter(tree))
   print(attribute)
    if test[attribute] in tree[attribute].keys():
        print(tree[attribute].keys())
        print(test[attribute])
        result = tree[attribute][test[attribute]]
        if isinstance(result, dict):
           return func(test, result)
        else:
           return result
   else:
        return default
ans = func(test, tree)
print(ans)
Given Play Tennis Data Set:
     Outlook Temperature Humidity
                                   Wind Play Tennis
                  Hot
0
      Sunny
                           High
                                   Weak
                                                  No
1
      Sunny
                    Hot
                            High Strong
                                                  No
2
   0vercast
                   Hot
                           High
                                  Weak
                                                 Yes
                  Mild
3
       Rain
                           High
                                                 Yes
                                  Weak
```

4

5

Rain

Rain

Overcast

Cool Normal Weak

Cool Normal Strong

Cool Normal Strong

Yes

No

Yes

```
OVC: CU3C
                 COOT 1401 11101 2 C1 O116
                                            . . .
7
                Mild High Weak
     Sunny
                                            No
8
                Cool Normal Weak
    Sunny
                                            Yes
                Mild Normal Weak
9
      Rain
                                            Yes
                Mild Normal Strong
10
    Sunny
                                            Yes
                Mild High Strong
11 Overcast
                                            Yes
12 Overcast
                 Hot Normal
                                            Yes
                               Weak
13
      Rain
                 Mild High Strong
                                             No
The resultant decision tree is:
{'Outlook': {'Overcast': 'Yes',
           'Rain': {'Wind': {'Strong': 'No', 'Weak': 'Yes'}},
           'Sunny': {'Humidity': {'High': 'No', 'Normal': 'Yes'}}}
Outlook
dict_keys(['Overcast', 'Rain', 'Sunny'])
Sunny
Humidity
dict_keys(['High', 'Normal'])
High
No
```