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]
        else:
            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
                           High
0
      Sunny
                                   Weak
                                                No
1
                           High Strong
      Sunny
                   Hot
                                                No
2
   Overcast
                   Hot
                          High
                                 Weak
                                               Yes
3
       Rain
                  Mild
                           High
                                   Weak
                                               Yes
4
       Rain
                  Cool Normal
                                 Weak
                                               Yes
5
       Rain
                  Cool Normal Strong
                                               No
                                               Yes
6
   Overcast
                  Cool
                         Normal Strong
7
      Sunny
                  Mild
                         High
                                Weak
                                               No
8
      Sunny
                  Cool Normal
                                               Yes
                                   Weak
9
                  Mild Normal
       Rain
                                   Weak
                                               Yes
10
      Sunny
                  Mild
                         Normal Strong
                                               Yes
11 Overcast
                  Mild
                         High Strong
                                               Yes
```

```
12 Overcast
                  Hot Normal Weak
                                                Yes
13
                   Mild
                           High Strong
       Rain
                                                 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
```