import pandas as pd

df = pd.read\_csv('/home/cmruuser/Downloads/tennisdata.csv')

print("\n Input Data Set is:\n", df)

t = df.keys()[-1]

print('Target Attribute is: ', t)

# Get the attribute names from input dataset

attribute\_names = list(df.keys())

#Remove the target attribute from the attribute names list

attribute\_names.remove(t)

print('Predicting Attributes: ', attribute\_names)

#Function to calculate the entropy of collection S

import math

def entropy(probs):

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

#Function to calulate the entropy of the given Data Sets/List with

#respect to target attributes

def entropy\_of\_list(ls,value):

from collections import Counter

cnt = Counter(x for x in ls)# Counter calculates the propotion of class

print('Target attribute class count(Yes/No)=',dict(cnt))

total\_instances = len(ls)

print("Total no of instances/records associated with {0} is: {1}".format(value,total\_instances ))

probs = [x / total\_instances for x in cnt.values()] # x means no of YES/NO

print("Probability of Class {0} is: {1:.4f}".format(min(cnt),min(probs)))

print("Probability of Class {0} is: {1:.4f}".format(max(cnt),max(probs)))

return entropy(probs) # Call Entropy

def information\_gain(df, split\_attribute, target\_attribute,battr):

print("\n\n-----Information Gain Calculation of ",split\_attribute, " --------")

df\_split = df.groupby(split\_attribute) # group the data based on attribute values

glist=[]

for gname,group in df\_split:

print('Grouped Attribute Values \n',group)

glist.append(gname)

glist.reverse()

nobs = len(df.index) \* 1.0

df\_agg1=df\_split.agg({target\_attribute:lambda x:entropy\_of\_list(x, glist.pop())})

df\_agg2=df\_split.agg({target\_attribute :lambda x:len(x)/nobs})

df\_agg1.columns=['Entropy']

df\_agg2.columns=['Proportion']

# Calculate Information Gain:

new\_entropy = sum( df\_agg1['Entropy'] \* df\_agg2['Proportion'])

if battr !='S':

old\_entropy = entropy\_of\_list(df[target\_attribute],'S-'+df.iloc[0][df.columns.get\_loc(battr)])

else:

old\_entropy = entropy\_of\_list(df[target\_attribute],battr)

return old\_entropy - new\_entropy

def id3(df, target\_attribute, attribute\_names, default\_class=None,default\_attr='S'):

from collections import Counter

cnt = Counter(x for x in df[target\_attribute])# class of YES /NO

## First check: Is this split of the dataset homogeneous?

if len(cnt) == 1:

return next(iter(cnt)) # next input data set, or raises StopIteration when EOF is hit.

## Second check: Is this split of the dataset empty? if yes, return a default value

elif df.empty or (not attribute\_names):

return default\_class # Return None for Empty Data Set

## Otherwise: This dataset is ready to be devied up!

else:

# Get Default Value for next recursive call of this function:

default\_class = max(cnt.keys()) #No of YES and NO Class

# Compute the Information Gain of the attributes:

gainz=[]

for attr in attribute\_names:

ig= information\_gain(df, attr, target\_attribute,default\_attr)

gainz.append(ig)

print('Information gain of ',attr,' is : ',ig)

index\_of\_max = gainz.index(max(gainz)) # Index of Best Attribute

best\_attr = attribute\_names[index\_of\_max] # Choose Best Attribute to split on

print("\nAttribute with the maximum gain is: ", best\_attr)

# Create an empty tree, to be populated in a moment

tree = {best\_attr:{}} # Initiate the tree with best attribute as a node

remaining\_attribute\_names =[i for i in attribute\_names if i != best\_attr]

# Split dataset-On each split, recursively call this algorithm.Populate the empty tree with subtrees, which

# are the result of the recursive call

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

subtree = id3(data\_subset,target\_attribute, remaining\_attribute\_names,default\_class,best\_attr)

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

return tree

from pprint import pprint

tree = id3(df,t,attribute\_names)

print("\nThe Resultant Decision Tree is:")

print(tree)

OUTPUT

Input Data Set is:

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong 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

6 overcast cool normal Strong yes

7 sunny mild high weak no

8 sunny cool normal weak yes

9 rain mild normal weak yes

10 sunny mild normal Strong yes

11 overcast mild high Strong yes

12 overcast hot normal weak yes

Target Attribute is: Target

Predicting Attributes: ['Outlook', 'Temperature', 'Humidity', 'Wind']

-----Information Gain Calculation of Outlook --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

2 overcast hot high weak yes

6 overcast cool normal Strong yes

11 overcast mild high Strong yes

12 overcast hot normal weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

3 rain mild high weak yes

4 rain cool normal weak yes

5 rain cool normal Strong no

9 rain mild normal weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong no

7 sunny mild high weak no

8 sunny cool normal weak yes

10 sunny mild normal Strong yes

Target attribute class count(Yes/No)= {'yes': 4}

Total no of instances/records associated with rain is: 4

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}

Total no of instances/records associated with overcast is: 4

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}

Total no of instances/records associated with sunny is: 5

Probability of Class no is: 0.4000

Probability of Class yes is: 0.6000

Target attribute class count(Yes/No)= {'no': 4, 'yes': 9}

Total no of instances/records associated with S is: 13

Probability of Class no is: 0.3077

Probability of Class yes is: 0.6923

Information gain of Outlook is : 0.2674250655956548

-----Information Gain Calculation of Temperature --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

4 rain cool normal weak yes

5 rain cool normal Strong no

6 overcast cool normal Strong yes

8 sunny cool normal weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong no

2 overcast hot high weak yes

12 overcast hot normal weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

3 rain mild high weak yes

7 sunny mild high weak no

9 rain mild normal weak yes

10 sunny mild normal Strong yes

11 overcast mild high Strong yes

Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}

Total no of instances/records associated with hot is: 4

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Target attribute class count(Yes/No)= {'no': 2, 'yes': 2}

Total no of instances/records associated with cool is: 4

Probability of Class no is: 0.5000

Probability of Class yes is: 0.5000

Target attribute class count(Yes/No)= {'yes': 4, 'no': 1}

Total no of instances/records associated with mild is: 5

Probability of Class no is: 0.2000

Probability of Class yes is: 0.8000

Target attribute class count(Yes/No)= {'no': 4, 'yes': 9}

Total no of instances/records associated with S is: 13

Probability of Class no is: 0.3077

Probability of Class yes is: 0.6923

Information gain of Temperature is : 0.05551064235231107

-----Information Gain Calculation of Humidity --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong no

2 overcast hot high weak yes

3 rain mild high weak yes

7 sunny mild high weak no

11 overcast mild high Strong yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

4 rain cool normal weak yes

5 rain cool normal Strong no

6 overcast cool normal Strong yes

8 sunny cool normal weak yes

9 rain mild normal weak yes

10 sunny mild normal Strong yes

12 overcast hot normal weak yes

Target attribute class count(Yes/No)= {'no': 3, 'yes': 3}

Total no of instances/records associated with high is: 6

Probability of Class no is: 0.5000

Probability of Class yes is: 0.5000

Target attribute class count(Yes/No)= {'yes': 6, 'no': 1}

Total no of instances/records associated with normal is: 7

Probability of Class no is: 0.1429

Probability of Class yes is: 0.8571

Target attribute class count(Yes/No)= {'no': 4, 'yes': 9}

Total no of instances/records associated with S is: 13

Probability of Class no is: 0.3077

Probability of Class yes is: 0.6923

Information gain of Humidity is : 0.11036014405977657

-----Information Gain Calculation of Wind --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

1 sunny hot high Strong no

5 rain cool normal Strong no

6 overcast cool normal Strong yes

10 sunny mild normal Strong yes

11 overcast mild high Strong yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

2 overcast hot high weak yes

3 rain mild high weak yes

4 rain cool normal weak yes

7 sunny mild high weak no

8 sunny cool normal weak yes

9 rain mild normal weak yes

12 overcast hot normal weak yes

Target attribute class count(Yes/No)= {'no': 2, 'yes': 3}

Total no of instances/records associated with Strong is: 5

Probability of Class no is: 0.4000

Probability of Class yes is: 0.6000

Target attribute class count(Yes/No)= {'no': 2, 'yes': 6}

Total no of instances/records associated with weak is: 8

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Target attribute class count(Yes/No)= {'no': 4, 'yes': 9}

Total no of instances/records associated with S is: 13

Probability of Class no is: 0.3077

Probability of Class yes is: 0.6923

Information gain of Wind is : 0.01780102730053701

Attribute with the maximum gain is: Outlook

-----Information Gain Calculation of Temperature --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

4 rain cool normal weak yes

5 rain cool normal Strong no

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

3 rain mild high weak yes

9 rain mild normal weak yes

Target attribute class count(Yes/No)= {'yes': 1, 'no': 1}

Total no of instances/records associated with cool is: 2

Probability of Class no is: 0.5000

Probability of Class yes is: 0.5000

Target attribute class count(Yes/No)= {'yes': 2}

Total no of instances/records associated with mild is: 2

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}

Total no of instances/records associated with S-rain is: 4

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Information gain of Temperature is : 0.31127812445913283

-----Information Gain Calculation of Humidity --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

3 rain mild high weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

4 rain cool normal weak yes

5 rain cool normal Strong no

9 rain mild normal weak yes

Target attribute class count(Yes/No)= {'yes': 1}

Total no of instances/records associated with high is: 1

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'yes': 2, 'no': 1}

Total no of instances/records associated with normal is: 3

Probability of Class no is: 0.3333

Probability of Class yes is: 0.6667

Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}

Total no of instances/records associated with S-rain is: 4

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Information gain of Humidity is : 0.12255624891826566

-----Information Gain Calculation of Wind --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

5 rain cool normal Strong no

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

3 rain mild high weak yes

4 rain cool normal weak yes

9 rain mild normal weak yes

Target attribute class count(Yes/No)= {'no': 1}

Total no of instances/records associated with Strong is: 1

Probability of Class no is: 1.0000

Probability of Class no is: 1.0000

Target attribute class count(Yes/No)= {'yes': 3}

Total no of instances/records associated with weak is: 3

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'yes': 3, 'no': 1}

Total no of instances/records associated with S-rain is: 4

Probability of Class no is: 0.2500

Probability of Class yes is: 0.7500

Information gain of Wind is : 0.8112781244591328

Attribute with the maximum gain is: Wind

-----Information Gain Calculation of Temperature --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

8 sunny cool normal weak yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong no

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

7 sunny mild high weak no

10 sunny mild normal Strong yes

Target attribute class count(Yes/No)= {'yes': 1}

Total no of instances/records associated with hot is: 1

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'no': 2}

Total no of instances/records associated with cool is: 2

Probability of Class no is: 1.0000

Probability of Class no is: 1.0000

Target attribute class count(Yes/No)= {'no': 1, 'yes': 1}

Total no of instances/records associated with mild is: 2

Probability of Class no is: 0.5000

Probability of Class yes is: 0.5000

Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}

Total no of instances/records associated with S-sunny is: 5

Probability of Class no is: 0.4000

Probability of Class yes is: 0.6000

Information gain of Temperature is : 0.5709505944546686

-----Information Gain Calculation of Humidity --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

1 sunny hot high Strong no

7 sunny mild high weak no

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

8 sunny cool normal weak yes

10 sunny mild normal Strong yes

Target attribute class count(Yes/No)= {'no': 3}

Total no of instances/records associated with high is: 3

Probability of Class no is: 1.0000

Probability of Class no is: 1.0000

Target attribute class count(Yes/No)= {'yes': 2}

Total no of instances/records associated with normal is: 2

Probability of Class yes is: 1.0000

Probability of Class yes is: 1.0000

Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}

Total no of instances/records associated with S-sunny is: 5

Probability of Class no is: 0.4000

Probability of Class yes is: 0.6000

Information gain of Humidity is : 0.9709505944546686

-----Information Gain Calculation of Wind --------

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

1 sunny hot high Strong no

10 sunny mild normal Strong yes

Grouped Attribute Values

Outlook Temperature Humidity Wind Target

0 sunny hot high weak no

7 sunny mild high weak no

8 sunny cool normal weak yes

Target attribute class count(Yes/No)= {'no': 1, 'yes': 1}

Total no of instances/records associated with Strong is: 2

Probability of Class no is: 0.5000

Probability of Class yes is: 0.5000

Target attribute class count(Yes/No)= {'no': 2, 'yes': 1}

Total no of instances/records associated with weak is: 3

Probability of Class no is: 0.3333

Probability of Class yes is: 0.6667

Target attribute class count(Yes/No)= {'no': 3, 'yes': 2}

Total no of instances/records associated with S-sunny is: 5

Probability of Class no is: 0.4000

Probability of Class yes is: 0.6000

Information gain of Wind is : 0.01997309402197489

Attribute with the maximum gain is: Humidity

The Resultant Decision Tree is:

{'Outlook': {'overcast': 'yes', 'rain': {'Wind': {'Strong': 'no', 'weak': 'yes'}}, 'sunny': {'Humidity': {'high': 'no', 'normal': 'yes'}}}}