# Ex.no:2 Decision Tree ID3 Algorithm In Python

## AIM:

A program to demonstrate the working of the decision tree based ID3 algorithm.

## ALGORITHM:

```
ID3(Examples, Target_attribute, Attributes)
```

Examples are the training examples.

Target\_attribute is the attribute whose value is to be predicted by the tree.

Attributes is a list of other attributes that may be tested by the learned decision tree.

Returns a decision tree that correctly classifies the given Examples.

Create a Root node for the tree

If all Examples are positive, Return the single-node tree Root, with label = +

If all Examples are negative, Return the single-node tree Root, with label = -

If Attributes is empty, Return the single-node tree Root, with label = most common value of Target attribute in Examples

### Otherwise Begin

A  $\leftarrow$  the attribute from Attributes that best\* classifies Examples

The decision attribute for Root  $\leftarrow$  A

For each possible value, vi, of A,

Add a new tree branch below Root, corresponding to the test  $\mathbf{A} = \mathbf{vi}$ 

Let Examples vi, be the subset of Examples that have value vi for  ${\tt A}$ 

If Examples vi , is empty

Then below this new branch add a leaf node with label = most common value of Target attribute in

#### Examples

Else

below this new branch add the subtree

ID3(Examples vi, Targe\_tattribute, Attributes - {A}))

```
End
Return Root
```

(\*NOTE: After completion of each block use the separate code IDE to perform the decision tree.

After completion on each block click on the code button to code the next block) Don't write the note in the observation note.

## PROGRAM:

```
import pandas as pd
import math
import numpy as np
data = pd.read csv("Dataset/4-dataset.csv")
features = [feat for feat in data]
features.remove("answer")
class Node:
def init (self):
self.children = []
self.value = ""
self.isLeaf = False
self.pred = ""
def entropy(examples):
pos = 0.0
neg = 0.0
for , row in examples.iterrows():
if row["answer"] == "yes":
pos += 1
else:
neq += 1
if pos == 0.0 or neg == 0.0:
return 0.0
else:
p = pos / (pos + neg)
n = neg / (pos + neg)
return -(p * math.log(p, 2) + n * math.log(n, 2))
```

```
def info gain (examples, attr):
uniq = np.unique(examples[attr])
#print ("\n",uniq)
gain = entropy(examples)
#print ("\n",gain)
for u in uniq:
subdata = examples[examples[attr] == u]
#print ("\n", subdata)
sub e = entropy(subdata)
gain -= (float(len(subdata)) / float(len(examples))) *
sub e
#print ("\n",gain)
return gain
def ID3(examples, attrs):
root = Node()
\max gain = 0
max feat = ""
for feature in attrs:
#print ("\n", examples)
gain = info gain(examples, feature)
if gain > max gain:
max_gain = gain
max feat = feature
root.value = max feat
#print ("\nMax feature attr", max feat)
uniq = np.unique(examples[max feat])
#print ("\n",uniq)
for u in uniq:
#print ("\n",u)
subdata = examples[examples[max feat] == u]
#print ("\n", subdata)
if entropy(subdata) == 0.0:
newNode = Node()
newNode.isLeaf = True
newNode.value = u
 newNode.pred = np.unique(subdata["answer"])
root.children.append(newNode)
else:
dummyNode = Node()
dummyNode.value = u
```

```
new attrs = attrs.copy()
  new attrs.remove(max feat)
  child = ID3(subdata, new attrs)
dummyNode.children.append(child)
      root.children.append(dummyNode)
return root
def printTree(root: Node, depth=0):
for i in range (depth):
print("\t", end="")
print(root.value, end="")
if root.isLeaf:
      print(" -> ", root.pred)
print()
for child in root.children:
printTree(child, depth + 1)
def classify(root: Node, new):
for child in root.children:
if child.value == new[root.value]:
if child.isLeaf:
            print ("Predicted Label for new example",
new," is:", child.pred)
exit
else:
classify (child.children[0], new)
root = ID3(data, features)
print("Decision Tree is:")
printTree(root)
print ("----")
new = {"outlook":"sunny", "temperature":"hot",
"humidity": "normal", "wind": "strong"}
classify (root, new)
```

```
Decision Tree is:
outlook
    overcast -> ['yes']

rain
    wind
    strong -> ['no']
    weak -> ['yes']

sunny
    humidity
    high -> ['no']
    normal -> ['yes']
```

Dataset Link: **CLICK HERE**