12/10/20

EXPERIMENT NO. 3

1KS17CS030

AIM: Write a program to demonstrate the working of decision tree based ID3 algorithm. Use an appropriate dataset for building the decision tree & apply this knowledge to classify a new sample

Dataset:

Outlook	Temparature	Humidity	Wind	PlayTennis
Sunny	Hot	High	Weak	NO
Sunny	Hot	High	Strong	NO
Overcast	Hot	High	Weak	Yes
Rain	Mild	High	weak	Yes
Rain	cool	Normal	weak	Yes
Rain	cool	Normal	Strong	NO
Overcast	cool	Normal	Strong	Yes
Sunny	Mild	High	weak	NO
Sunny	cool	Normal	weak	Yes
Rain	Mild	Normal	weak	Yes
Sunny	Mild	Normal	strong	Yes
Overcast	Mild	High	Strong	Yes
Overcast	Hot	Normal	weak	Yes
Rain	Mild	High	Strong	NO

Algorithm:

ID3 (Algorithm, 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 = +ve
- If all Examples are negative, Return the single-node tree Root, with label = -ve
- If attribute is empty, Return the single-node tree Root, with label = most common value of Target_attribute in Examples
- Otherwise Begin.

o A ← the attribute from Attributes the best* classifies Examples.
o The decision attribute for Root ← A.
o For each possible value, v_i of A.

→ Add a new tree branch below Root, corresponding to the test A = v_i
→ Let Examples v_i be the subset of Examples that have value v_i for A.
→ If Examples v_i is empty.

- Then below this new branch add a leaf node with label = most common value of Target_attributes in Examples.
- Else below this new branch add the subtree

ID3 (Examples v_i, Target_address, Attributes − {A})

END.
Return Root

PROGRAM:

```
import math
import csv
def majorclass(attributes, data, target):
    freq = {}
    index = attributes.index(target)
    for tuple in data:
        if tuple[index] in freq:
            freq[tuple[index]] += 1.
        else:
            freq[tuple[index]] = 1.
```

```
    max=0
    major=" "
    for key in freq.keys():
        if freq[key] > max:
            max= freq[key]
            major = key.
    return major

def entropy(attributes, data, targetAttr):
    freq={}.
    dataEntropy=0.0
    i=0
    for entry in data:
        if entry[i] in freq:
            freq[entry[i]]+=1.0
        else:
            freq[entry[i]]=1.0.
    for freq in freq.values():
        dataEntropy+=(-freq/len(data))*math.log(freq/len(data),2)
    return dataEntropy.

def info_gain(attributes, data, attr, targetAttr):
    freq={}
    subsetEntropy=0.0
    i= attributes.index(attr)
    for entry in data:
        if entry[i] in freq:
            freq[entry[i]]+=1.0
        else:
            freq[entry[i]]=1.0
    for val in freq.keys():
        valProb=freq[val]/sum(freq.values())
        dataSubset=[entry for entry in data if entry[i]==val]
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            subsetEntropy += valProb * entropy(attributes, dataSubset, targetAttr)
        return (entropy(attributes, data, targetAttr) - subsetEntropy)

def attr_choose(data, attributes, target):
        best = attributes[0]
        maxGain = 0
        for attr in attributes:
                newGain = info_gain(attributes, data, attr, target)
                if newGain > maxGain:
                        maxGain = newGain
                        best = attr.
        return best

def get_values(data, attributes, attr):
        index = attributes.index(attr)
        values = []
        for entry in data:
                if entry[index] not in values:
                        values.append(entry[index])
        return values.

def get_data(data, attributes, best, val):
        new_data = [[]]
        index = attributes.index(best).
        for entry in data:
                if entry[index] == val:
                        newEntry = []
                        for i in range(0, len(entry)):
                                if (i != index):
                                        newEntry.append(entry[i])
                        new_data.append(newEntry).
```

```
    new_data.remove([])
    return new_data

def build_tree(data, attributes, target):
    data = data[:]
    vals = [record[attributes.index(target)] for record in data]
    default = majorclass(attributes, data, target)
    if not data or len(attributes)-1 <= 0:
        return default
    elif vals.count(vals[0]) == len(vals):
        return vals[0]
    else:
        best = attr_choose(data, attributes, target)
        tree = {best:{}}
        for val in get_values(data, attributes, best):
            new_data = get_data(data, attributes, best, val)
            newAttr = attributes[:]
            newAttr.remove(best)
            subtree = build_tree(new_data, newAttr, target)
            tree[best][val] = subtree
    return tree

def execute_decision_tree():
    data = []
    with open("weather.csv") as tsv:
        for line in csv.reader(tsv):
            data.append(tuple(line))
    attributes = ['Outlook', 'Temparature', 'Humidity', 'Wind', 'PlayTennis']
    target = attributes[-1]
    acc = []
    training_set = [x for i, x in enumerate(data)]
    tree = build_tree(training_set, attributes, target)
```

```
Print(tree)
 results =[]
 test_set= [('Sunny', 'Hot', 'High', 'Weak')]
 for entry in test_set:
        tempDict = tree.copy()
        result =" "
        while (isinstance(tempDict, dict)):
              child = []
              nodeval= next(iter(tempDict))
              child = tempDict[next(iter(tempDict))].keys()
              tempDict = tempDict[next(iter(tempDict))]
              index= attributes.index(nodeval)
              value= entry[index]
              if value in tempDict.keys():
                    result = tempDict[value]
                    tempDict = tempDict[value]
              else:
                    result = "Null"
                    break.
        if result != "Null"
              results.append(result == entry[-1])
              Print(result)
if __name__ == "__main__":
      execute_decision_tree()
```

Output:

```
{'Wind': {'Wind': 'PlayTennis', 'Weak': {'Humidity': {'High': {'Temparature': {'Hot': {'Outlook': {'Sunny': 'No', 'Overcast': 'Yes'}}, 'Mild': {'Outlook': {'Rain': 'Yes', 'Sunny': 'No'}}}}, 'Normal': 'Yes'}}, 'Strong': {'Humidity': {'High': {'Outlook': {'Sunny': 'No', 'Overcast': 'Yes', 'Rain': 'No'}}, 'Normal': {'Outlook': {'Rain': 'No', 'Overcast': 'Yes', 'Sunny': 'Yes'}}}}}}
No
```