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Lab 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(Using .csv file)

```
In [3]: import math
          import csv
          def load_csv(filename):
            lines=csv.reader(open(filename,"r"));
             dataset = list(lines)
             headers = dataset.pop(0)
              return dataset, headers
          class Node:
              def __init__(self,attribute):
                   self.attribute=attribute
                   self.children=[]
                   self.answer="
          def subtables(data,col,delete):
               coldata=[row[col] for row in data]
               attr=list(set(coldata))
               counts=[0]*len(attr)
              r=len(data)
               c=len(data[0])
               for x in range(len(attr)):
                  for y in range(r):
                       if data[y][col]==attr[x]:
                            counts[x]+=1
                \begin{array}{ll} \mbox{for $x$ in $range(len(attr))$:} \\ \mbox{dic}[attr[x]] = [[0 \mbox{ for i in $range(c)$] for $j$ in $range(counts[x])$]} \end{array} 
                   pos=0
                   for y in range(r):
    if data[y][col]==attr[x]:
                               del data[y][col]
```

```
der data[y][col]
dic[attr[x]][pos]=data[y]
                pos+=1
    return attr,dic
def entropy(5):
    attr=list(set(S))
    if len(attr)==1:
       return 0
    counts=[0,0]
    for i in range(2):
       counts[i]=sum([1 for x in S if attr[i]==x])/(len(S)*1.0)
    for cnt in counts:
       sums+=-1*cnt*math.log(cnt,2)
    return sums
def compute_gain(data,col):
    attr,dic = subtables(data,col,delete=False)
    total_size=len(data)
    entropies=[0]*len(attr)
    ratio=[0]*len(attr)
    total_entropy=entropy([row[-1] for row in data])
    for x in range(len(attr)):
       ratio[x]=len(dic[attr[x]])/(total_size*1.0)
        entropies[x]=entropy([row[-1] for row in dic[attr[x]]])
        total_entropy-=ratio[x]*entropies[x]
    return total_entropy
def build_tree(data,features):
    lastcol=[row[-1] for row in data]
    if(len(set(lastcol)))==1:
       node=Node("")
       node.answer=lastcol[0]
        return node
    n=len(data[0])-1
    gains=[0]*n
   for col in range(n):
```

```
attr,dic=subtables(data,split,delete=True)
    for x in range(len(attr)):
       child=build_tree(dic[attr[x]],fea)
        node.children.append((attr[x],child))
    return node
def print_tree(node,level):
   if node.answer!="":
    print(" "*level,node.answer)
        return
    print(" "*level,node.attribute)
    for value,n in node.children:
    print(" "*(level+1),value)
        print_tree(n,level+2)
def classify(node,x_test,features):
   if node.answer!="":
        print(node.answer)
        return
    pos=features.index(node.attribute)
    for value, n in node.children:
       if x_test[pos]==value:
            classify(n,x_test,features)
'''Main program'''
dataset, features=load csv("id3.csv")
node1=build_tree(dataset,features)
print("The decision tree for the dataset using ID3 algorithm is")
print_tree(node1,0)
testdata,features=load_csv("id3.csv")
for xtest in testdata:
   print("The test instance:",xtest)
    print("The label for test instance:",end=" ")
    classify(node1,xtest,features)
```

The decision tree for the dataset using ID3 algorithm is $\ensuremath{\text{Outlook}}$

```
Outlook
   rain
     Wind
        strong
         no
       weak
         yes
   overcast
     yes
   sunny
     Humidity
       normal
        high
The test instance: ['sunny', 'hot', 'high', 'weak', 'no']
The label for test instance: no
The test instance: ['sunny', 'hot', 'high', 'strong', 'no']
The label for test instance: no
The test instance: ['overcast', 'hot', 'high', 'weak', 'yes']
The label for test instance: yes
The test instance: ['rain', 'mild', 'high', 'weak', 'yes']
The label for test instance: yes
The test instance: ['rain', 'cool', 'normal', 'weak', 'yes']
The label for test instance: yes
The test instance: ['rain', 'cool', 'normal', 'strong', 'no']
The label for test instance: no
The test instance: ['overcast', 'cool', 'normal', 'strong', 'yes']
The label for test instance: yes
The test instance: ['sunny', 'mild', 'high', 'weak', 'no']
The label for test instance: no
The test instance: ['sunny', 'cool', 'normal', 'weak', 'yes']
The label for test instance: yes
The test instance: ['rain', 'mild', 'normal', 'weak', 'yes']
The label for test instance: yes
The test instance: ['sunny', 'mild', 'normal', 'strong', 'yes']
The label for test instance:
The label for test instance: yes
The test instance: ['overcast', 'mild', 'high', 'strong', 'yes']
The label for test instance: yes
The test instance: ['overcast', 'hot', 'normal', 'weak', 'yes']
The label for test instance: yes
The test instance: ['rain', 'mild', 'high', 'strong', 'no']
The label for test instance: no
```