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1  """AI&ML
2  LAB-4:
3  Write a program to demonstrate the working of the decision tree based on ID3 algorithm.
4  Use an appropriate data set for building the decision tree and apply this knowledge to classify a new
   sample
5  """
6  import pandas as pd
7  from collections import Counter
8  import math
9
10 tennis=pd.read_csv('playtennis.csv')
11 print("\n Given play tennis data set:\n\n",tennis)
12
13 def entropy(alist):
14     c=Counter(x for x in alist)
15     instances=len(alist)
16     prob=[x/instances for x in c.values()]
17     return sum([-p*math.log(p,2) for p in prob])
18
19 def information_gain(d,split,target):
20     splitting=d.groupby(split)
21     n=len(d.index)
22     agent=splitting.agg({target:[entropy,lambda x:len(x)/n]})[target]
23     agent.columns=['entropy','observations']
24     new_entropy=sum(agent['entropy']*agent['observations'])
25     old_entropy=entropy(d[target])
26     return old_entropy-new_entropy
27 def id3(sub,target,a):
28     count=Counter(x for x in sub[target])
29     if len(count)==1:
30         return next(iter(count))
31     else:
32         gain=[information_gain(sub,attr,target)for attr in a]

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33     print("Gain=",gain)
34     maximum=gain.index(max(gain))
35     best=a[maximum]
36     print("Best Attribute:",best)
37     tree={best:{}}
38     remaining=[i for i in a if i!=best]
39     for val,subset in sub.groupby(best):
40         subtree=id3(subset,target,remaining)
41         tree[best][val]=subtree
42     return tree
43
44 names=list(tennis.columns)
45 print("List of Attributes:",names)
46 names.remove('playtennis')
47 print("Predicting Attributes:",names)
48 tree =id3(tennis,'playtennis',names)
49 print("\n\nThe Resultant Decision Tree is:\n")
50 print(tree)
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66 """Output:
67   Given play tennis data set:
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69   playtennis  outlook temperature humidity  wind
70 0           no    sunny          hot    high  weak
71 1           no    sunny          hot    high strong
72 2          yes overcast          hot    high  weak
73 3          yes    rain          mild    high  weak
74 4          yes    rain          cool normal  weak
75 5           no    rain          cool normal strong
76 6          yes overcast          cool normal strong
77 7           no    sunny          mild    high  weak
78 8          yes    sunny          cool normal  weak
79 9          yes    rain          mild normal  weak
80 10          yes    sunny          mild normal strong
81 11          yes overcast          mild    high strong
82 12          yes overcast          hot    normal  weak
83 13          no    rain          mild    high strong
84 list of Attributes: ['playtennis', 'outlook', 'temperature', 'humidity', 'wind']
85 Predicting Attributes: ['outlook', 'temperature', 'humidity', 'wind']
86 Gain= [0.2467498197744391, 0.029222565658954647, 0.15183550136234136, 0.04812703040826927]
87 Best Attribute: outlook
88 Gain= [0.01997309402197489, 0.01997309402197489, 0.9709505944546686]
89 Best Attribute: wind
90 Gain= [0.5709505944546686, 0.9709505944546686, 0.01997309402197489]
91 Best Attribute: humidity
92
93 The Resultant Decision Tree is:
94
95 {'outlook': {'overcast': 'yes', 'rain': {'wind': {'strong': 'no', 'weak': 'yes'}}}, 'sunny': {'humidity
   ': {'high': 'no', 'normal': 'yes'}}}}
96 """

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