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import pandas as pd
In [1]:
         import math
         df = pd.read csv('PlayTennis.csv')
         print("\n Input Data Set is:\n", df)
         t = df.keys()[-1]
         print('Target Attribute is: ', t)
         attribute names = list(df.keys())
         attribute names.remove(t)
         print('Predicting Attributes: ', attribute names)
         def entropy(probs):
             return sum( [-prob*math.log(prob, 2) for prob in probs])
         def entropy of list(ls,value):
             from collections import Counter
             cnt = Counter(x for x in ls)
             total instances = len(ls)
             probs = [x / total instances for x in cnt.values()]
             return entropy(probs)
         def information gain(df, split attribute, target_attribute,battr):
             df split = df.groupby(split attribute)
             alist=[]
             for gname, group in df split:
                 glist.append(gname)
             glist.reverse()
             nobs = len(df.index) * 1.0
             df aggl=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']
             new entropy = sum( df agg1['Entropy'] * df agg2['Proportion'])
             if battr !='S':
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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])
    if len(cnt) == 1:
        return next(iter(cnt))
    elif df.empty or (not attribute names):
        return default class
    else:
        default class = max(cnt.keys())
        gainz=[]
        for attr in attribute names:
            ig= information gain(df, attr, target attribute, default attr)
            gainz.append(ig)
        index of max = gainz.index(max(gainz))
        best attr = attribute names[index of max]
        tree = {best attr:{}}
        remaining attribute names =[i for i in attribute names if i != best attr]
        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)
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def classify(instance, tree,default=None):
             attribute = next(iter(tree))
             if instance[attribute] in tree[attribute].keys():
                 result = tree[attribute][instance[attribute]]
                 if isinstance(result, dict):
                     return classify(instance, result)
                 else:
                     return result
             else:
                 return default
         df new=pd.read csv('PlayTennisTest.csv')
         df new['predicted'] = df new.apply(classify, axis=1, args=(tree,'?'))
         print(df new)
         Input Data Set is:
              Outlook Temperature Humidity
                                              Wind PlayTennis
        0
               Sunny
                             Hot
                                     High
                                             Weak
                                                           No
        1
                             Hot
                                     High Strong
                                                           No
               Sunny
            0vercast
                             Hot
                                     High
                                             Weak
                                                          Yes
        3
                Rain
                            Mild
                                     High
                                             Weak
                                                          Yes
        4
                Rain
                            Cool
                                   Normal
                                             Weak
                                                          Yes
        5
                Rain
                            Cool
                                   Normal Strong
                                                          No
        6
            0vercast
                            Cool
                                   Normal Strong
                                                          Yes
        7
               Sunny
                            Mild
                                     High
                                                          No
                                             Weak
        8
               Sunny
                            Cool
                                   Normal
                                             Weak
                                                          Yes
        9
                Rain
                            Mild
                                   Normal
                                             Weak
                                                          Yes
        10
                                   Normal Strong
               Sunnv
                            Mild
                                                          Yes
        11 Overcast
                            Mild
                                     High Strong
                                                          Yes
        12 Overcast
                             Hot
                                   Normal
                                             Weak
                                                          Yes
                Rain
        13
                            Mild
                                     High Strong
                                                          No
        Target Attribute is: PlayTennis
        Predicting Attributes: ['Outlook', 'Temperature', 'Humidity', 'Wind']
        The Resultant Decision Tree is:
        {'Outlook': {'Overcast': 'Yes', 'Rain': {'Wind': {'Strong': 'No', 'Weak': 'Yes'}}, 'Sunny': {'Humidity': {'High': 'N
        o', 'Normal': 'Yes'}}}
          Outlook Temperature Humidity Wind PlayTennis predicted
            Sunny
                          Hot
                                  High
                                        Weak
                                                                No
                         Mild
                                        Weak
             Rain
                                  High
                                                               Yes
In [ ]:
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