8. ID3 (Decision Tree) Algorithm

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In [1]: from math import log
        from pprint import pprint
        import numpy as np
        import pandas as pd
/anaconda3/lib/python3.6/importlib/_bootstrap.py:219: RuntimeWarning: numpy.dtype size changed
  return f(*args, **kwds)
/anaconda3/lib/python3.6/importlib/_bootstrap.py:219: RuntimeWarning: numpy.dtype size changed
 return f(*args, **kwds)
In [2]: # Creating a Decision Tree class
        class DecisionTree(object):
            # Constructor to initialize the dataframe of class
            def __init__(self, df):
                self.df = df
                self.entropy = self.getentropy(df.iloc[:, -1])
            def getdata(self):
                print(self.df)
            def proportion(self, c, n):
                if c == 0:
                    return 0
                pc = c/n
                return -pc*log(pc, 2)
            # Compute the entroy of a given value vector
            def getentropy(self, values):
                n = len(values)
                value_dist = values.value_counts()
                entropy = 0
                for i in value_dist.index:
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entropy += self.proportion(value_dist[i], n)
    return entropy
# Compute the entropy for each attribute of dataset
def getgain(self, df):
    attribute = df.columns[0]
    info = {}
    avg = 0
   n = len(df.iloc[:, 0])
    value_dist = df.iloc[:, 0].value_counts()
    for i in value_dist.index:
        x = df.loc[df[attribute] == i]
        entropy = self.getentropy(x.iloc[:, -1])
        info[i] = entropy
        y = value_dist[i]/n * entropy
        avg += y
    info['average'] = avg
    return info
# Compute the info gain for each attribute of the dataset
def getinfogain(self):
   df = self.df
    infogain = {}
    target = df.columns[-1]
    for col in df.columns[:-1]:
        x = df.loc[:, [col, target]]
        info = self.getgain(x)
        infogain[col] = self.entropy - info['average']
    return infogain
# Find the root attribute having minimum infoquin
def findroot(self):
    ifgain = self.getinfogain()
    attr, maxig = None, 0
    for k in ifgain.keys():
        if ifgain[k] > maxig:
            maxig = ifgain[k]
            attr = k
    return attr
# Create the decision tree recursively
def createTree(self):
    df = self.df
    if self.entropy == 0:
        target = df.iloc[:, -1].value_counts().index[0]
        return target
    rootattr = self.findroot()
    tree = {}
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for i in df[rootattr].value_counts().index:
                   cols = list(df.columns)
                   cols.remove(rootattr)
                   x = df.loc[df[rootattr]==i, cols]
                   subtree = DecisionTree(x)
                   tree[i] = subtree.createTree()
               return {rootattr: tree}
In [3]: # Initialising the class with tennis data
       data = pd.read_csv('data.csv')
       dtree = DecisionTree(data)
       dtree.getdata()
    outlook temp humidity
                             windy play
              hot
0
      sunny
                      high
                              weak
1
      sunny
              hot
                      high strong
                                    no
2
   overcast
              hot
                      high
                              weak yes
3
      rainy mild
                      high
                              weak yes
4
      rainy cool
                   normal
                              weak yes
5
      rainy cool
                   normal strong
                                   no
6
   overcast cool
                   normal strong yes
7
      sunny mild
                      high
                              weak
8
      sunny cool normal
                              weak yes
9
      rainy mild normal
                              weak yes
      sunny mild normal strong yes
10
   overcast mild
11
                      high strong yes
12 overcast
            hot
                   normal
                              weak yes
13
      rainy mild
                      high strong
                                    no
In [4]: # Display the computed Desicion Tree
       pprint(dtree.createTree())
{'outlook': {'overcast': 'yes',
            'rainy': {'windy': {'strong': 'no', 'weak': 'yes'}},
            'sunny': {'humidity': {'high': 'no', 'normal': 'yes'}}}
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