```
In [5]:
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```
import pandas as pd
data=pd.read csv('/content/golfboll.csv')
print (data)
y=list(map(lambda v: 1 if v=='Yes' else 0 ,data['PlayGolf'].values ))
X = data[['Outlook', 'Temperature', 'Humidity', 'Windy', 'PlayGolf']].values
print(X)
y train = y[:10]
y \ val = y[10:]
X \text{ train} = X[:10]
X \text{ val} = X[10:]
class NaiveBayesClassifier:
    def init (self, X, y):
        self.X, self.y = X, y
        self.N = len(self.X) # Length of the training set
        self.dim = len(self.X[0]) # Dimension of the vector of features
        self.attrs = [[] for in range(self.dim)] # Here we'll store the columns of the
training set
        self.output dom = {} # Output classes with the number of ocurrences in the train
ing set. In this case we have only 2 classes
        self.data = [] # To store every row [Xi, yi]
        for i in range(len(self.X)):
            for j in range(self.dim):
                 # if we have never seen this value for this attr before,
                 # then we add it to the attrs array in the corresponding position
                if not self.X[i][j] in self.attrs[j]:
                    self.attrs[j].append(self.X[i][j])
            # if we have never seen this output class before,
            # then we add it to the output dom and count one occurrence for now
            if not self.y[i] in self.output dom.keys():
                self.output dom[self.y[i]] = 1
            # otherwise, we increment the occurrence of this output in the training set b
y 1
            else:
                self.output dom[self.y[i]] += 1
            self.data.append([self.X[i], self.y[i]])
        print(self.data)
    def classify(self, entry):
        solve = None # Final result
        \max \text{ arg} = -1 \# partial maximum}
        for y in self.output dom.keys():
            prob = self.output_dom[y]/self.N # P(y)
            for i in range(self.dim):
                cases = [x \text{ for } x \text{ in self.data if } x[0][i] == \text{ entry}[i] \text{ and } x[1] == y] # a
11 \text{ rows with } Xi = xi
                n = len(cases)
                prob *= n/self.N \# P *= P(Xi = xi)
            # if we have a greater prob for this output than the partial maximum...
            if prob > max arg:
                max arg = prob
                solve = y
        return solve
```

```
nbc = NaiveBayesClassifier(X train, y train)
total cases = len(y val)
good = 0
bad = 0
for i in range(total cases):
    predict = nbc.classify(X val[i])
    if y val[i] == predict:
       good += 1
    else:
       bad += 1
print('TOTAL EXAMPLES:', total cases)
print('RIGHT:', good)
print('WRONG:', bad)
print('ACCURACY:', good/total cases)
           Outlook Temperature Humidity Windy PlayGolf
    index
                                  High False
0
             Rainy
                           Hot
1
        1
                                   High
                                         True
                                                      No
             Rainy
                           Hot.
2
        2
                                   High False
          Overcast
                           Hot
                                                     Yes
3
        3
                                  High False
                          Mild
                                                     Yes
             Sunny
                                Normal False
4
       4
              Sunny
                          Cool
                                                     Yes
                                Normal
5
       5
                          Cool
                                          True
                                                     No
             Sunny
                                Normal
6
        6
                                          True
                                                     Yes
          Overcast
                          Cool
                          Mild High False
Cool Normal False
7
             Rainy
       7
                                                      No
8
       8
             Rainy
                                                     Yes
9
       9
             Sunny
                          Mild Normal False
                                                     Yes
10
      10
            Rainy
                          Mild Normal True
                                                     Yes
      11 Overcast
11
                          Mild
                                 High True
                                                     Yes
12
                          Hot Normal False
                                                     Yes
      12 Overcast
13
      13
             Sunny
                          Mild High True
                                                     No
[['Rainy' 'Hot' 'High' False 'No']
['Rainy' 'Hot' 'High' True 'No']
['Overcast' 'Hot' 'High' False 'Yes']
 ['Sunny' 'Mild' 'High' False 'Yes']
 ['Sunny' 'Cool' 'Normal' False 'Yes']
 ['Sunny' 'Cool' 'Normal' True 'No']
 ['Overcast' 'Cool' 'Normal' True 'Yes']
 ['Rainy' 'Mild' 'High' False 'No']
 ['Rainy' 'Cool' 'Normal' False 'Yes']
 ['Sunny' 'Mild' 'Normal' False 'Yes']
 ['Rainy' 'Mild' 'Normal' True 'Yes']
 ['Overcast' 'Mild' 'High' True 'Yes']
 ['Overcast' 'Hot' 'Normal' False 'Yes']
['Sunny' 'Mild' 'High' True 'No']]
[[array(['Rainy', 'Hot', 'High', False, 'No'], dtype=object), 0], [array(['Rainy', 'Hot',
'High', True, 'No'], dtype=object), 0], [array(['Overcast', 'Hot', 'High', False, 'Yes'],
dtype=object), 1], [array(['Sunny', 'Mild', 'High', False, 'Yes'], dtype=object), 1], [ar
ray(['Sunny', 'Cool', 'Normal', False, 'Yes'], dtype=object), 1], [array(['Sunny', 'Cool'
, 'Normal', True, 'No'], dtype=object), 0], [array(['Overcast', 'Cool', 'Normal', True, '
Yes'], dtype=object), 1], [array(['Rainy', 'Mild', 'High', False, 'No'], dtype=object), 0
], [array(['Rainy', 'Cool', 'Normal', False, 'Yes'], dtype=object), 1], [array(['Sunny',
'Mild', 'Normal', False, 'Yes'], dtype=object), 1]]
TOTAL EXAMPLES: 4
RIGHT: 4
WRONG: 0
```

ACCURACY: 1.0