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In [1]: import pandas as pd
         import numpy as np
         from math import log2
         import pprint
In [2]: | data = pd.DataFrame([
             ['Sunny', 'Hot', 'High', 'Weak', 'No'],
             ['Sunny', 'Hot', 'High', 'Strong', 'No'],
             ['Overcast', 'Hot', 'High', 'Weak', 'Yes'],
             ['Rain', 'Mild', 'High', 'Weak', 'Yes'],
             ['Rain', 'Cool', 'Normal', 'Strong', 'Yes'],
             ['Rain', 'Cool', 'Normal', 'Weak', 'Yes'],
             ['Overcast', 'Cool', 'Normal', 'Strong', 'Yes'],
             ['Sunny', 'Mild', 'High', 'Weak', 'No'],
             ['Sunny', 'Cool', 'Normal', 'Weak', 'Yes'],
             ['Rain', 'Mild', 'Normal', 'Weak', 'Yes'],
             ['Sunny', 'Mild', 'Normal', 'Strong', 'Yes'],
             ['Overcast', 'Mild', 'High', 'Strong', 'Yes'],
             ['Overcast', 'Hot', 'Normal', 'Weak', 'Yes'],
             ['Rain', 'Mild', 'High', 'Strong', 'No']
         ], columns=['Outlook', 'Temperature', 'Humidity', 'Wind', 'PlayTennis'])
        def entropy(target_col):
In [3]:
            values, counts = np.unique(target_col, return_counts=True)
            entropy = sum([-p * log2(p) for p in counts / counts.sum()])
             return entropy
In [4]: def info_gain(data, split_attribute_name, target_name="PlayTennis"):
            total_entropy = entropy(data[target_name])
            vals, counts = np.unique(data[split_attribute_name], return_counts=True)
            weighted_entropy = sum(
                 (counts[i] / sum(counts)) *
                 entropy(data[data[split_attribute_name] == vals[i]][target_name])
                 for i in range(len(vals))
            information_gain = total_entropy - weighted_entropy
             return information_gain
In [5]: def ID3(data, original_data, features, target_attribute_name="PlayTennis", parent_node_class=None):
            if len(np.unique(data[target_attribute_name])) <= 1:</pre>
                 return np.unique(data[target_attribute_name])[0]
            elif len(data) == 0:
                 return np.unique(original_data[target_attribute_name])[np.argmax(
                     np.unique(original_data[target_attribute_name],            return_counts=True)[1])]
            elif len(features) == 0:
                 return parent_node_class
            else:
                 parent_node_class = np.unique(data[target_attribute_name])[np.argmax(
                     np.unique(data[target_attribute_name], return_counts=True)[1])]
                 item_values = [info_gain(data, feature, target_attribute_name) for feature in features]
                 best_feature_index = np.argmax(item_values)
                 best_feature = features[best_feature_index]
                 tree = {best_feature: {}}
                 for value in np.unique(data[best_feature]):
                     sub_data = data[data[best_feature] == value]
                     new_features = features[:best_feature_index] + features[best_feature_index + 1:]
                     subtree = ID3(sub_data, original_data, new_features, target_attribute_name, parent_node_class)
                     tree[best_feature][value] = subtree
                 return tree
        features = list(data.columns)
         features.remove("PlayTennis")
         tree = ID3(data, data, features)
In [7]: # Print tree
         import pprint
         pprint.pprint(tree)
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{'Humidity': {'High': {'Outlook': {'Overcast': 'Yes',
                                            'Rain': {'Wind': {'Strong': 'No',
                                                              'Weak': 'Yes'}},
                                           'Sunny': 'No'}},
                       'Normal': 'Yes'}}
In [8]: def predict(query, tree, default='Yes'):
            for attr in query:
                if attr in tree:
                    try:
                        result = tree[attr][query[attr]]
                    except:
                        return default
                    if isinstance(result, dict):
                        return predict(query, result)
                    else:
                        return result
            return default
        sample = {'Outlook': 'Sunny', 'Temperature': 'Cool', 'Humidity': 'High', 'Wind': 'Strong'}
In [9]:
        prediction = predict(sample, tree)
        print("\nPredicted Output for sample is:", prediction)
        Predicted Output for sample is: No
In [ ]:
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