Building a Decision Tree: Takeaways 🖻

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Syntax

• Using Python to calculate entropy:

```
def calc_entropy(column):
"""

Calculate entropy given a pandas series, list, or numpy array.
"""

counts = numpy.bincount(column)

probabilities = counts / len(column)

entropy = 0

for prob in probabilities:
    if prob > 0:
        entropy += prob * math.log(prob, 2)

return -entropy
```

• Using Python to calculate information gain:

```
def calc_information_gain(data, split_name, target_name):
    """

Calculate information gain given a data set, column to split on, and target.
    """

original_entropy = calc_entropy(data[target_name])

column = data[split_name]

median = column.median()

left_split = data[column <= median]

right_split = data[column > median]

to_subtract = 0

for subset in [left_split, right_split]:
    prob = (subset.shape[0] / data.shape[0])

    to_subtract += prob * calc_entropy(subset[target_name])

return original_entropy - to_subtract
```

- Finding the best column to split on:

```
def find_best_column(data, target_name, columns):
    """

Find the best column to split on given a data set, target variable, and list of columns.
information_gains = []
for col in columns:
    information_gain = calc_information_gain(data, col, "high_income")
    information_gains.append(information_gain)
highest_gain_index = information_gains.index(max(information_gains))
highest_gain = columns[highest_gain_index]
return highest_gain
```

• Applying a function to a data frame:

Concepts

- Pseudocode is a piece of plain-text outline of a piece of code explaining how the code works. Exploring the pseudocode is a good way to understand it before tying to code it.
- Pseudocode for the ID3 algorithm:

```
def id3(data, target, columns)
   1 Create a node for the tree
   2 If all values of the target attribute are 1, Return the node, with label = 1
   3 If all values of the target attribute are 0, Return the node, with label = 0
   4 Using information gain, find A, the column that splits the data best
   5 Find the median value in column A
   6 Split column A into values below or equal to the median (0), and values above the median (1)
   7 For each possible value (0 or 1), vi, of A,
   8 Add a new tree branch below Root that corresponds to rows of data where A = vi
   9 Let Examples(vi) be the subset of examples that have the value vi for A
   10 Below this new branch add the subtree id3(data[A==vi], target, columns)
   11 Return Root
```

- We can store the entire tree in a nested dictionary by representing the root node with a dictionary and branches with keys for the left and right node.
- Dictionary for a decision tree:

```
{
   "left":{
  "left":{
     "left":{
        "number":4,
        "label":0
     "column": "age",
     "median":22.5,
     "number":3,
     "right":{
        "number":5,
        "label":1
  },
  "column": "age",
  "median":25.0,
  "number":2,
  "right":{
     "number":6,
     "label":1
  }
   },
```

```
"column":"age",
"median":37.5,
"number":1,
"right":{
"left":{
  "left":{
     "number":9,
     "label":0
  },
  "column":"age",
  "median":47.5,
  "number":8,
  "right":{
     "number":10,
     "label":1
  }
},
"column": "age",
"median":55.0,
"number":7,
"right":{
  "number":11,
  "label":0
}
```

Resources

- Recursion
- ID3 Algorithm



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