

Machine Learning

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- Iterative process
- Take a bunch of x
- Compute a frontier to separate x
- Compare on which side x are with where they should be
- Propagate the error

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Overfitting

- If the learning process keeps going...
- No errors on seen data
- But what about other \mathbf{x} (unseen in the training phase)

Diff between Decision and Regression

- Decision = predict a label
- Regression = predict a value

Diff between Decision and Regression

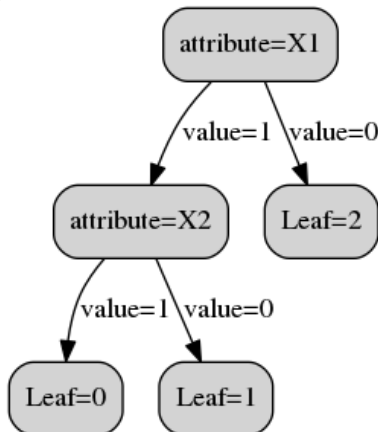
- Decision = predict a label
- Regression = predict a value

Regression Trees

- Similar to Decision Trees
- Problem is more complex

Decision Trees

- Frontier is based on decision rules
- Model has a tree hierarchy
- Each level is a test on attributes



How to choose the order of attributes?

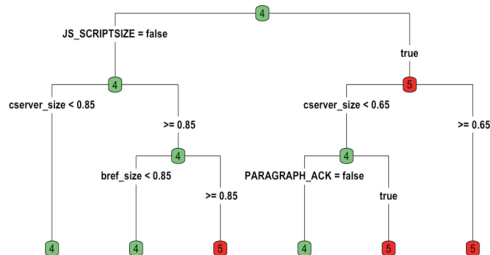
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Information gain

- Entropy based heuristic
- What is the attribute that discriminate most ?



Coming back at overfitting

- Few prediction errors on the training set
- Much more on the test set

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How to avoid it ?

- Constrain the growth of the tree
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Pruning

- based on heuristic
- Impurity measure
- Trade-off between the depth of the tree and number of errors

Input: attributes, examples, selection_heuristic, pruning_heuristic, strucutre_constraints

Output: tree : a decision tree

```
1: tree =  $\emptyset$ 
2: attribute = choose_best(attributes, examples, heuristic)
3: tree = create_root(attribute)
4: attributes = attributes - attribute
5: while attributes  $\neq \emptyset$  || strucutre_constraintsmet do
6:   attribute = choose_best(attributes, examples, heuristic)
7:   tree = add_child(attribute)
8:   attributes = attributes - attribute
9: end while
10: tree = prune(tree, pruning_heuristic)
```

Exploitation

- For each incoming new data
- Begin at the root of the tree
- Follow path depending on results of tests
- The ending leaf gives the class

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Decision rules

- New constraints can be extracted
- Simply follow the path !

