

#### Classification or Decision Tree

**CART** 

CLASSIFICATION AND REGRESSION TREE

#### **Training**

Customer ID	Gender	Responded to Email Marketing?
1	Male	Yes
2	Female	No

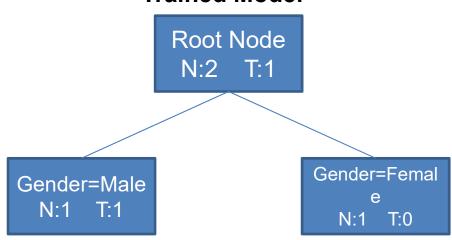
#### **Testing**

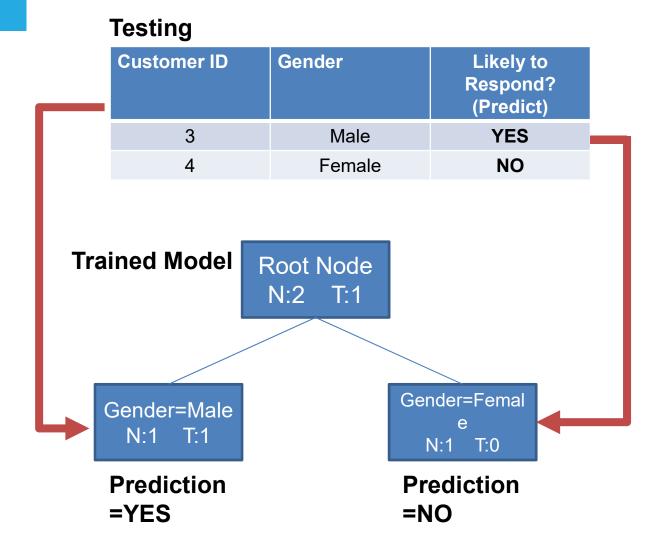
Customer ID	Gender	Likely to Respond? (Predict)
3	Male	?
4	Female	?

**Training** 

Customer ID	Gender	Responded to Email Marketing?
1	Male	Yes
2	Female	No

#### **Trained Model**

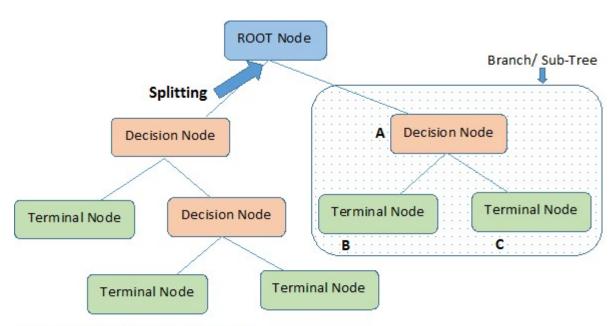






- What if there are more Independent Variables? –
  Choose the Best Variable using a splitting criterion
- What are the splitting criteria available? Gini
- How many splits to perform for a data? Depends on Purity of a Node
- Should we always perform a Binary Split? Binary & Multiway Split Models are available

# Decision Tree Terminology



Note:- A is parent node of B and C.

### Decision Trees greatlearning

- Supervised Machine Learning algorithm
- Mainly used for classification
- Works for both categorical and continuous output variables
- Terminology:
  - Root Node: Represents entire Training Dataset
  - Decision Node/Parent Node: Node that is split into sub-nodes
  - Splitting: Dividing Decision node into sub-nodes
  - Leaf/Terminal Node: Nodes that can no longer split
  - Branch: Sub-section of entire tree
  - Child node The resulting Nodes after splitting a decision node

### Decision Trees greatlearning

- Advantages
  - Easy to interpret
  - Automated field selection
  - No data processing required
    - Variable transformation not required
    - Can handle outliers
    - Missing value tolerant
- Disadvantages
  - They are unstable
  - Often inaccurate and poor compared to other models (Solution – Random Forest)
  - Generally not preferred for continuous prediction

## Decision Tree – Model Design

- Data should have both o (Bad) and 1 (Good) data
- Remove indeterminate values (NAs)
- Look for categorical variables
- Look for meaningful trend. Eg: Height should increase with Age.
- Look for default values like -999. Convert them to missing values. Maybe remove it.
- Look for capping/floor values Age > 100
- Reasons to create meaningful group Group all small Northeastern states

# Classification Techniques

#### Classification and Regression Tree (CART):

- Binary Decision Tree
- Classification (Categorical output variable)
- Regression (Continuous output variable)
- Uses Gini Index

#### CHAID – CHI-squared Automatic Interaction Detector:

- Non-Binary Decision Tree
- Use statistical significance of proportions

#### **LET US BUILD A DECISION TREE!!!**

### Gini Index Calculations

$$Gini(D) = 1 - \sum_{i=1}^{m} p_i^2$$

**m**: Number of Classes

**p**: Probability that a record in D belongs to class Ci

Gini Index consists of binary split ( $D_1$  and  $D_2$ ) for each attribute A

$$Gini_A(D) = \frac{D_1}{D}Gini(D_1) + \frac{D_2}{D}Gini(D_2)$$

**Reduction in Impurity** is given by:

$$\Delta Gini(A) = Gini(D) - Gini_A(D)$$

The attribute that Maximizes the reduction in impurity is chosen as the Splitting Attribute

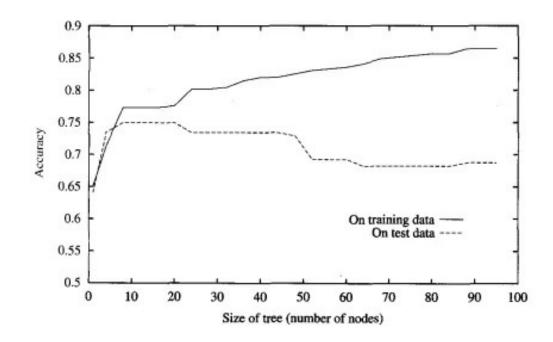
### Limitations of Decision Greatlearning

Vulnerable to over-fitting
 Solution – Pruning

Greedy Algorithm
 Solution – Cross Validation

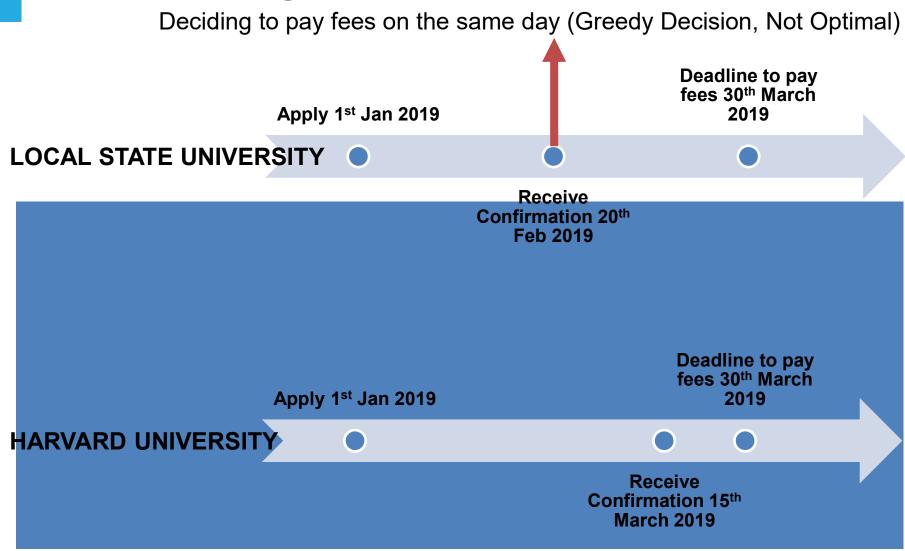
### Over Fitting **greatlearning**

- Works extremely well on Training dataset
- Performs poorly on unseen dataset



### Greedy Algorithm

#### greatlearning



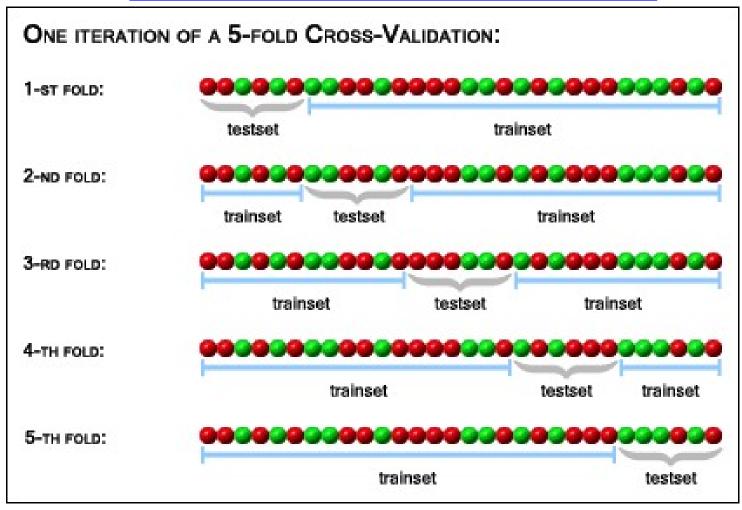
### Greedy Algorithm greatlearning

 When a Split happens using the best independent variable, the model does not consider the future states

 What-if the model has higher accuracy if a different independent variable is chosen instead of the best one?

### Cross Validation greatlearning

Source: https://stats.stackexchange.com/questions/1826/cross-validation-in-plain-english



### Cross-Validation greatlearning

- Helps overcome Greedy Algorithm problem
- How good is the model with unseen data?
- Also helps address 'Over Fitting'



#### Thank You