

## Specification and Estimation of CART

$$\hat{f}(x) = \sum_{m=1}^{|T|} I_{R_m}(x) \cdot \hat{y}_m$$

- $|T|$  is the number of leaves in the tree
- $R_m$  is the set of values of  $x$  in the  $m$ 'th leaf
- $I_{R_m}(x) = \begin{cases} 1 & \text{if } x \text{ is in region } R_m \\ 0 & \text{otherwise} \end{cases}$

•  $\hat{y}_m$  is the estimated function value for leaf  $m$ .

Temp. vs. O3 exam pk

### Parameters to Estimate:

Split Points: where do we make splits? Determines  $R_m$

Regression Constants: In each leaf, what is  $\hat{y}_m$ ?

### Optimization Target for Regression:

$$RSS = \sum_{i=1}^n (\hat{y}_i - y_i)^2 = \sum_{m=1}^{|T|} \sum_{i:x_i \in R_m} (\hat{y}_{im} - y_i)^2$$

### Optimization Target for Classification:

Often use Gini Index:

$$1 - \sum_{k=1}^K p_{km} \quad \text{where } p_{km} \text{ is proportion of obs. in region } m \text{ that are in class } k.$$

## Top down Estimation Algorithm:

1. Initialize tree with no splits
  - $\hat{y}$  is mean of all observations
  - calculate RSS for this "tree"
  
2. Repeat until a stopping criteria is met:
  - for every leaf, try every possible split at the midpoint of values of  $x$  in that leaf;
  - calculate RSS based on that split
  - select the split that gives the largest reduction in RSS
    - ⇒ reduction in RSS

Possible stopping criteria:

- all leaves have 5 or fewer obs.  
or some other #
- a maximum # of leaves has been reached
- a max. depth has been reached
- No reduction in RSS larger than  $\lambda$  can be achieved.

Regularization / Penalization:

$$\text{minimize } \text{RSS} + \lambda |T| \quad |T| \text{ # of leaves}$$

R package minimizes

$$\approx -R^2 + \lambda |T| = -\left(1 - \frac{\text{RSS}}{\text{TSS}}\right) + \lambda |T|$$

$$= \frac{\text{RSS}}{\text{TSS}} - 1 + \lambda |T|$$