### Classification and Regression Trees

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March 1, 2017

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**Overview - Regression Tree** 

Regression Tree

Medical Applications of CART

# Overview - Regression Tree

### • Regression Tree:

- Given the dataset  $\mathcal{D} = (x_1, y_1), ..., (x_n, y_n)$  where  $x_i \in \mathbb{R}, y_i \in \mathcal{Y} = \mathbb{R}$
- minimize the error (e.g. square loss error) in each leaf
- the parameterized function is

$$\hat{f}(x) = \sum_{j=1}^{K} eta_j \cdot \mathbb{I}(x \in \mathcal{R}_j)$$

• Using squared loss, optimal parameters are:

$$\hat{\beta}_j = \frac{\sum_i y_i \cdot \mathbb{I}(x_i \in \mathcal{R}_j)}{\sum_i \mathbb{I}(x_i \in \mathcal{R}_j)}$$

which is sample mean.

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Methodology

### Assign the prediction for each leaf

For each leaf, we need to assign the prediction y which minimizes the loss for the regression problem.

• Regression Tree:

$$\hat{y} = \arg\min_{y \in \mathbb{R}} \sum_{i \in \mathcal{R}_k} (y - y_i)^2 = \frac{1}{\sum \mathbb{I}(x_i \in \mathcal{R}_k)} \cdot \sum \mathbb{I}(x_i \in \mathcal{R}_k) \cdot y_i$$

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Methodology

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# Growing the Tree: Overview

- Ideally, would like to find partition that achieves minimal risk: lowest mean-squared error for regression problem.
- Number of potential partitions is too large to search exhaustively.
- Greedy search heuristics for a good partition:
  - Start at root.
  - Determine the best feature and value to split.
  - Recurse on children of node.
  - Stop at some point (with heuristic pruning rules).

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Growth Heuristic for Regression Trees

## Algorithm for Regression Trees

- Start with  $\mathcal{R}_1 = \mathbb{R}^d$
- For each feature j = 1, ..., d, for each value  $v \in \mathbb{R}$  that we can split on:
  - Split the data set:

$$I_{<} = \{i : x_{ij} < v\}$$
 and  $I_{>} = \{i : x_{ij} \ge v\}$ 

• Estimate parameters:

$$eta_<=rac{\sum_{i\in I_<} y_i}{|I_<|} ext{ and } eta_>=rac{\sum_{i\in I_>} y_i}{|I_>|}$$

• Quality of split is measured by the squared loss:

$$\sum_{i \in I_{<}} (y_i - eta_{<})^2 + \sum_{i \in I_{>}} (y_i - eta_{>})^2$$

- Choose split with minimal loss.
- Recurse on both children, with  $(x_i, y_i)_{i \in I_{\leq}}$  and  $(x_i, y_i)_{i \in I_{>}}$ .

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#### **Example of Regression Trees**

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#### **Example of Regression Trees**

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