# Gradient Boosted Decision Trees Intro & Regression

## Boosting

Boosting – a method for combining outputs of many "weak" classifiers or regressors to produce a powerful "ensemble"

Variants of boosting: AdaBoost, BrownBoost, LogitBoost, Gradient Boosting

### Big Data:

- Large number of examples
- High dimensionality

One can train very complex (and accurate!) machine learning models by two approaches:

- 1. Start with a complex model from the very beginning and fit its parameters neural network
- 2. Build a model iteratively, each step requires training of simple models boosting

### Regression

Given a training set:  $Z = \{(x_1, y_1), ..., (x_n, y_n)\}$  $x_i$ - features,  $y_i$ - targets (real values)

Goal is to find f(x) using training set, such as

$$\min \sum_{(x, y) \in T} (f(x) - y)^2$$

at test set  $T = \{(x_1, y_1), ..., (x_n, y_n)\}$ 

How to build f(x)?

### Gradient Boosted Trees for Regression

$$f(\mathbf{x}) = \sum_{m=1}^{M} h_m(\mathbf{x})$$

 $h_m(x)$  - a decision tree

### Algorithm: Gradient Boosted Trees for Regression

Input: training set  $\{(x_1, y_1), \dots, (x_n, y_n)\}$ 

M – number of iterations

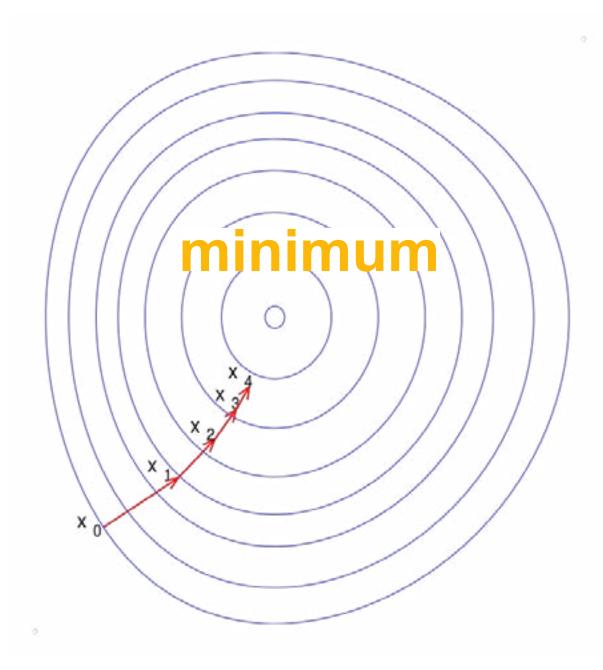
1. 
$$f_0(x) = \frac{1}{n} \sum_{i=1}^n y_i$$

- 2. For m=1...M:
- 3.  $\hat{y}_i = y_i f_{m-1}(x_i)$  (residual)
- 4. Fit a decision tree  $h_m(x)$  to the targets  $\hat{y}_i$  (auxiliary training set  $\{(x_1, \hat{y}_1), ..., (x_n, \hat{y}_n)\}$ )
- 5.  $f_m(x) = f_{m-1}(x) + vh_m(x)$
- 6. Return  $f_m(x)$

v - regularization (learningRate), recommended ≤ 0.1

# Optimization theory

Function maximization with gradient descent



$$f_{m}(x) = f_{0}(x) + vh_{1}(x) + vh_{2}(x) + ...$$

Boosting – minimization in the functional space

### Summary

- ▶ Boosting a method for combining outputs of many "weak" classifiers or regressors to produce a powerful "ensemble"
- ► Gradient Boosting a gradient descent minimization of the target function in the functional space
- ► Gradient Boosting with Decision Trees the best algorithm for general purpose classification or regression