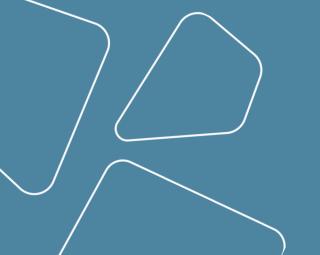
Machine Learning Lecture 6: Gradient boosting





Outline

- 1. Boosting intuitions
- 2. Gradient boosting
- 3.

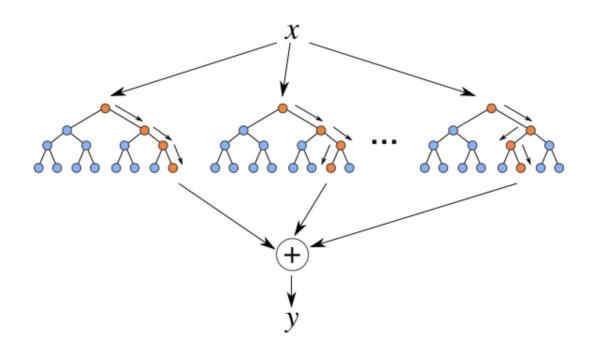


Based on: ml-mipt 2017 by Victor Kantor, Evgeny Sokolov HSE 2018, mlcourse_open by ODS

Random Forest



Bagging + RSM = Random Forest



Random Forest

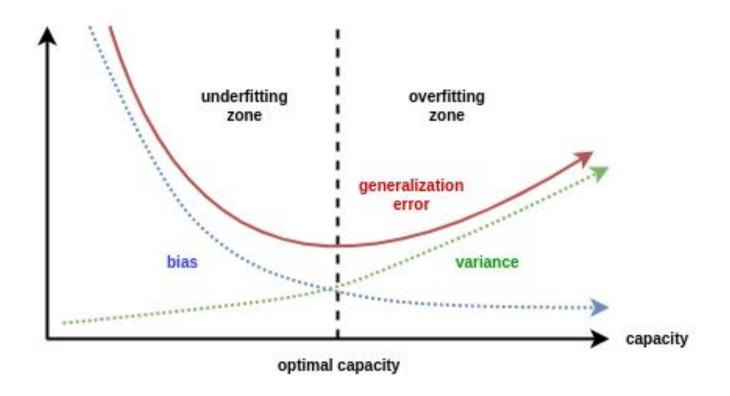


- One of the greatest "universal" models.
- There are some modifications: Extremely Randomized Trees, Isolation Forest, etc.
- Allows to use train data for validation: OOB

OOB =
$$\sum_{i=1}^{\ell} L\left(y_i, \frac{1}{\sum_{n=1}^{N} [x_i \notin X_n]} \sum_{n=1}^{N} [x_i \notin X_n] b_n(x_i)\right)$$

Bias-variance tradeoff

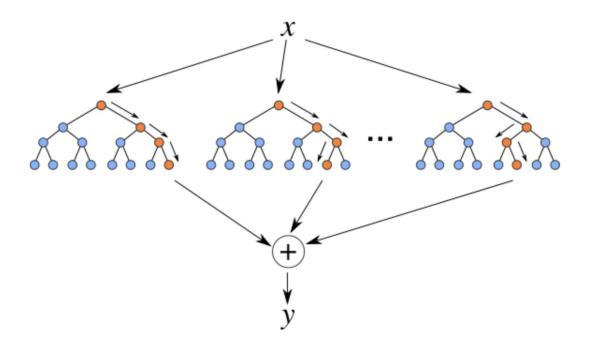




Random Forest



Is Random Forest decreasing bias or variance by building the trees ensemble?



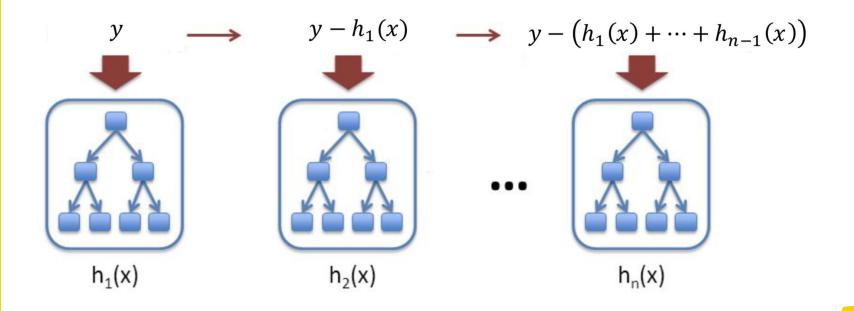
Boosting

girafe ai

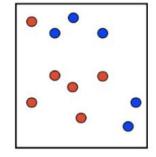
Boosting

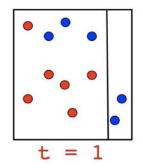


$$a_n(x) = h_1(x) + \dots + h_n(x)$$



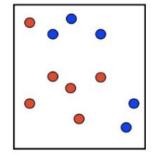
Binary classification

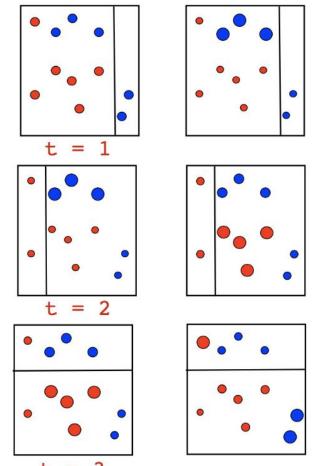






Binary classification

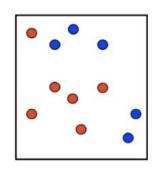






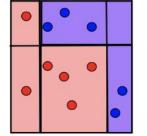


Binary classification

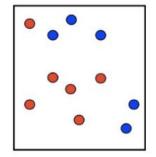


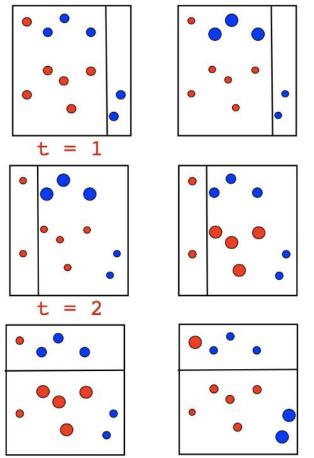
$$ho_1 -
ho_2 -
ho_3 -
ho_3$$

$$\hat{f}_T(x) = \sum_{t=1}^T \rho_t h_t(x) =$$



Binary classification







Gradient boosting

girafe ai



Denote dataset $\{(x_i,y_i)\}_{i=1,\dots,n}$, loss function L(y,f)

Optimal model:

$$\hat{f}(x) = \underset{f(x)}{\operatorname{arg\,min}} L(y, f(x)) = \underset{f(x)}{\operatorname{arg\,min}} \mathbb{E}_{x,y}[L(y, f(x))]$$

Let it be from parametric family:

$$\hat{f}(x) = f(x, \hat{\theta}),$$

$$\hat{\theta} = \arg\min \mathbb{E}_{x,y}[L(y, f(x, \theta))]$$



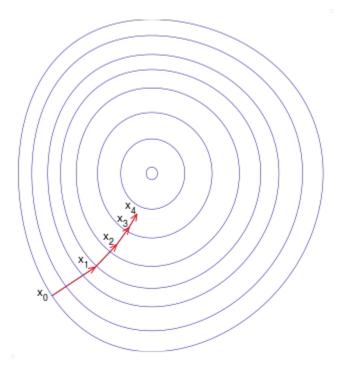
$$\hat{f}(x) = \sum_{i=0}^{t-1} \hat{f}_i(x),$$

$$(\rho_t, \theta_t) = \underset{\rho, \theta}{\operatorname{arg\,min}} \mathbb{E}_{x,y}[L(y, \hat{f}(x) + \rho \cdot h(x, \theta))],$$

$$\hat{f}_t(x) = \rho_t \cdot h(x, \theta_t)$$

What if we could use gradient descent in space of our models?





What if we could use gradient descent in space of our models?



$$\hat{f}(x) = \sum_{i=1}^{t-1} \hat{f}_i(x),$$

$$r_{it} = -\left[\frac{\partial L(y_i, f(x_i))}{\partial f(x_i)}\right]_{f(x) = \hat{f}(x)}, \quad \text{for } i = 1, \dots, n,$$

$$\theta_t = \underset{\theta}{\operatorname{arg\,min}} \sum_{i=1}^n (r_{it} - h(x_i, \theta))^2,$$

$$\rho_t = \underset{\rho}{\operatorname{arg\,min}} \sum_{i=1}^n L(y_i, \hat{f}(x_i) + \rho \cdot h(x_i, \theta_t))$$



In linear regression case with MSE loss:

$$r_{it} = -\left[\frac{\partial L(y_i, f(x_i))}{\partial f(x_i)}\right]_{f(x) = \hat{f}(x)} = -2(\hat{y}_i - y_i) \propto \hat{y}_i - y_i$$

Gradient boosting: beautiful demo



Great demo:

http://arogozhnikov.github.io/2016/06/24/gradient_boosting_explained.html

Gradient boosting



What we need:

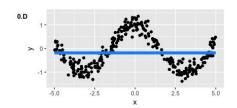
- Data.
- Loss function and its gradient.
- Family of algorithms (with constraints if necessary).
- Number of iterations M.
- Initial value (GBM by Friedman): constant.

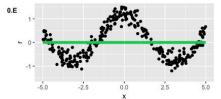


What we need:

- Data: toy dataset $y = cos(x) + \epsilon, \epsilon \sim \mathcal{N}(0, \frac{1}{5}), x \in [-5, 5]$
- Loss function: MSF
- Family of algorithms: decision trees with depth 2
- Number of iterations M = 3
- Initial value: just mean valu





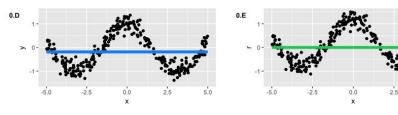


Left: full ensemble on each step.

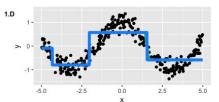
Right: additional tree decisions.

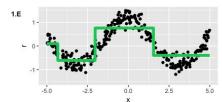
Example by ODS; source: https://habr.com/ru/company/od s/blog/327250/





Left: full ensemble on each step.

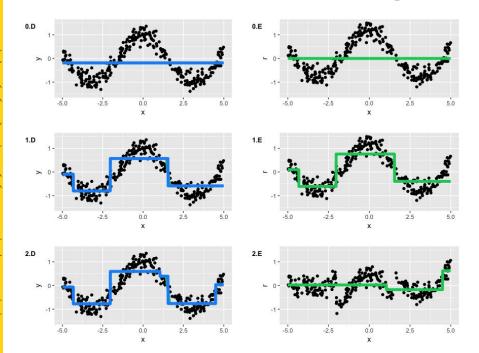




Right: additional tree decisions.

Example by ODS; source: https://habr.com/ru/company/ods/ s/blog/327250/



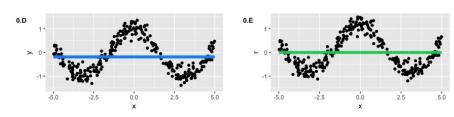


Left: full ensemble on each step.

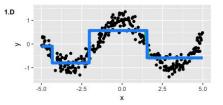
Right: additional tree decisions.

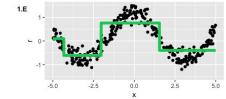
Example by ODS; source: https://habr.com/ru/company/ods/https://habr.com/ru/c



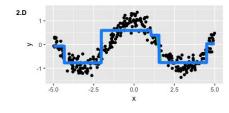


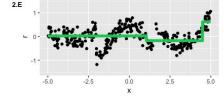
Left: full ensemble on each step.



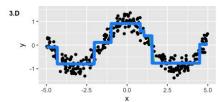


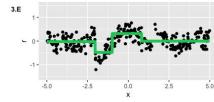
Right: additional tree decisions.





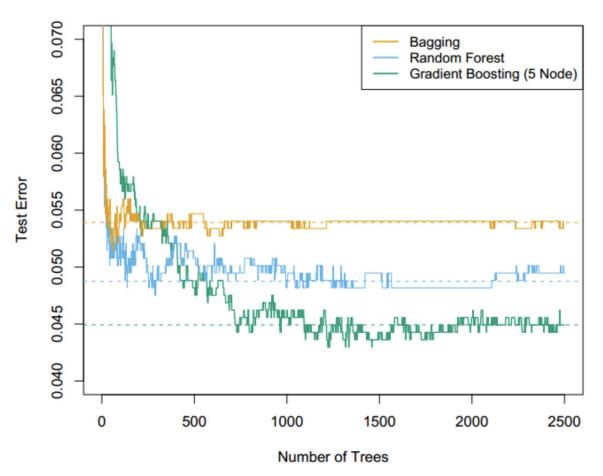






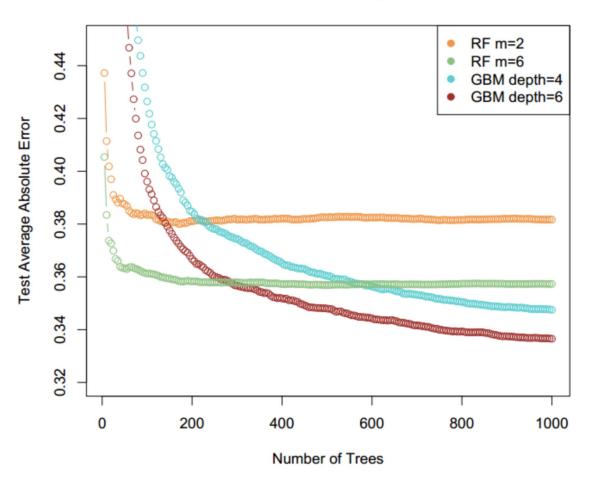
Spam Data





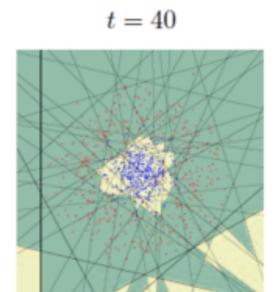
California Housing Data

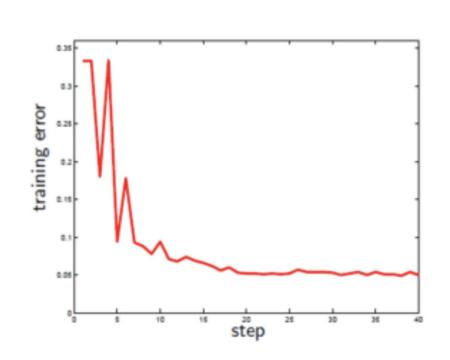




Boosting with linear classification methods







Technical side: training in parallel



Which of the ensembling methods could be parallelized?

- Random Forest: parallel on the forest level (all trees are independent)
- Gradient boosting: parallel on one tree level

Revise

- 1. Boosting intuitions
- 2. Gradient boosting



Thanks for attention!

Questions?



