

# XGBoost

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XGBoost (eXtreme Gradient Boost) is an algorithm which follows the principle of gradient boosting. The major difference with GBM is that XGBoost used a more regularized model formalization to control overfitting, which gives it better performance. It uses boosted trees for regression or classification problems.

**Input:**  $\{x_i, y_i\}_{i=1}^m$ , a loss function  $L(y, f(x))$ , max iterations = M, a learning rate  $\alpha$ .

1. Initialize the model with a constant value

$$f_0(x) = \operatorname{argmin}_{\theta} \left( \sum_{i=1}^m L(y_i, \theta) \right)$$

which we can calculate by  $\frac{df_0(x)}{d\theta} = 0$

2. Iterate k = 1 to M:

- a. Compute the gradients ( $g_k(x_i)$ ) and Hessians ( $h_k(x_i)$ )

for i = 1 to m:

$$\left| \begin{array}{l} g_k(x_i) = \left[ \frac{\partial L(y_i, f_{k-1}(x_i))}{\partial f_{k-1}(x_i)} \right] \\ h_k(x_i) = \left[ \frac{\partial^2 L(y_i, f_{k-1}(x_i))}{\partial f_{k-1}(x_i)^2} \right] \end{array} \right|$$

end

- b. Fit a base learner (tree)  $\phi_k(x)$  with input  $\left\{ x_i, -\frac{g_k(x_i)}{h_k(x_i)} \right\}_{i=1}^m$  by

solving:

$$\phi_k(x) = \operatorname{argmin}_{\phi} \left( \sum_{i=1}^m \frac{1}{2} h_k(x_i) \left[ -\frac{g_k(x_i)}{h_k(x_i)} - \phi(x_i) \right]^2 \right)$$

c. Update the model:

$$f_k(x) = f_{k-1}(x) + \alpha \phi_k(x)$$

3. Output function:  $f_M(x)$ .

**Questions:**

1. XGBoost is a/an:

- (a) Supervised learning algorithm
- (b) Unsupervised learning algorithm

**Ans. (a)**

2. XGBoost can be used for:

- (a) Regression
- (b) Classification
- (c) Both of these
- (d) None of these

**Ans. (c)**

3. Which of the following is an important assumption in XGBoost algorithm?

- (a) The value of every feature of data should be between 0 and 1.
- (b) Each sample model ( $\phi_k(x)$ ) is a stump.
- (c) Loss function can be approximated by a second order approximation.
- (d) None of the above

**Ans. (c)**

4. Why is regularization used in XGBoost?

**Ans.** Regularization prevents overfitting of the model by reducing the dependency of hypothesis function on one particular training data.

5. What are the advantages of using XGBoost?

**Ans.** It does not need normalized features and work with non-linear data also.

6. What are the disadvantages of using XGBoost?

**Ans.** It is sensitive to outliers. It does not perform so well on sparse and unstructured data.