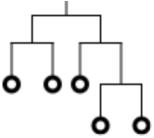
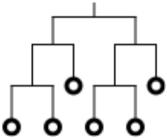
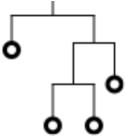


Gradient Boosting Machine (GBM)











$$T_M(\mathbf{x})$$



Algorithm 1 Forward Stagewise Additive Training 1: Initialize $f_0(\mathbf{x}) = 0$. 2: for i = 1 to M do

3: $\hat{\Theta}_i = \arg\min \sum_{i=1}^n L(y_i, f_{i-1}(x_i) + T(x_i; \Theta_i))$

5: end for

6: $f(\mathbf{x}) = f_M(\mathbf{x})$

4: $f_i(\mathbf{x}) = f_{i-1} + T(\mathbf{x}; \hat{\Theta}_i)$

GBM Algorithm

Algorithm 10.3 Gradient Tree Boosting Algorithm.

- 1. Initialize $f_0(x) = \arg\min_{\gamma} \sum_{i=1}^{N} L(y_i, \gamma)$.
- 2. For m=1 to M:
 - (a) For $i = 1, 2, \ldots, N$ compute

$$r_{im} = -\left[\frac{\partial L(y_i, f(x_i))}{\partial f(x_i)}\right]_{f=f_{m-1}}.$$

- (b) Fit a regression tree to the targets r_{im} giving terminal regions $R_{jm}, j = 1, 2, ..., J_m$.
- (c) For $j = 1, 2, \ldots, J_m$ compute

$$\gamma_{jm} = \arg\min_{\gamma} \sum_{x_i \in R_{jm}} L(y_i, f_{m-1}(x_i) + \gamma).$$

- (d) Update $f_m(x) = f_{m-1}(x) + \sum_{j=1}^{J_m} \gamma_{jm} I(x \in R_{jm})$.
- 3. Output $\hat{f}(x) = f_M(x)$.

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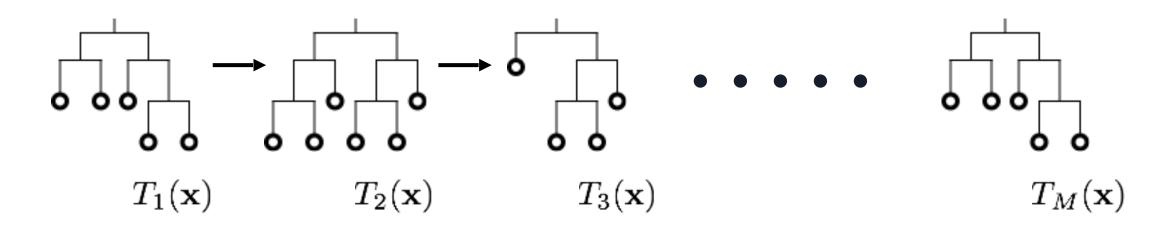
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Data Mining, Inference, and Prediction

Second Edition



Gradient Boosting Machine (GBM)



Algorithm 1 Forward Stagewise Additive Training

- 1: Initialize $f_0(\mathbf{x}) = 0$.
- 2: **for** i = 1 to M **do**

3:
$$\hat{\Theta}_i = \underset{\Theta_i}{\operatorname{arg\,min}} \sum_{i=1}^n L(y_i, f_{i-1}(x_i) + T(x_i; \Theta_i))$$

- 4: $f_i(\mathbf{x}) = f_{i-1} + T(\mathbf{x}; \hat{\Theta}_i)$
- 5: end for
- 6: $f(\mathbf{x}) = f_M(\mathbf{x})$

