## Algorithm

Input: training set  $\{(x_i, y_i)\}_{i=1}^n$  a differentiable loss function L(y, F(x)), number of iterations  $\underline{M}$ .

1. Initialize 
$$f_0(x) = \underset{i=1}{\underbrace{\operatorname{arg min}}} \sum_{i=1}^{N} L(y_i, \gamma).$$

$$\rightarrow 2$$
. For  $m = 1$  to  $\widehat{M}$ :

2. For 
$$\underline{m} = 1$$
 to  $\underline{M}$ :

(a) For  $i = 1, 2, ..., N$  compute

 $r_{im} = - \left[ \frac{\partial L}{\partial x_i} \right]$ 

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

- (b) Fit a regression tree to the targets  $r_{im}$  giving terminal regions  $R_{im}, j = 1, 2, \dots, J_m$
- (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 
$$\underline{f_m(x)} = \underline{f_{m-1}(x)} + \sum_{j=1}^{J_m} \gamma_{jm} I(x \in R_{jm}).$$

3. Output 
$$\hat{f}(x) = f_M(x)$$
.

$$\oint f(x) = f(x) + qT$$

$$f_2(x) = f_1(x) + dT^2$$

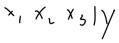
$$f_0(x) + dT$$

## Additive Modelling

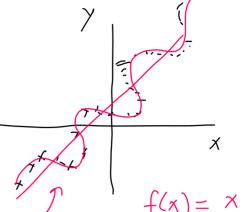
Monday, September 20, 2021 8:28 AM

$$\frac{1}{x} \xrightarrow{\chi} \rightarrow f()$$

$$\Rightarrow$$
  $y = f(x)$ 







additive

$$\lambda = X \quad \lambda = \sin(\chi)$$

## Explanation

$$F(x) = 0 \Rightarrow (y - 192) + (y - 194) + y$$

$$F(x) = \frac{f_0(x)}{y} + \frac{f_1(x)}{y} + \frac{f_2(x)}{y} + \cdots + \frac{f_m(x)}{y}$$

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$$F(x) = -\left[\frac{\partial L(y_i, f(x_i))}{\partial f(x_i)}\right] + \frac{\partial L(y_i, f(x_i))}{\partial f(x_i)} + \frac{\partial L(y_i, f($$

$$T_{11} = (Y_1 - Y_1) \int_{f=f_0}^{f=f_0} f \cdot (Y_1 - Y_1) \int_{f=f_0}^{f=f_0}$$

$$\frac{\gamma_{jm} = argmin}{\gamma} \leq L(\gamma_i, f_{m-1}(x_i) + \gamma)$$

$$\frac{\gamma_{jn}}{\gamma_{j1}} \Rightarrow \frac{\gamma_{i1}}{\gamma_{i1}} = \frac{argmin}{\gamma} \leq L(\gamma_i, f_{m-1}(\chi_i) + \gamma)$$

$$\frac{\chi_{ic}}{\gamma_{i1}} = \frac{\chi_{ic}}{\gamma_{i1}}$$

$$\gamma_{11} \quad \gamma_{21}$$

$$\gamma_{11} = \alpha \gamma_{1}^{min} \quad \frac{1}{2} (\gamma_{1} - (f_{0}(x_{1}) + \gamma)^{2})$$

$$\frac{dL}{d\gamma} = \frac{1}{2} \times \chi (\gamma_{1} - (f_{0}(x) + \gamma)) \frac{d}{d\gamma} (\frac{\gamma_{1}}{\gamma_{1}} - \frac{f_{0}(x) - \gamma}{d\gamma}) = 0$$

$$= -(\gamma_{1} - f_{0}(x) - \gamma) = 0$$

$$= \gamma_{1} - f_{0}(x) - \gamma = 0$$

$$\gamma_{11} = \alpha \gamma_{1}^{min} \quad \frac{1}{2} (\gamma_{1} - (f_{0}(x_{1}) + \gamma)^{2})$$

$$= -(\gamma_{1} - f_{0}(x) - \gamma) = 0$$

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$$\gamma_{11} = \alpha \gamma_{1}^{min} \quad \frac{1}{2} (\gamma_{1} - (f_{0}(x_{1}) + \gamma)^{2})$$

$$\gamma_{21} = \operatorname{argmin} \leq L(y_i, f_o(x_i) + \gamma)$$
 $\gamma_{1} \in R_{21}$ 

$$= \underset{\gamma}{\operatorname{arg min}} \frac{1}{2} \stackrel{\sim}{\leq} (y_i - (f_0(x_i) + \gamma)^2)$$

$$= -\frac{2}{\xi} \left( y_i - f_0(x_i) - \gamma \right) = 0$$

$$= \sum_{i=1}^{2} (\gamma_{i} - f_{0}(x_{i}) - \gamma) = 0$$

$$= y_1 - f_0(x_1) - \gamma + y_2 - f_0(x_2) - \gamma = 0$$

$$= 192 - 142 - \gamma + 144 - 142 - \gamma = 0$$

$$336 - 284$$
 $52 - 2\gamma = 0$ 
 $\gamma = \frac{52}{2}$