Boosting Algorithm

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1 Gradient Tree Boosting Algorithm

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Algorithm 1 Gradient Tree Boosting Algorithm
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1: procedure GBT(x, y)
         Initialize:
 2:
         f_0(x) = argmin_{\gamma} \sum_{i=1}^{N} L(y_i, \gamma)
         for m = 1 to M: do
 3:
              for i = 1 to N: compute do r_{im} = -\Big[\frac{\delta L(y_i, f(x_i))}{\delta f(x_i)}\Big]
 4:
              Fit a regression tree to the targets r_{im} giving terminal regions R_{jm}, j = 1,2,...,J_m
 6:
              for j = 1 to J_m: compute do
 7:
                   \gamma_{jm} = argmin_{\gamma} \sum_{x_i \in R_{jm}} L(y_i, f_{m-1}(x_i) + \gamma)
 8:
              Update:
 9:
         f_m(x) = f_{m-1}(x) + \sum_{j=1}^{J_m} \gamma_{jm} I(x \epsilon R_{jm})
         Output f(x) = f_M(x)
10:
```

2 Nestrov Descent

Starting with x_0 and y_0

$$x_{t+1} = y_t - w\nabla f(y_t) \tag{1}$$

$$y_{t+1} = (1 - \gamma_t)x_{t+1} + \gamma_t x_t \tag{2}$$

where w is the step size, λ_0 , $\lambda_t=\frac{1+\sqrt{1+4\lambda_{t-1}}^2}{2}$ and $\gamma_t=\frac{1-\lambda_t}{\lambda_{t+1}}$

3 Accelerated Gradient Boosting

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Algorithm 2 Accelerated Gradient Boosting Algorithm
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1: procedure AGB(x, y)
             Initialize:
 2:
             f_0(x) = argmin_{\gamma} \sum_{i=1}^{N} L(y_i, \gamma)
             for m = 1 to M: do
 3:
                    for i = 1 to N: compute do r_{im} = -\Big[\frac{\delta L(y_i, f(x_i))}{\delta f(x_i)}\Big]
 4:
 5:
                    Fit a regression tree to the targets r_{im} giving terminal regions R_{jm}, j = 1,2,...,J_m
 6:
                    for j = 1 to J_m: compute do
 7:
                           \gamma_{jm} = argmin_{\gamma} \sum_{x_i \in R_{jm}} L(y_i, f_{m-1}(x_i) + \gamma)
 8:
                    Update:
                   f_m(x) = g_{m-1}(x) + \sum_{j=1}^{J_m} \gamma_{jm} I(x \epsilon R_{jm})
g_m(x) = (1 - \gamma) f_m(x) + \gamma_t F_t
\lambda_t = \frac{1 + \sqrt{1 + 4\lambda_{t-1}}^2}{2} \text{ and } \lambda_{t+1} = \frac{1 + \sqrt{1 + 4\lambda_t}^2}{2}
and \gamma_t = \frac{1 - \lambda_t}{\lambda_{t+1}}
10:
11:
12:
13:
             Output \hat{f(x)} = f_M(x)
14:
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