

Compboost

Modular framework for component-wise boosting

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What is Component-Wise Boosting

Component-Wise Boosting: Terminology

- Loss Function:

$$L : \mathcal{Y} \times \mathcal{X} \rightarrow \mathbb{R}$$

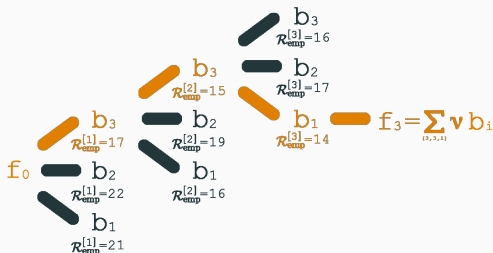
- Empirical Risk:

$$\mathcal{R}_{\text{emp}}(\theta) = \frac{1}{n} \sum_{i=1}^n L\left(y^{(i)}, f(x^{(i)})\right)$$

- Estimated model/parameter at iteration m :

$$\hat{f}^{[m]}, \theta^{[m]}$$

Component-Wise Boosting: The Idea



Iteration 1: $\hat{f}^{[1]}(x) = \beta b_3(x_3, \theta^{[1]})$

Iteration 2: $\hat{f}^{[2]}(x) = \beta b_3(x_3, \theta^{[1]}) + \beta b_3(x_3, \theta^{[2]})$

Iteration 2: $\hat{f}^{[3]}(x) = \beta b_3(x_3, \theta^{[1]}) + \beta b_3(x_3, \theta^{[2]}) + \beta b_1(x_1, \theta^{[3]})$

$$\Rightarrow \hat{f}^{[3]}(x) = \beta \left(b_3(x_3, \theta^{[1]} + \theta^{[2]}) + b_1(x_1, \theta^{[3]}) \right)$$

Component-Wise Boosting: The Algorithm

```
Result: Component-wise boosting model  $\hat{f}(x)$   
Initialize  $\hat{f}^{[0]}(x) = \arg \min_{c \in \mathbb{R}} \mathcal{R}_{\text{emp}}(c)$  ;  
for  $m \in \{1, \dots, M\}$  do  
    // Update pseudo residuals:  
     $r^{[m](i)} = - \left[ \frac{\delta}{\delta f(x^{(i)})} L \left( y^{(i)}, f(x^{(i)}) \right) \right]_{f=f^{[m-1]}} , \forall i \in \{1, \dots, n\}$  ;  
    // Get index  $j^*$  of  $m$ -th base-learner from optimizer:  
    for  $j \in \{1, \dots, J\}$  do  
        // Fit each base-learner  $b_j^{[m]}$  to the pseudo residuals:  
         $\hat{\theta}_j^{[m]} = \arg \min_{\theta_j} \sum_{i=1}^n \left( r^{[m](i)} - b_j^{[m]}(x^{(i)}, \theta_j) \right)^2$  ;  
        // Calculate the SSE of the fitted base-learner:  
         $\text{SSE}_j = \sum_{i=1}^n \left( r^{[m](i)} - b_j^{[m]}(x^{(i)}, \hat{\theta}_j) \right)^2$  ;  
    end  
    // Add selected component to model:  
     $\hat{f}^{[m]}(x) = \hat{f}^{[m-1]}(x) + \beta b_{j^*}^{[m]}(x, \theta_{j^*}^{[m]})$   
end  
Returns:  $\hat{f}(x) = \hat{f}^{[m]}(x)$ ;
```

- Tree-based implementations:
 - `xgboost`
 - `catboost`
 - `gbm`
- Model-based implementations:
 - `mboost`

So, why another boosting implementation?

About Comboost

About Comboost

Small Usecase

Next Steps

Questions?