

$$y_i \in \{0, 1\}, p_i = \frac{e^{pred_i}}{1 + e^{pred_i}} \quad (1)$$

$$\frac{\partial p}{\partial pred} = p(1 - p) \quad (2)$$

1 Initial

$$L(Y, pred) = - \sum (y_i * pred - \log(1 + e^{pred})) \quad (3)$$

$$g = \frac{\partial L}{\partial pred} = - \sum (y_i - p) \quad (4)$$

$$\arg \min_{pred} L(Y, pred) \Rightarrow \frac{\partial L}{\partial pred} = 0 \Rightarrow pred = \log \frac{1_{(yi=1)}}{1_{(yi=0)}} = \log \frac{count(yi=1)}{count(yi=0)} \quad (5)$$

2 Build Tree, Focus on the leaf , a.k.a terminal region

NOTE $pred_i$ is the begin predict state of every sample on the tree beginning, δ is the leaf value. REMEMBER: We have $pred_i$, we need to get δ .

$$Y = \{y_i\}, \arg \min_{\delta} L(Y, \{pred_i + \delta\}) \quad (6)$$

$$\Rightarrow \text{using the second taylor expansion at } pred_i, \text{ then } \frac{\partial L}{\partial \delta} = 0 \quad (7)$$

$$\Rightarrow \delta = -g/H, g_i = \frac{\partial L}{\partial pred_i} = y_i - p_i, H = \frac{\partial g_i}{\partial pred_i} = -\frac{\partial p_i}{\partial pred_i} = -p_i(1 - p_i) \quad (8)$$

$$\Rightarrow \text{final}, \delta = \frac{y_i - p_i}{p_i(1 - p_i)} \quad (9)$$

On predicting, use the following :

$$p_i = \frac{e^{pred_i}}{1 + e^{pred_i}}, \quad (10)$$

where $pred_i = pred_{t0} + \lambda(pred_{t1} + \dots + pred_{tn})$; t_i is tree $_i$, λ is learning-rate. (11)

3 Inspire

- Pick any leaf in the first tree, it maybe contains 10 or 100 train-samples, How many y-values does the leaf contains? How about the second tree?
- When we have 100 trees ,every tree have 8 leaf , How many values we can predict in permutation? (Predict-Sapce)