

τ_B : Branch nodes

τ_L : Leaf nodes

L_t : The optimal misclassification loss in each leaf node t ,

which is equal to the number of points in the node less the number of points of the most common label.

\hat{L} : Simply predicting the most popular class for the entire dataset.

d_t : $\mathbf{1}(\text{node } t \text{ applies a split})$ to track which branch nodes apply splits

$d_{p(t)}$: parents node

a_t : binary variables that sum to 1

α : complexity parameter

N_{kt} : to be the number of points of label k in node t

N_t : to be the total number of points in node t

N_{min} : the minimum leaf size

z_{it} : $\mathbf{1}(\text{xi is in node } t)$ to track the points assigned to each leaf node

l_t : $\mathbf{1}(\text{leaf } t \text{ contains any points})$ to enforce a minimum number of points at each leaf

n : number of training samples

Y_{ik} : an incorrect label prediction has cost 1, and a correct label prediction has cost 0, +1 if $y_i = k$, -1 otherwise

c_{kt} : binary variables to track the prediction of each node, where $c_{kt} = \mathbf{1}(c_t = k)$

ϵ_j : for feature j , the smallest non-zero distance between adjacent values of this feature

ϵ_{max} : maximum ϵ_j for all features