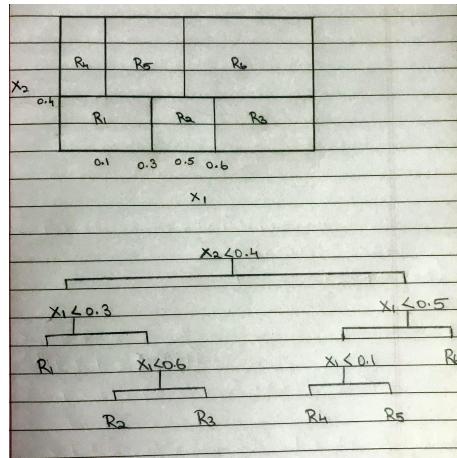
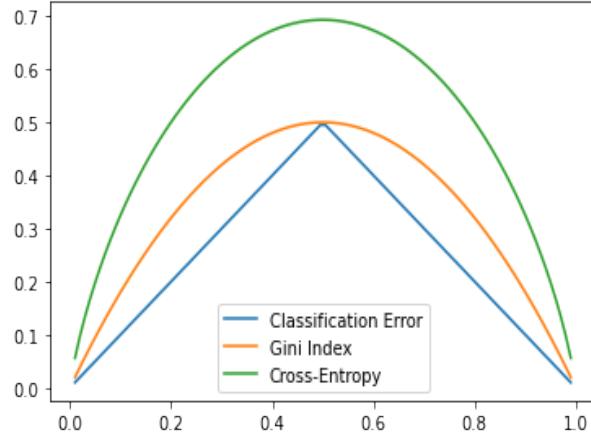


## Chapter 8 - Tree-Based Models

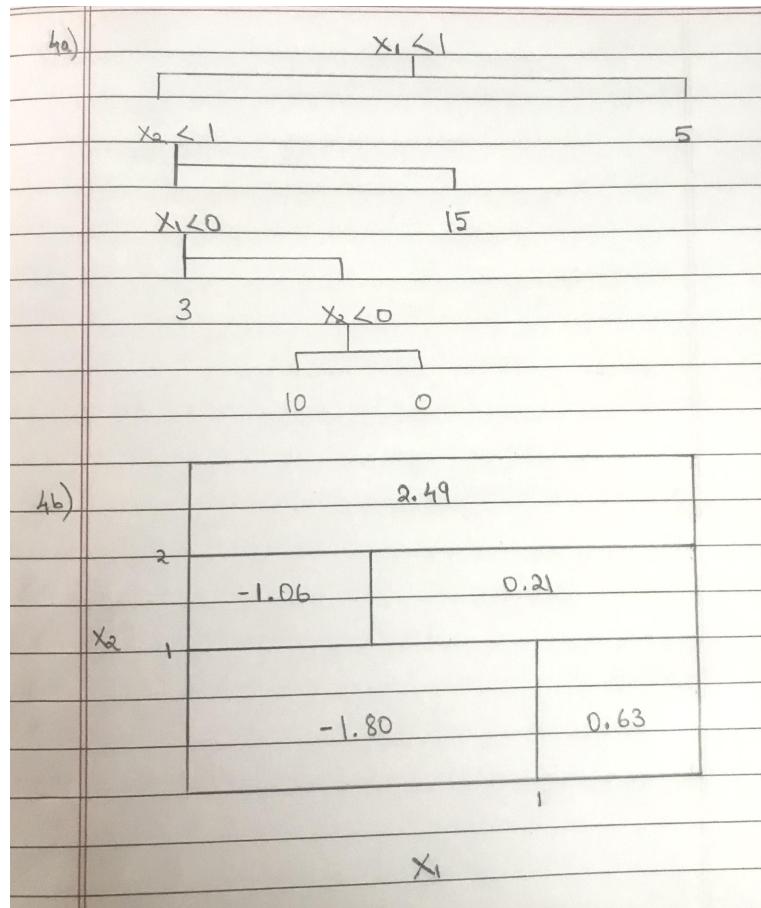
1. The below image contains 6 regions and 6 cut-points.



2.  $f(X) = \sum_{b=1}^B \lambda f^b(X)$ . When boosting is done using depth-one trees, then each tree only considers one variable. Thus,  $f^b(X_i) = \beta_0 + \beta_1 I(X_i < S)$ . Hence,  $f(X) = \beta_0 + \beta_1 I(X_1 < S_1) + \beta_2 I(X_2 < S_2) \dots$  which is additive in nature.
3. Here  $\hat{p}_{m1}$  is on the  $x$ -axis, and the errors are on the  $y$ -axis.



4. The below image contains solutions for 4a and 4b.



5.  $P(RED|X) = [0.1, 0.15, 0.2, 0.2, 0.55, 0.6, 0.6, 0.65, 0.7, 0.75]$ .  
 Average of  $P(RED|X) = 0.45$ , thus  $X$  does not belong to class  $RED$ .  
 Max Vote is for class  $RED$  (6 vs 4), thus  $X$  belongs to class  $RED$ .
6. Firstly a large tree is created using a greedy approach called recursive binary splitting to divide the predictor space into  $J$  regions by minimizing RSS given by  $\sum_{j=1}^J \sum_{i \in R_j} (y_i - \hat{y}_{R_j})^2$ . Now prune the tree to get a set of subtrees as a function of  $\alpha$ . Now using cross validation we can choose the best among the subtrees.