



Growing a Tree: Splitting Criteria

- If starting with criterion to minimize, then learning an optimal decision tree is an NP-complete problem
- heuristic algorithms employ greedy procedure
- splitting criteria:

Classification: k classes

Entropy:

Gini:

Squared error:

Regression

Squared error

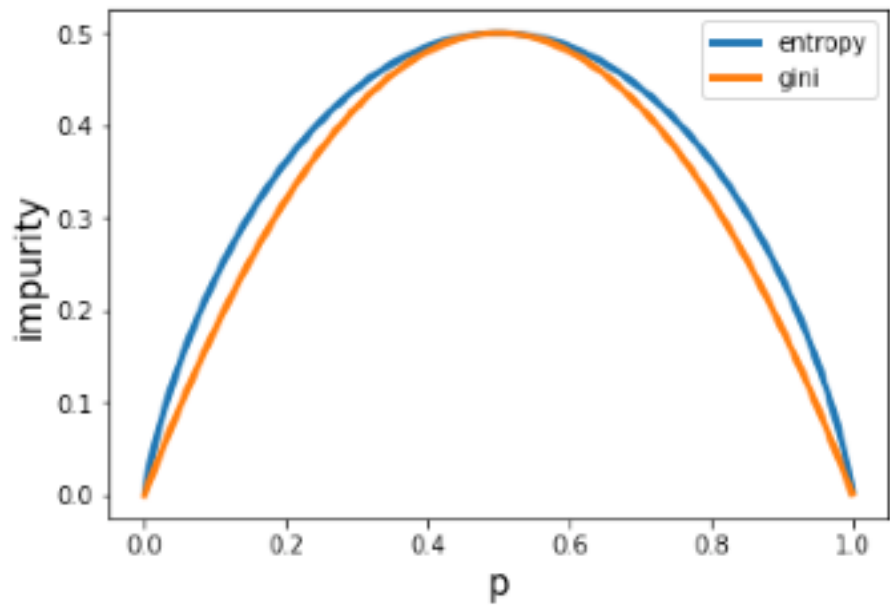
$$\sum_{i=1}^k \hat{p}_i(1-\hat{p}_i)$$

$$\sum_{i=1}^k \hat{p}_i \log(\hat{p}_i)$$

n

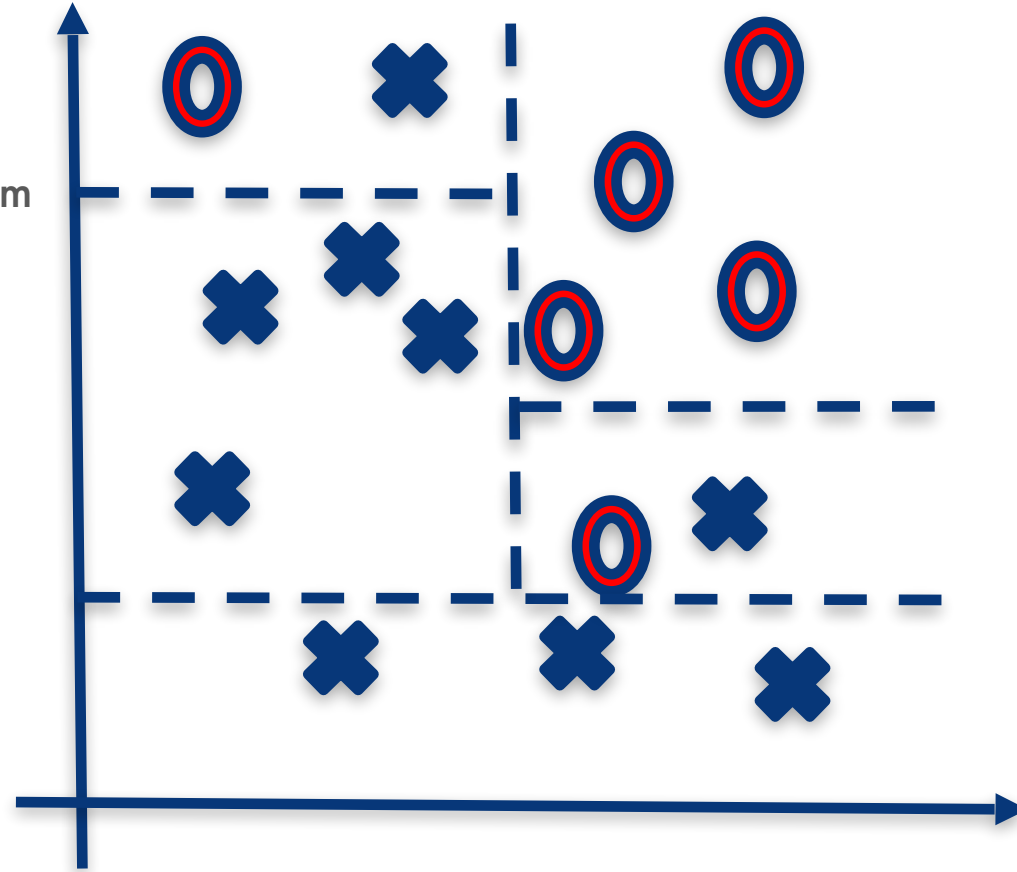
$$\sum_{i=1}^n (y_i - \hat{y})^2$$

$$\sum_{i=1}^n (y_i - \hat{y})^2$$



Growing a Tree: Stopping Criteria

- 1) All instances in leaf have the same y value
- 2) A maximum tree depth has been reached
- 3) Number of instances in leaf is below some minimum
- 4) Splitting criteria is below some threshold



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