

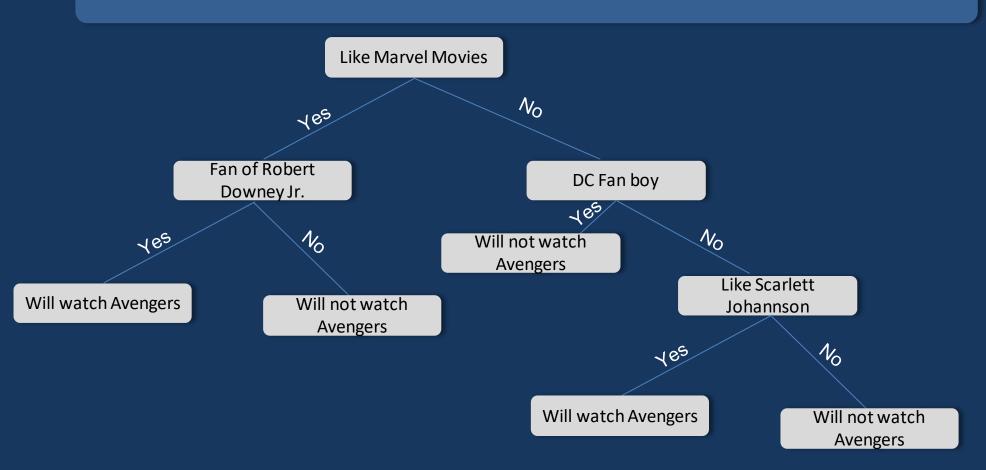
Classification with Tree Based Models



Decision Tree Algorithm



Decision Tree Algorithm is a supervised learning method used for both classification and regression



Decision Tree Algorithm-CART



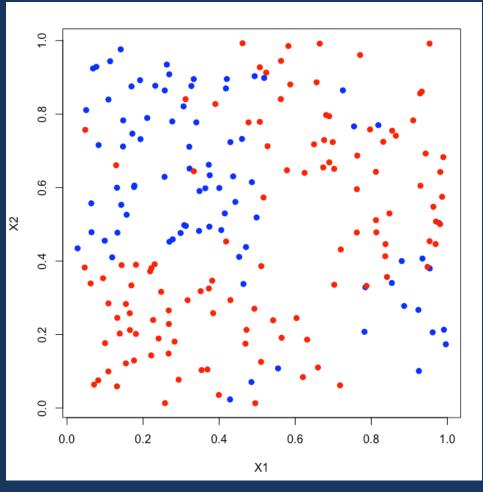
X1	X2	Υ	-	Categorical in Nature
0.2	0.3	Good		, reading
0.4	0.3	Bad		
0.2	0.1	Good		
0.6	0.5	Bad		
0.5	0.5	Good		

X1	X2	Υ	
(0.2	0.3	56
(0.4	0.3	34
(0.2	0.1	76
(0.6	0.5	12
(0.5	0.5	45

Numerical in Nature

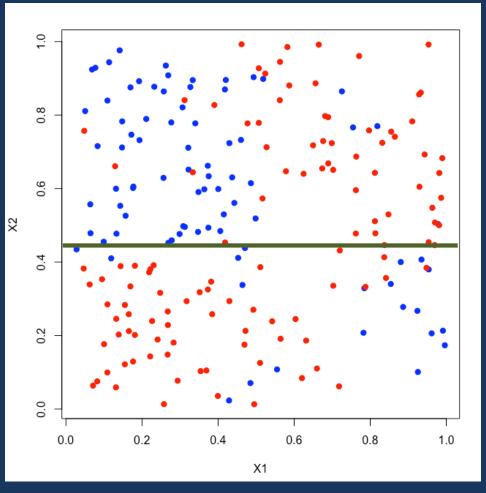
- Spam/Not Spam
- Tumor/No Tumor
- Lend Money/Deny

- Predict Stock Returns
- Predicting Sports Scores
- Pricing a house

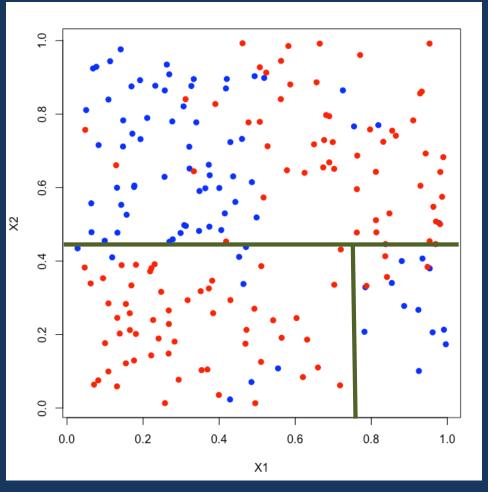


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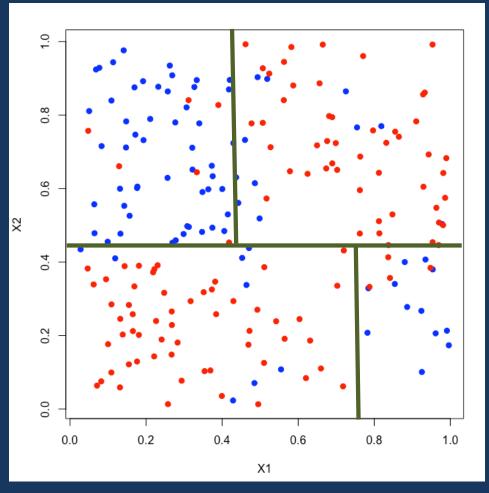




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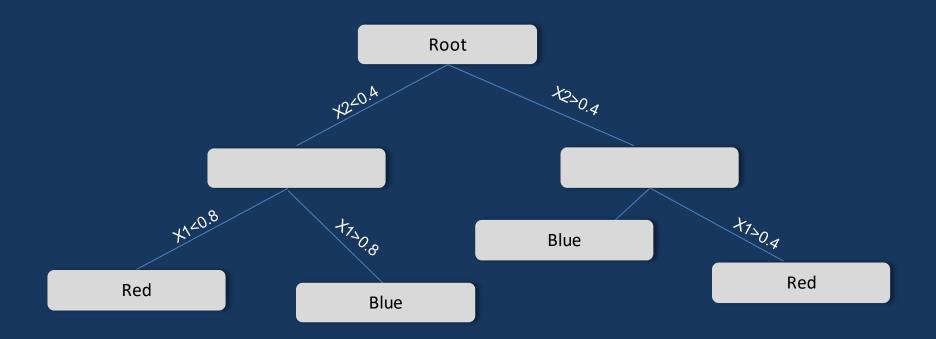
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How are decision trees built?





Measures of Impurity



These metrics measure how similar a region or a node is. They are said to measure the impurity of a region

Larger these impurity metrics the larger the "dissimilarity" of a nodes/regions

Gini Impurity

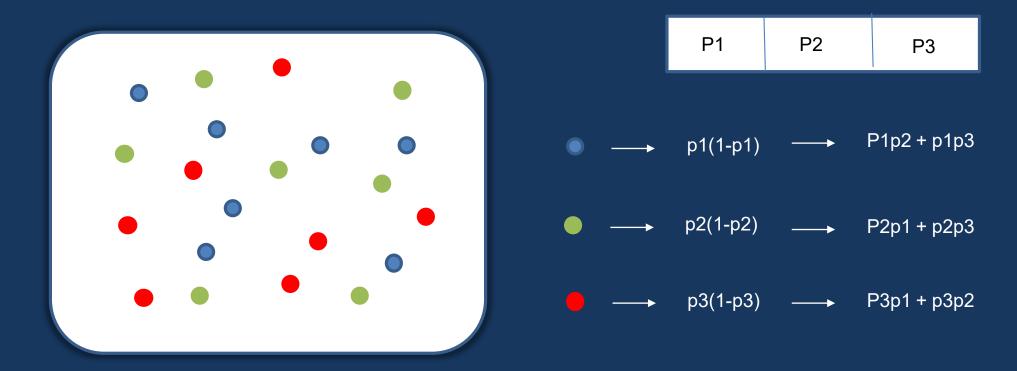
Entropy

Variance

Gini Impurity

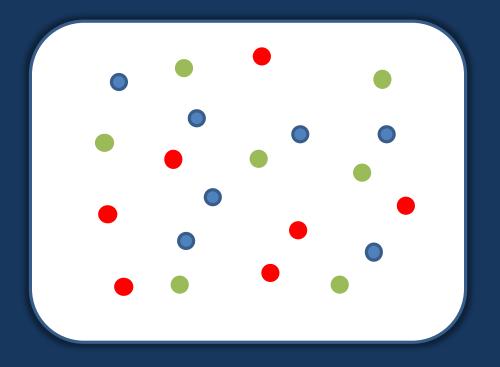


Gini Impurity is a measure of how often a randomly chosen element from the set would be incorrectly labeled if it was randomly labeled according to the distribution of labels in the subset



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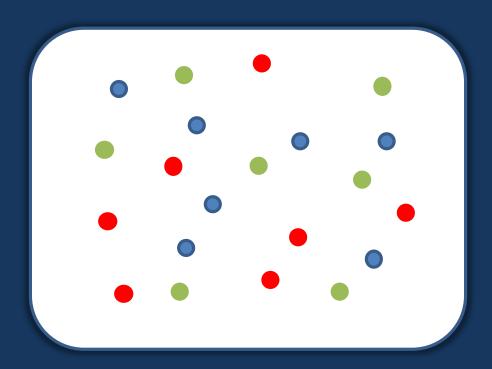


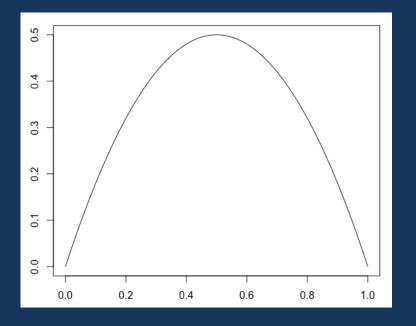
$$1 - \sum_{i=1}^J p_i^2$$

Gini Impurity

greatlearning Power Ahead

Gini Impurity is a measure of how often a randomly chosen element from the set would be incorrectly labeled if it was randomly labeled according to the distribution of labels in the subset





Tree to a Forest



Decision trees are very sensitive to even small changes in the data - usually called unstable

Can we get a whole bunch of decision trees to work together to yield a better and more robust prediction?

Then for prediction we could use the mean for regression trees and mode for classification trees

Bagging and Random Forest



