

Decision Trees

Calculating Node Impurity

- Calculate the Entropy, Gini Index, and Classification Error for each of the following nodes

Node N_1	Count
Class = 0	0
Class = 1	6

$$\text{Gini} = 1 - \left[\left(\frac{0}{6} \right)^2 + \left(\frac{6}{6} \right)^2 \right] = 0$$

$$\text{Entropy} = - \left[\left(\frac{0}{6} \right) \cdot \log_2 \left(\frac{0}{6} \right) + \left(\frac{6}{6} \right) \cdot \log_2 \left(\frac{6}{6} \right) \right] = 0$$

$$\text{Classification Error} = 1 - \max \left(\left(\frac{0}{6} \right), \left(\frac{6}{6} \right) \right) = 0$$

Node N_2	Count
Class = 0	1
Class = 1	5

$$\text{Gini} = 1 - \left[\left(\frac{1}{6} \right)^2 + \left(\frac{5}{6} \right)^2 \right] = 0.278$$

$$\text{Entropy} = - \left[\left(\frac{1}{6} \right) \cdot \log_2 \left(\frac{1}{6} \right) + \left(\frac{5}{6} \right) \cdot \log_2 \left(\frac{5}{6} \right) \right] = 0.650$$

$$\text{Classification Error} = 1 - \max \left(\left(\frac{1}{6} \right), \left(\frac{5}{6} \right) \right) = 0.167$$

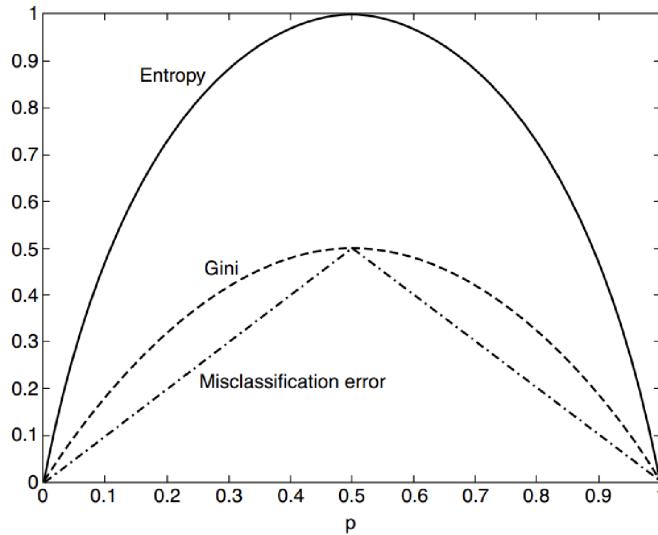
Node N_3	Count
Class = 0	3
Class = 1	3

$$\text{Gini} = 1 - \left[\left(\frac{3}{6} \right)^2 + \left(\frac{3}{6} \right)^2 \right] = 0.5$$

$$\text{Entropy} = - \left[\left(\frac{3}{6} \right) \cdot \log_2 \left(\frac{3}{6} \right) + \left(\frac{3}{6} \right) \cdot \log_2 \left(\frac{3}{6} \right) \right] = 1$$

$$\text{Classification Error} = 1 - \max \left(\left(\frac{3}{6} \right), \left(\frac{3}{6} \right) \right) = 0.5$$

Node Impurity



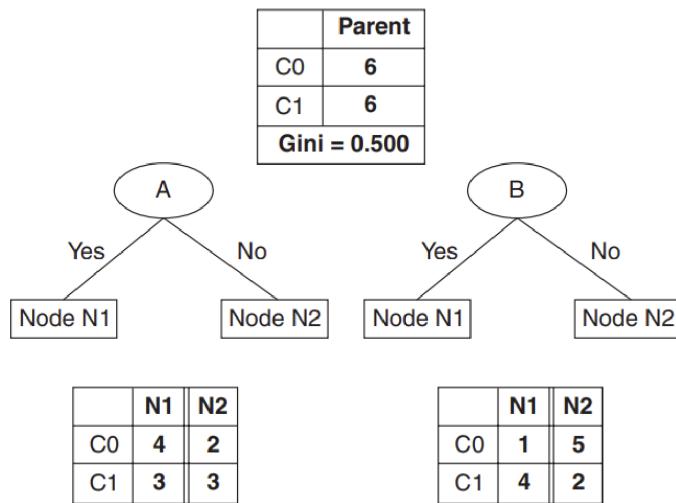
- Figure compares the values of the impurity measures for binary classification problems
 - p refers to the fraction of records that belong to one of the two classes
 - Observe that all three measures attain their maximum value when the class distribution is uniform (i.e., when $p = 0.5$)
 - The minimum values for the measures are attained when all the records belong to the same class (i.e., when $p = 0$ or $p = 1$)
 - To determine how well a test condition performs:
 - Compare the impurity of the parent node (before splitting) with the impurity of the child nodes (after splitting)
 - The larger their difference, the better the test condition
 - The gain, Δ , is a criterion that can be used to determine the goodness of a split:
- $$\Delta = I(\text{parent}) - \sum_{j=1}^k \frac{N(v_j)}{N} I(v_j)$$
- Where $I(\cdot)$ is the impurity measure of a given node, N is the total number of records at the parent node, k is the number of children, and $N(v_j)$ is the number of records associated with the child node, v_j

Measuring Node Impurity

- Decision tree induction algorithms often choose a test condition that maximizes the gain Δ
- Since $I(\text{parent})$ is the same for all test conditions, maximizing the gain is equivalent to minimizing the weighted average impurity measures of the child nodes
- When entropy is used as the impurity measure, the difference in entropy is known as the information gain, Δ_{info}

Splitting of Binary Attributes

- We will use Gini Index to select the best splitting of binary attributes
- A binary attribute splits into two partitions:
 - The parent node has 12 objects
 - There are two binary attributes, A and B , and their splitting results are shown in the figure below
- Which attribute (A or B) is better to split on?



Test	Node	Gini of Individual Children	Weighted Avg. Impurity of all Children
A	N1	$1 - \left(\left(\frac{4}{7}\right)^2 + \left(\frac{3}{7}\right)^2 \right) = 0.49$	$\left(\frac{7}{12} \cdot 49\right) + \left(\frac{5}{12} \cdot 48\right) = 0.486$
	N2	$1 - \left(\left(\frac{2}{5}\right)^2 + \left(\frac{3}{5}\right)^2 \right) = 0.32$	
B	N1	$1 - \left(\left(\frac{1}{5}\right)^2 + \left(\frac{4}{5}\right)^2 \right) = 0.32$	$\left(\frac{5}{12} \cdot 0.32\right) + \left(\frac{7}{12} \cdot 0.41\right) = 0.373$
	N2	$1 - \left(\left(\frac{5}{7}\right)^2 + \left(\frac{2}{7}\right)^2 \right) = 0.41$	

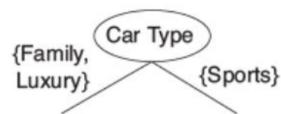
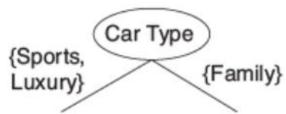
- Gain:

$$\Delta A = 0.5 - 0.486 = 0.014$$

$$\Delta B = 0.5 - 0.373 = 0.127$$

- We want to maximize the gain by minimizing the children's degree impurity

Splitting of Nominal Attributes



Car Type		
	{Sports, Luxury}	{Family}
C0	9	1
C1	7	3
Gini		

Car Type		
	{Sports}	{Family, Luxury}
C0	8	2
C1	0	10
Gini		

Car Type		
Family	Sports	Luxury
C0	1	8
C1	3	0
Gini		

(a) Binary split

(b) Multiway split

Test	Node	Gini of Individual Children	Weighted Avg. Impurity of all Children
A.1	Sport / Luxury Family	$1 - \left(\left(\frac{9}{16}\right)^2 + \left(\frac{7}{16}\right)^2 \right) = 0.492$	$\left(\frac{16}{20} \cdot 0.492\right) + \left(\frac{4}{20} \cdot 0.375\right)$
		$1 - \left(\left(\frac{1}{4}\right)^2 + \left(\frac{3}{4}\right)^2 \right) = 0.375$	$= 0.468$
A.2	Sport Family / Luxury	$1 - \left(\left(\frac{8}{8}\right)^2 + \left(\frac{0}{8}\right)^2 \right) = 0$	$\left(\frac{8}{20} \cdot 0\right) + \left(\frac{12}{20} \cdot 0.278\right)$
		$1 - \left(\left(\frac{2}{12}\right)^2 + \left(\frac{10}{12}\right)^2 \right) = 0.278$	$= 0.167$
B	Family Sports Luxury	$1 - \left(\left(\frac{1}{4}\right)^2 + \left(\frac{3}{4}\right)^2 \right) = 0.375$	$\left(\frac{4}{20} \cdot 0.375\right) + \left(\frac{8}{20} \cdot 0\right) +$
		$1 - \left(\left(\frac{8}{8}\right)^2 + \left(\frac{0}{8}\right)^2 \right) = 0$	$\left(\frac{8}{20} \cdot 0.219\right) = 0.163$
		$1 - \left(\left(\frac{1}{8}\right)^2 + \left(\frac{7}{8}\right)^2 \right) = 0.219$	