

Assignment 2.1

University of San Diego

ADS 502

Dingyi Duan

Introduction to Data Mining: Exercises 3.11 – Page 186: Question #3

3. Consider the training examples shown in Table 3.6 for a binary classification problem.

Table 3.6. Data set for Exercise 3.

Instance	a1	a2	a3	Target Class
1	T	T	1.0	+
2	T	T	6.0	+
3	T	F	5.0	-
4	F	F	4.0	+
5	F	T	7.0	-
6	F	T	3.0	-
7	F	F	8.0	-
8	T	F	7.0	+
9	F	T	5.0	-

a. What is the entropy of this collection of training examples with respect to the class attribute?

$$P(\text{positive}) = \frac{4}{9}.$$

$$P(\text{negative}) = \frac{5}{9}$$

Entropy for positive class

$$= - \left(\frac{4}{9} \log_2 \left(\frac{4}{9} \right) + \frac{5}{9} \log_2 \left(\frac{5}{9} \right) \right)$$

$$= \boxed{0.991}$$

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b. What are the information gains of a_1 and a_2 relative to these training examples?

a_1	+	-
T	3	1
F	1	4

$$\text{entropy for } a_1: \frac{4}{9} \left[-\frac{3}{4} \log_2 \left(\frac{3}{4} \right) - \frac{1}{4} \log_2 \left(\frac{1}{4} \right) \right] + \frac{5}{9} \left[-\frac{1}{5} \log_2 \left(\frac{1}{5} \right) - \frac{4}{5} \log_2 \left(\frac{4}{5} \right) \right]$$

$$= 0.762$$

$$\text{Information gain: } 0.991 - 0.762$$

$$= \boxed{0.229}$$

a_2	+	-
T	2	3
F	2	2

$$\text{entropy for } a_2: \frac{5}{9} \left[-\frac{2}{5} \log_2 \left(\frac{2}{5} \right) - \frac{3}{5} \log_2 \left(\frac{3}{5} \right) \right] + \frac{4}{9} \left[-\frac{2}{4} \log_2 \left(\frac{2}{4} \right) - \frac{2}{4} \log_2 \left(\frac{2}{4} \right) \right]$$

$$= 0.984$$

$$\text{Information gain: } 0.991 - 0.984$$

$$= \boxed{0.007}$$

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c. For a3, which is a continuous attribute, compute the information gain for every possible split.

a3:	1	3	4	5	6	7	8	
split pos.	0.5	2	3.5	4.5	5.5	6.5	7.5	8.5
	<=	>	<=	>	<=	>	<=	>
+	0	4	1	3	1	3	2	2
-	0	5	0	5	1	4	1	4

Split 1:

$$\text{Entropy} = - \left[\left(\frac{4}{9} \log_2 \left(\frac{4}{9} \right) + \frac{5}{9} \log_2 \left(\frac{5}{9} \right) \right) \right]$$

$$= 0.991$$

$$\text{Infor. gain} = 0.991 - 0.991 = \boxed{0}$$

Split 2:

$$<= \text{Entropy} = - [1 \cdot \log_2(1) + 0 \cdot \log_2(0)] = 0$$

$$> \text{Entropy} = - \left[\frac{3}{8} \cdot \log_2 \frac{3}{8} + \frac{5}{8} \cdot \log_2 \frac{5}{8} \right] = 0.954$$

~~Infor. gain~~

$$\text{Info. gain} = 0.991 - \left(\frac{1}{9} \cdot 0 + \frac{8}{9} \cdot 0.954 \right) = \boxed{0.143}$$

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Split 3:

$$\leq \text{Entropy} = -\left[\frac{1}{2} \cdot \log_2 \frac{1}{2} + \frac{1}{2} \cdot \log_2 \frac{1}{2}\right] = 1$$

$$> \text{Entropy} = -\left(\frac{3}{7} \cdot \log_2 \frac{3}{7} + \frac{4}{7} \cdot \log_2 \frac{4}{7}\right) = 0.985.$$

$$\text{Info gain} = 0.991 - \left(\frac{2}{9} \cdot 1 + \frac{7}{9} \cdot 0.985\right) = \boxed{0.00249}$$

Split 4:

$$\leq \text{Entropy} = -\left(\frac{2}{3} \cdot \log_2 \frac{2}{3} + \frac{1}{3} \cdot \log_2 \frac{1}{3}\right) = 0.918$$

$$> \text{Entropy} = -\left(\frac{2}{6} \cdot \log_2 \frac{2}{6} + \frac{4}{6} \cdot \log_2 \frac{4}{6}\right) = 0.918.$$

$$\text{Info gain} = 0.991 - \left(\frac{3}{9} \cdot 0.918 + \frac{6}{9} \cdot 0.918\right) = \boxed{0.0727}$$

Split 5:

$$\leq \text{Entropy} = -\left(\frac{2}{5} \cdot \log_2 \frac{2}{5} + \frac{3}{5} \cdot \log_2 \frac{3}{5}\right) = 0.971$$

$$> \text{Entropy} = -\left(\frac{2}{4} \cdot \log_2 \frac{2}{4} + \frac{2}{4} \cdot \log_2 \frac{2}{4}\right) = 1$$

$$\text{Info. gain} = 0.991 - \left(\frac{5}{9} \cdot 0.971 + \frac{4}{9} \cdot 1\right) = \boxed{0.00714}$$

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Split 6 :

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

$$\rightarrow \text{Entropy} = -\left(\frac{1}{3} \cdot \log_2 \frac{1}{3} + \frac{2}{3} \cdot \log_2 \frac{2}{3}\right) = 0.918.$$

$$\text{Info. gain} = 0.991 - \left(\frac{6}{9} \cdot 1 + \frac{3}{9} \cdot 0.918\right) = \boxed{0.6182}$$

Split 7 :

$$\Leftarrow \text{Entropy} = -\left(\frac{4}{8} \cdot \log_2 \frac{4}{8} + \frac{4}{8} \cdot \log_2 \frac{4}{8}\right) = 1$$

$$\rightarrow \text{Entropy} = -(0 \cdot \log_2 0 + 1 \cdot \log_2 1) = 0.$$

$$\text{Info. gain} = 0.991 - \left(\frac{8}{9} \cdot 1 + \frac{1}{9} \cdot 0\right) = \boxed{0.102}$$

Split 8 :

$$\Leftarrow \text{Entropy} = -\left(\frac{4}{9} \cdot \log_2 \frac{4}{9} + \frac{5}{9} \cdot \log_2 \frac{5}{9}\right) = 0.992.$$

$$\rightarrow \text{Entropy} = -(0 \cdot \log_2 0 + 0 \cdot \log_2 0) = 0.$$

$$\text{Info. gain} = \boxed{0.}$$

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d. What is the best split (among a_1 , a_2 and a_3) according to the information gain?

a_1 , due to its higher gain of 2.229

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e. What is the best split (between a_1 and a_2) according to the misclassification error rate?

Classification Error Rate:

$$a_1 = 1 - \left(\frac{7}{9}, \frac{2}{9} \right) = 1 - \frac{7}{9} = \frac{2}{9}. \quad \checkmark$$

$$a_2 = 1 - \left(\frac{5}{9}, \frac{4}{9} \right) = 1 - \frac{5}{9} = \frac{4}{9}.$$

a_1 is the better split for its lower MER
of $\frac{2}{9} \approx 0.222$

DP

f. What is the best split (between a_1 and a_2) according to the Gini index?

Gini Index :

$$a_1 = 1 - \left[\left(\frac{7}{9} \right)^2 + \left(\frac{2}{9} \right)^2 \right] = 0,346 . \quad \checkmark$$

$$a_2 = 1 - \left[\left(\frac{4}{9} \right)^2 + \left(\frac{5}{9} \right)^2 \right] = 0,494$$

$\boxed{a_1}$ is the best split for its lower GI.
of 0,346 .

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