

DSAA 5002 - Data Mining and Knowledge Discovery in Data Science

(Fall Semester 2023)

Homework 2 Solutions

1.

(a) (10 marks) Try threshold 2, 5, and 8 for attributes A (that is, use the “A > 2, A < 2”, “A > 5, A < 5”, and “A > 8, A < 8” respectively). Use the Gini score to determine the best one θ_a among them.

Value 2: 0.4018

Value 5: 0.4844

Value 8: 0.4872

The best choice is 2.

(b) (15 marks) Use θ_a obtained above, and the Gini score, determine which attributes should firstly be used for developing a decision tree.

I(Parent) = 0.4922

Gini(A) = 0.4018, Gain(A) = 0.0904

Gini(B) = 0.4844, Gain(B) = 0.0078

Gini(C) = 0.4643, Gain(C) = 0.0279

① for "A > 2, A < 2"

< 2	> 2
0	2
1	0

$$I(A < 2) = 1 - \left(\frac{2}{2}\right)^2 - 0 = 0$$

$$I(A > 2) = 1 - \left(\frac{2}{16}\right)^2 - \left(\frac{9}{16}\right)^2 = \frac{45}{96}$$

$$\text{Weighted Gini} = \frac{2}{16} \times 0 + \frac{14}{16} \times \frac{45}{96} = \frac{45}{112} \approx 0.4018$$

$$\Delta = \frac{63}{128} - \frac{45}{112} = \frac{81}{896} \approx 0.0904$$

② for "A > 5, A < 5"

< 5	> 5
0	3
1	4

$$I(A < 5) = 1 - \left(\frac{3}{8}\right)^2 - \left(\frac{5}{8}\right)^2 = \frac{15}{32}$$

$$I(A > 5) = 1 - \left(\frac{1}{2}\right)^2 - \left(\frac{1}{2}\right)^2 = \frac{1}{2}$$

$$\text{Weighted Gini} = \frac{1}{2} \times \frac{15}{32} + \frac{1}{2} \times \frac{1}{2} = \frac{31}{64} \approx 0.4844$$

$$\Delta = \frac{63}{128} - \frac{31}{64} = \frac{1}{128} \approx 0.0078$$

③ for "A > 8, A < 8"

< 8	> 8
0	6
1	7

$$I(A < 8) = 1 - \left(\frac{6}{13}\right)^2 - \left(\frac{7}{13}\right)^2 = \frac{84}{169}$$

$$I(A > 8) = 1 - \left(\frac{1}{3}\right)^2 - \left(\frac{2}{3}\right)^2 = \frac{4}{9}$$

$$\text{Weighted Gini} = \frac{13}{16} \times \frac{84}{169} + \frac{3}{16} \times \frac{4}{9} = \frac{19}{39} \approx 0.4872$$

$$\Delta = \frac{63}{128} - \frac{19}{39} = \frac{25}{4352} \approx 0.0058$$

$\therefore \text{Gini}_2 < \text{Gini}_5 < \text{Gini}_8$

\therefore The best threshold is 2: "A > 2, A < 2".

① for attribute "A". from part (a) we know that

$$\text{Weighted Gini} = \frac{2}{16} \text{Gini}(2,0) + \frac{14}{16} \text{Gini}(5,9) = \frac{45}{112}$$

② for attribute "B".

Yes	No
0	4
1	4

$$I(B = \text{Yes}) = 1 - \left(\frac{4}{8}\right)^2 - \left(\frac{4}{8}\right)^2 = \frac{1}{2}$$

$$I(B = \text{No}) = 1 - \left(\frac{3}{8}\right)^2 - \left(\frac{5}{8}\right)^2 = \frac{15}{32}$$

$$\text{Weighted Gini} = \frac{1}{2} \times \frac{1}{2} + \frac{1}{2} \times \frac{15}{32} = \frac{31}{64}$$

③ for attribute "C".

Yes	No
0	3
1	6

$$I(C = \text{Yes}) = 1 - \left(\frac{3}{7}\right)^2 - \left(\frac{4}{7}\right)^2 = \frac{4}{9}$$

$$I(C = \text{No}) = 1 - \left(\frac{4}{7}\right)^2 - \left(\frac{3}{7}\right)^2 = \frac{24}{49}$$

$$\text{Weighted Gini} = \frac{9}{16} \times \frac{4}{9} + \frac{7}{16} \times \frac{24}{49} = \frac{13}{28}$$

$\therefore I(\text{parent}) = 1 - \left(\frac{7}{16}\right)^2 - \left(\frac{9}{16}\right)^2 = \frac{63}{128}$

$\therefore \Delta_A = \frac{63}{128} - \frac{45}{112} = \frac{81}{896} \approx 0.0904$

$\Delta_B = \frac{63}{128} - \frac{31}{64} = \frac{1}{128} \approx 0.0078$

$\Delta_C = \frac{63}{128} - \frac{13}{28} = \frac{25}{896} \approx 0.0279$

$\therefore \Delta_A > \Delta_C > \Delta_B$

\therefore A should firstly be used for developing a decision tree

2.

(a) (10 marks) Consider the procedures of building a decision tree with Gini score. If we plan only to use the attributes a3 and a5 to predict the decision d2, which attribute should we use first?

	yes	no
$d_2 = \text{yes}$	9	1
$d_2 = \text{no}$	4	6

$$\text{Gini} = \frac{13}{20} \left[1 - \left(\frac{9}{13} \right)^2 - \left(\frac{4}{13} \right)^2 \right] + \frac{7}{20} \left[1 - \left(\frac{1}{7} \right)^2 - \left(\frac{6}{7} \right)^2 \right] = 0.3626.$$

If we use a5,

	yes	no
$d_2 = \text{yes}$	1	9
$d_2 = \text{no}$	3	7

$$\text{Gini} = \frac{4}{20} \left[1 - \left(\frac{1}{4} \right)^2 - \left(\frac{3}{4} \right)^2 \right] + \frac{16}{20} \left[1 - \left(\frac{9}{16} \right)^2 - \left(\frac{7}{16} \right)^2 \right] = 0.4688$$

Therefore, a3 should be used in the first step.

- (b) (20 marks) Use the naïve Bayes algorithm, the attributes a1 (with the threshold $\theta_1 = 37.95$), a2, and a3 only, to predict the decision d2 for the following data of a new patient. (For simplicity you do NOT need to use the Laplacian correction.)

$$P(d_2=\text{yes}) = 1/2$$

$$P(d_2=\text{no}) = 1/2$$

$$P(a_1 > 37.95 | d_2=\text{yes}) = 9/10$$

$$P(a_1 > 37.95 | d_2=\text{no}) = 1/10$$

$$P(a_2=\text{yes} | d_2=\text{yes}) = 0$$

$$P(a_2=\text{yes} | d_2=\text{no}) = 1/10$$

$$P(a_3=\text{no} | d_2=\text{yes}) = 1/10$$

$$P(a_3=\text{no} | d_2=\text{no}) = 6/10$$

Using the naïve Bayes assumption,

$$P(d_2=\text{yes} | a_1 > 37.95, a_2=\text{no}, a_3=\text{no}) = 0$$

$$P(d_2=\text{no} | a_1 > 37.95, a_2=\text{no}, a_3=\text{no}) = 0.003$$

3.

- (a) (7 marks) What is $P(\neg A, B, \neg C, D)$?

$$P(\neg A, B, \neg C, D) = P(\neg C | \neg A) P(D | \neg A, B) P(\neg A) P(B) = (1 - 0.2)(0.6)(1 - 0.1)(0.5) = 0.216$$

- (b) (8 marks) What is $P(A | B, C, D)$?

$$P(A | B, C, D) = P(A, B, C, D) / P(B, C, D) \quad P(A, B, C, D) = (0.7)(0.9)(0.1)(0.5) = 0.0315$$

$$P(B, C, D) = P(A, B, C, D) + P(\neg A, B, C, D) = 0.0315 + (0.2)(0.6)(0.9)(0.5) = 0.0855$$

$$\text{So, } P(A | B, C, D) = 0.0315 / 0.0855 = 0.368$$

4.

- (a) (5 marks)

$$y = \text{Relu} \left(\underset{1 \times 2}{V} \cdot \underset{2 \times 3}{\text{Relu} \left(\underset{3 \times 1}{W \cdot x} \right)} \right) \underset{\substack{\text{scalar} \\ \text{output}}}{1 \times 1}$$

set: $z = W \cdot x$
 $a = \text{Relu}(z)$
 $\therefore y = \text{Relu}(V \cdot a)$

(c) (10 marks)

forward pass $x = (1, 2, 1)^T$

$$z = W \cdot x = \begin{pmatrix} 1 & 0 & 1 \\ -3 & -1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \\ -5 \end{pmatrix}$$

$$a = \text{Relu}(W \cdot x) = \begin{pmatrix} 2 \\ 0 \end{pmatrix}$$

$$y = \text{Relu}(V \cdot a) = \text{Relu} \left[\begin{pmatrix} 0 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 0 \end{pmatrix} \right] = 0$$

(d) (15 marks) \hat{y} refers to o here

backward pass

$$J = \frac{1}{2} (y - \hat{y})^2$$

$$y = \text{Relu} \left(\underset{1 \times 2}{V} \cdot \underset{2 \times 1}{a} \right)$$

$$a = \text{Relu}(\underset{2 \times 1}{z}), \quad z = \underset{2 \times 1}{W} \cdot \underset{2 \times 3}{x}$$

$$\frac{\partial J}{\partial V} = \frac{\partial J}{\partial y} \cdot \frac{\partial y}{\partial V} = \frac{\partial}{\partial y} \left[\frac{1}{2} (y - \hat{y})^2 \right] \cdot \frac{\partial y}{\partial V}$$

$$= (y - \hat{y}) \text{Relu}'(V \cdot a) \cdot a^T$$

$$= (1 - 0) \text{Relu}'(0) \cdot \begin{pmatrix} 2 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 0 \end{pmatrix}$$