



جامعة الجلالة
GALALA UNIVERSITY

Artificial Intelligence Science Program

Chapter 4: Learning from Examples

Gain Ratio for Attribute Selection (C4.5)

- Information gain measure is **biased** towards attributes with a large number of values.
- C4.5 (a successor of ID3) **uses gain ratio** to overcome the problem (normalization to information gain).

$$SplitInfo_A(D) = -\sum_{j=1}^v \frac{|D_j|}{|D|} \times \log_2\left(\frac{|D_j|}{|D|}\right)$$

- $GainRatio(A) = Gain(A)/SplitInfo(A)$
- Ex.
$$SplitInfo_{income}(D) = -\frac{4}{14} \times \log_2\left(\frac{4}{14}\right) - \frac{6}{14} \times \log_2\left(\frac{6}{14}\right) - \frac{4}{14} \times \log_2\left(\frac{4}{14}\right) = 1.557.$$
 - $gain_ratio(income) = 0.029/1.557 = 0.019$
- The attribute with the **maximum gain ratio** is **selected** as the splitting **attribute**

$$Info(D) = - \sum_{i=1}^m p_i \log_2(p_i)$$

$$Info_A(D) = \sum_{j=1}^v \frac{|D_j|}{|D|} \times Info(D_j)$$

- Calculate **Entropy** of Class attribute:

$$SplitInfo_A(D) = - \sum_{j=1}^v \frac{|D_j|}{|D|} \times \log_2\left(\frac{|D_j|}{|D|}\right)$$

$$Gain(A) = Info(D) - Info_A(D)$$

$$GainRatio(A) = \frac{Gain(A)}{SplitInfo(A)}$$

buys_computer

yes	no
9	5

$$Info(D) = I(9,5) = -\frac{9}{14} \log_2\left(\frac{9}{14}\right) - \frac{5}{14} \log_2\left(\frac{5}{14}\right) = \underline{0.9403}$$

- Calculate **Gain Ratio** of all other attributes:

$$-\frac{2}{5} \log_2\left(\frac{2}{5}\right) - \frac{3}{5} \log_2\left(\frac{3}{5}\right)$$

		Class		
		yes	no	
age	youth	2	3	5
	middle_aged	4	0	4
	senior	3	2	5
				14

$$Info_{age}(D) = \frac{5}{14} I(2,3) + \frac{4}{14} I(4,0) + \frac{5}{14} I(3,2)$$

$$= \frac{5}{14} \cdot 0.971 + \frac{4}{14} \cdot 0 + \frac{5}{14} \cdot 0.971 = 0.3467 + 0 + 0.3467 = \underline{0.6934}$$

$$Gain(age) = Info(D) - Info_{age}(D) = 0.9403 - 0.6934 = \underline{0.2469}$$

$$SplitInfo_{age}(D) = -\frac{5}{14} \log_2\left(\frac{5}{14}\right) - \frac{4}{14} \log_2\left(\frac{4}{14}\right) - \frac{5}{14} \log_2\left(\frac{5}{14}\right) = \underline{1.5774}$$

$$GainRatio(age) = \frac{Gain(A)}{SplitInfo(A)} = \frac{0.246}{1.5774} = \underline{0.1559}$$

		Class		
		yes	no	
income	low	3	1	4
	medium	4	2	6
	high	2	2	4
				14

$$Info_{income}(D) = \frac{4}{14} I(3,1) + \frac{6}{14} I(4,2) + \frac{4}{14} I(2,2)$$

$$= \frac{4}{14} \cdot 0.8113 + \frac{6}{14} \cdot 0.9183 + \frac{4}{14} \cdot 1 = 0.2318 + 0.3935 + 0.2857 = \underline{0.911}$$

$$Gain(income) = 0.9403 - 0.911 = \underline{0.0293}$$

$$SplitInfo_{income}(D) = -\frac{4}{14} \log_2\left(\frac{4}{14}\right) - \frac{6}{14} \log_2\left(\frac{6}{14}\right) - \frac{4}{14} \log_2\left(\frac{4}{14}\right) = \underline{1.5566}$$

$$GainRatio(income) = \frac{0.0293}{1.5566} = \underline{0.0188}$$

age	income	student	credit_rating	buys_computer
youth	high	no	fair	no
youth	high	no	excellent	no
middle_aged	high	no	fair	yes
senior	medium	no	fair	yes
senior	low	yes	fair	yes
senior	low	yes	excellent	no
middle_aged	low	yes	excellent	yes
youth	medium	no	fair	no
youth	low	yes	fair	yes
senior	medium	yes	fair	yes
youth	medium	yes	excellent	yes
middle_aged	medium	no	excellent	yes
middle_aged	high	yes	fair	yes
senior	medium	no	excellent	no

$$Info(D) = -\sum_{i=1}^m p_i \log_2(p_i)$$

$$Info_A(D) = \sum_{j=1}^v \frac{|D_j|}{|D|} \times Info(D_j)$$

- Calculate **Entropy** of Class attribute:

buys_computer	
yes	no
9	5

$$Info(D) = I(9,5) = -\frac{9}{14} \log_2\left(\frac{9}{14}\right) - \frac{5}{14} \log_2\left(\frac{5}{14}\right) = \mathbf{0.9403}$$

$$SplitInfo_A(D) = -\sum_{j=1}^v \frac{|D_j|}{|D|} \log_2\left(\frac{|D_j|}{|D|}\right)$$

$$Gain(A) = Info(D) - Info_A(D)$$

$$GainRatio(A) = \frac{Gain(A)}{SplitInfo(A)}$$

- Calculate **Gain Ratio** of all other attributes:

		Class		
		yes	no	
student	yes	6	1	7
	no	3	4	7
				14

$$Info_{student}(D) = \frac{7}{14} I(6,1) + \frac{7}{14} I(3,4)$$

$$= \frac{7}{14} * 0.5917 + \frac{7}{14} * 0.9852 = 0.2958 + 0.4926 = \mathbf{0.7884}$$

$$Gain(student) = 0.9403 - 0.7884 = \mathbf{0.1519}$$

$$SplitInfo_{student}(D) = -\frac{7}{14} * \log_2\left(\frac{7}{14}\right) - \frac{7}{14} * \log_2\left(\frac{7}{14}\right) = 1$$

$$GainRatio(student) = \frac{0.1519}{1} = \mathbf{0.1519}$$

		Class		
		yes	no	
credit_rating	fair	6	2	8
	excellent	3	3	6
				14

$$Info_{credit_rating}(D) = \frac{8}{14} I(6,2) + \frac{6}{14} I(3,3)$$

$$= \frac{8}{14} * 0.8113 + \frac{6}{14} * 1 = 0.4636 + 0.4286 = \mathbf{0.8922}$$

$$Gain(credit_rating) = 0.9403 - 0.8922 = \mathbf{0.0481}$$

$$SplitInfo_{credit_rating}(D) = -\frac{8}{14} * \log_2\left(\frac{8}{14}\right) - \frac{6}{14} * \log_2\left(\frac{6}{14}\right) = 0.9852$$

$$GainRatio(credit_rating) = \frac{0.0481}{0.9852} = \mathbf{0.0488}$$

age	income	student	credit_rating	buys computer
youth	high	no	fair	no
youth	high	no	excellent	no
middle_aged	high	no	fair	yes
senior	medium	no	fair	yes
senior	low	yes	fair	yes
senior	low	yes	excellent	no
middle_aged	low	yes	excellent	yes
youth	medium	no	fair	no
youth	low	yes	fair	yes
senior	medium	yes	fair	yes
youth	medium	yes	excellent	yes
middle_aged	medium	no	excellent	yes
middle_aged	high	yes	fair	yes
senior	medium	no	excellent	no

$$Info(D) = - \sum_{i=1}^m p_i \log_2(p_i)$$

$$Info_A(D) = \sum_{j=1}^v \frac{|D_j|}{|D|} \times Info(D_j)$$

- Calculate **Entropy** of Class attribute:

buys_computer	
yes	no
9	5

$$Info(D) = I(9,5) = -\frac{9}{14} \log_2\left(\frac{9}{14}\right) - \frac{5}{14} \log_2\left(\frac{5}{14}\right) = \mathbf{0.9403}$$

$$SplitInfo_A(D) = - \sum_{j=1}^v \frac{|D_j|}{|D|} \log_2\left(\frac{|D_j|}{|D|}\right)$$

$$Gain(A) = Info(D) - Info_A(D)$$

$$GainRatio(A) = \frac{Gain(A)}{SplitInfo(A)}$$

- Calculate **Gain Ratio** of all other attributes:

		Class		
		yes	no	
student	yes	6	1	7
	no	3	4	7
				14

$$Info_{student}(D) = \frac{7}{14} I(6,1) + \frac{7}{14} I(3,4)$$

$$= \frac{7}{14} * 0.5917 + \frac{7}{14} * 0.9852 = 0.2958 + 0.4926 = \mathbf{0.7884}$$

$$Gain(student) = 0.9403 - 0.7884 = \mathbf{0.1519}$$

$$SplitInfo_{student}(D) = -\frac{7}{14} \log_2\left(\frac{7}{14}\right) - \frac{7}{14} \log_2\left(\frac{7}{14}\right) = 1$$

$$GainRatio(student) = \frac{0.1519}{1} = \mathbf{0.1519}$$

		Class		
		yes	no	
credit_rating	fair	6	2	8
	excellent	3	3	6
				14

$$Info_{credit_rating}(D) = \frac{8}{14} I(6,2) + \frac{6}{14} I(3,3)$$

$$= \frac{8}{14} * 0.8113 + \frac{6}{14} * 1 = 0.4636 + 0.4286 = \mathbf{0.8922}$$

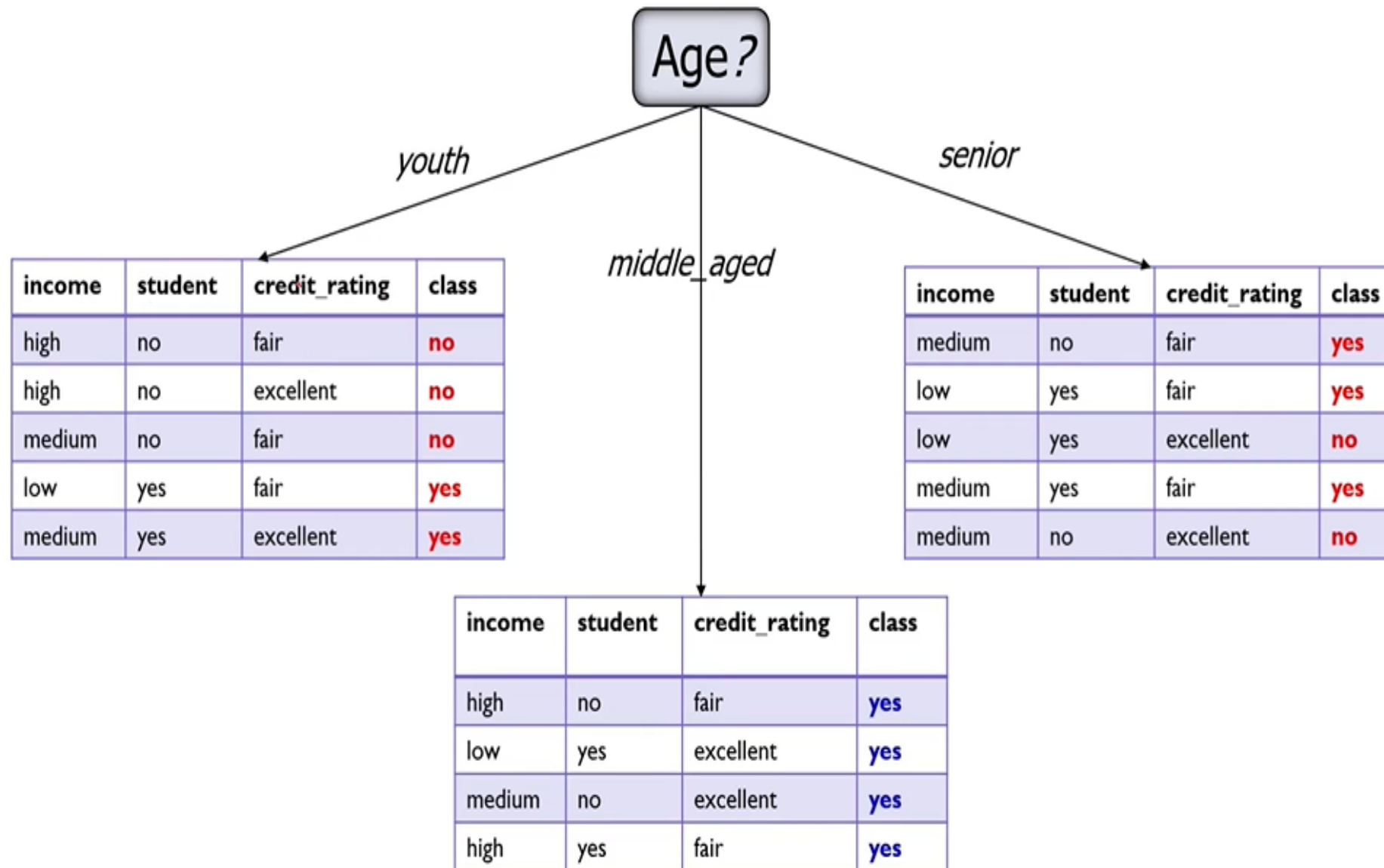
$$Gain(credit_rating) = 0.9403 - 0.8922 = \mathbf{0.0481}$$

$$SplitInfo_{credit_rating}(D) = -\frac{8}{14} \log_2\left(\frac{8}{14}\right) - \frac{6}{14} \log_2\left(\frac{6}{14}\right) = 0.9852$$

$$GainRatio(credit_rating) = \frac{0.0481}{0.9852} = \mathbf{0.0488}$$

age	income	student	credit_rating	buys_computer
youth	high	no	fair	no
youth	high	no	excellent	no
middle_aged	high	no	fair	yes
senior	medium	no	fair	yes
senior	low	yes	fair	yes
senior	low	yes	excellent	no
middle_aged	low	yes	excellent	yes
youth	medium	no	fair	no
youth	low	yes	fair	yes
senior	medium	yes	fair	yes
youth	medium	yes	excellent	yes
middle_aged	medium	no	excellent	yes
middle_aged	high	yes	fair	yes
senior	medium	no	excellent	no

- As, the Gain Ratio of "age" is highest,
- So "**age**" is the best attribute & becomes the root node of the decision tree.



- For Left subtree: Calculate **Entropy** of Class attribute:

buys_computer	
yes	no
2	3

$$Info(D) = I(2,3) = -\frac{2}{5} \log_2 \left(\frac{2}{5} \right) - \frac{3}{5} \log_2 \left(\frac{3}{5} \right) = \underline{0.971} \checkmark$$

- Calculate **Gain Ratio** of all other attributes:

		Class		
		yes	no	
income	low	1	0	1
	medium	1	1	2
	high	0	2	2
				5

$$Info_{income}(D) = \frac{1}{5} I(1,0) + \frac{2}{5} I(1,1) + \frac{2}{5} I(0,2)$$

$$= \frac{1}{5} * 0 + \frac{2}{5} * 1 + \frac{2}{5} * 0 = 0 + 0.4 + 0 = \underline{0.4} \checkmark$$

$$Gain(income) = 0.971 - 0.4 = \underline{0.571} \checkmark$$

$$SplitInfo_{income}(D) = -\frac{1}{5} * \log_2 \left(\frac{1}{5} \right) - \frac{2}{5} * \log_2 \left(\frac{2}{5} \right) - \frac{2}{5} * \log_2 \left(\frac{2}{5} \right) = \underline{1.5219} \checkmark$$

$$GainRatio(income) = \frac{0.571}{1.5219} = \underline{0.3751} \checkmark$$

		Class		
		yes	no	
credit_rating	fair	1	2	3
	excellent	1	1	2
				5

$$Info_{credit_rating}(D) = \frac{3}{5} I(1,2) + \frac{2}{5} I(1,1)$$

$$= \frac{3}{5} * 0.9183 + \frac{2}{5} * 1 = 0.3443 + 0.4 = \underline{0.7443}$$

$$Gain(credit_rating) = 0.971 - 0.7443 = \underline{0.2267}$$

$$SplitInfo_{credit_rating}(D) = -\frac{3}{5} * \log_2 \left(\frac{3}{5} \right) - \frac{2}{5} * \log_2 \left(\frac{2}{5} \right) = \underline{0.9709}$$

$$GainRatio(credit_rating) = \frac{0.2267}{0.9709} = \underline{0.2335} \checkmark$$

$$Info(D) = -\sum_{i=1}^m p_i \log_2(p_i)$$

$$Info_A(D) = \sum_{j=1}^v \frac{|D_j|}{|D|} \times Info(D_j)$$

$$Gain(A) = Info(D) - Info_A(D)$$

$$GainRatio(A) = \frac{Gain(A)}{SplitInfo(A)}$$

income	student	credit_rating	class
high	no	fair	no
high	no	excellent	no
medium	no	fair	no
low	yes	fair	yes
medium	yes	excellent	yes

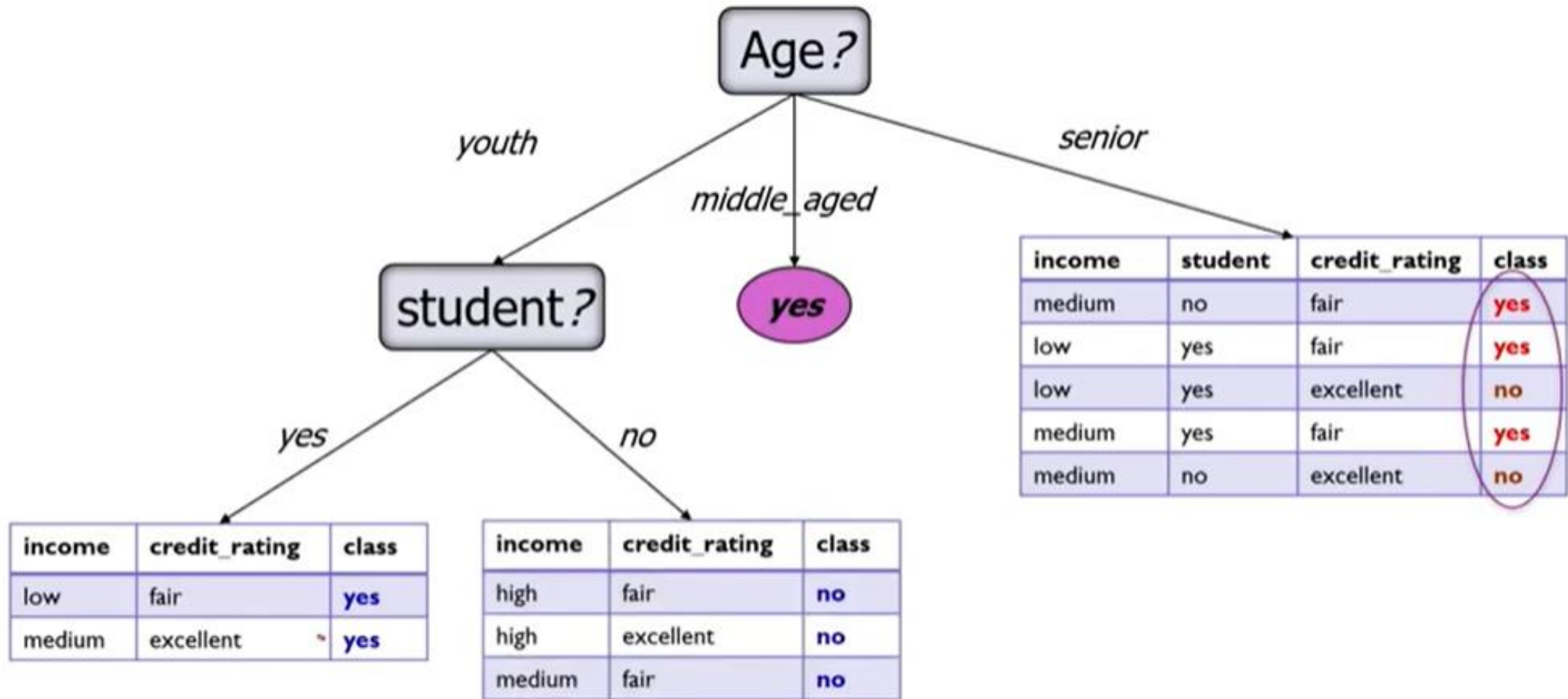
		Class		
		yes	no	
student	yes	2	0	2
	no	0	3	3
				5

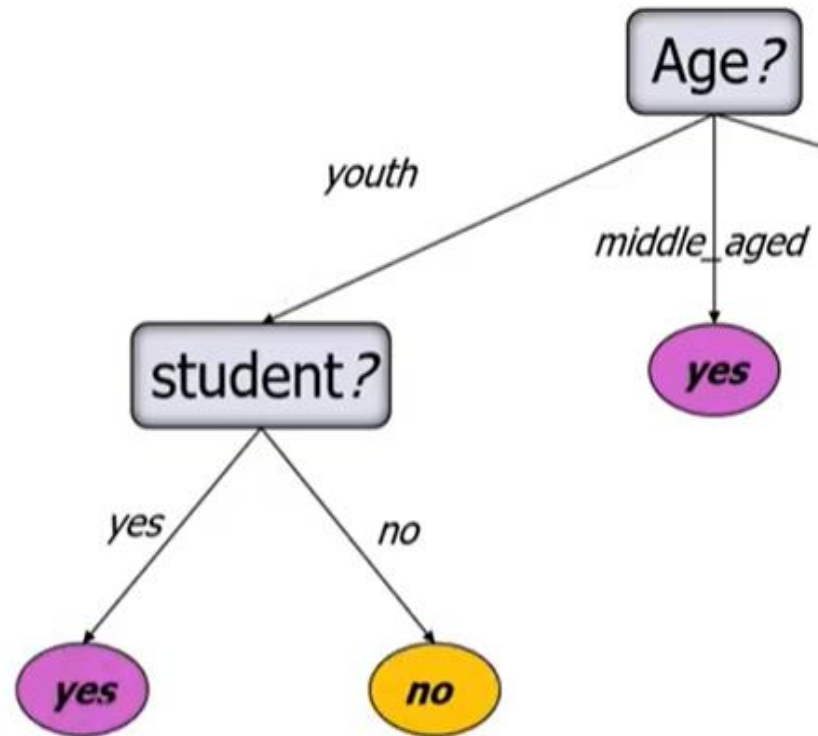
$$Info_{student}(D) = \frac{2}{5} I(2,0) + \frac{3}{5} I(0,3) = \frac{2}{5} * 0 + \frac{3}{5} * 0 = \underline{0}$$

$$Gain(student) = 0.971 - 0 = \underline{0.971}$$

$$SplitInfo_{student}(D) = -\frac{2}{5} * \log_2 \left(\frac{2}{5} \right) - \frac{3}{5} * \log_2 \left(\frac{3}{5} \right) = \underline{0.9709}$$

$$GainRatio(student) = \frac{0.971}{0.9709} = \underline{1} \checkmark$$





income	student	credit_rating	class
medium	no	fair	yes
low	yes	fair	yes
low	yes	excellent	no
medium	yes	fair	yes
medium	no	excellent	no

Gain Ratio [C4.5] - Example

- For Right subtree: Calculate **Entropy** of Class attribute:

buys_computer	
yes	no
3	2

$$Info(D) = I(3,2) = -\frac{3}{5} \log_2 \left(\frac{3}{5} \right) - \frac{2}{5} \log_2 \left(\frac{2}{5} \right) = 0.971 \checkmark$$

- Calculate **Gain Ratio** of all other attributes:

		Class		
		yes	no	
income	low	1	1	2
	medium	2	1	3
	high	0	0	0
				5

$$Info_{income}(D) = \frac{2}{5} I(1,1) + \frac{3}{5} I(2,1)$$

$$= \frac{2}{5} * 1 + \frac{3}{5} * 0.9183 = 0.4 + 0.551 = 0.951$$

$$Gain(income) = 0.971 - 0.951 = 0.02$$

$$SplitInfo_{income}(D) = -\frac{2}{5} * \log_2 \left(\frac{2}{5} \right) - \frac{3}{5} * \log_2 \left(\frac{3}{5} \right) = 0.9709$$

$$GainRatio(income) = \frac{0.02}{0.9709} = 0.0205 \checkmark$$

		Class		
		yes	no	
credit_rating	fair	3	0	3
	excellent	0	2	2
				5

$$Info_{credit_rating}(D) = \frac{3}{5} I(3,0) + \frac{2}{5} I(0,2) = \frac{3}{5} * 0 + \frac{2}{5} * 0 = 0$$

$$Gain(credit_rating) = 0.971 - 0 = 0.971$$

$$SplitInfo_{credit_rating}(D) = -\frac{3}{5} * \log_2 \left(\frac{3}{5} \right) - \frac{2}{5} * \log_2 \left(\frac{2}{5} \right) = 0.9709$$

$$GainRatio(credit_rating) = \frac{0.971}{0.9709} = 1$$

$$Info(D) = -\sum_{i=1}^n p_i \log_2(p_i)$$

$$Info_A(D) = \sum_{j=1}^v \frac{|D_j|}{|D|} \times Info(D_j)$$

$$Gain(A) = Info(D) - Info_A(D)$$

income	student	credit_rating	class
medium	no	fair	yes
low	yes	fair	yes
low	yes	excellent	no
medium	yes	fair	yes
medium	no	excellent	no

		Class		
		yes	no	
student	yes	2	1	3
	no	1	1	2
				5

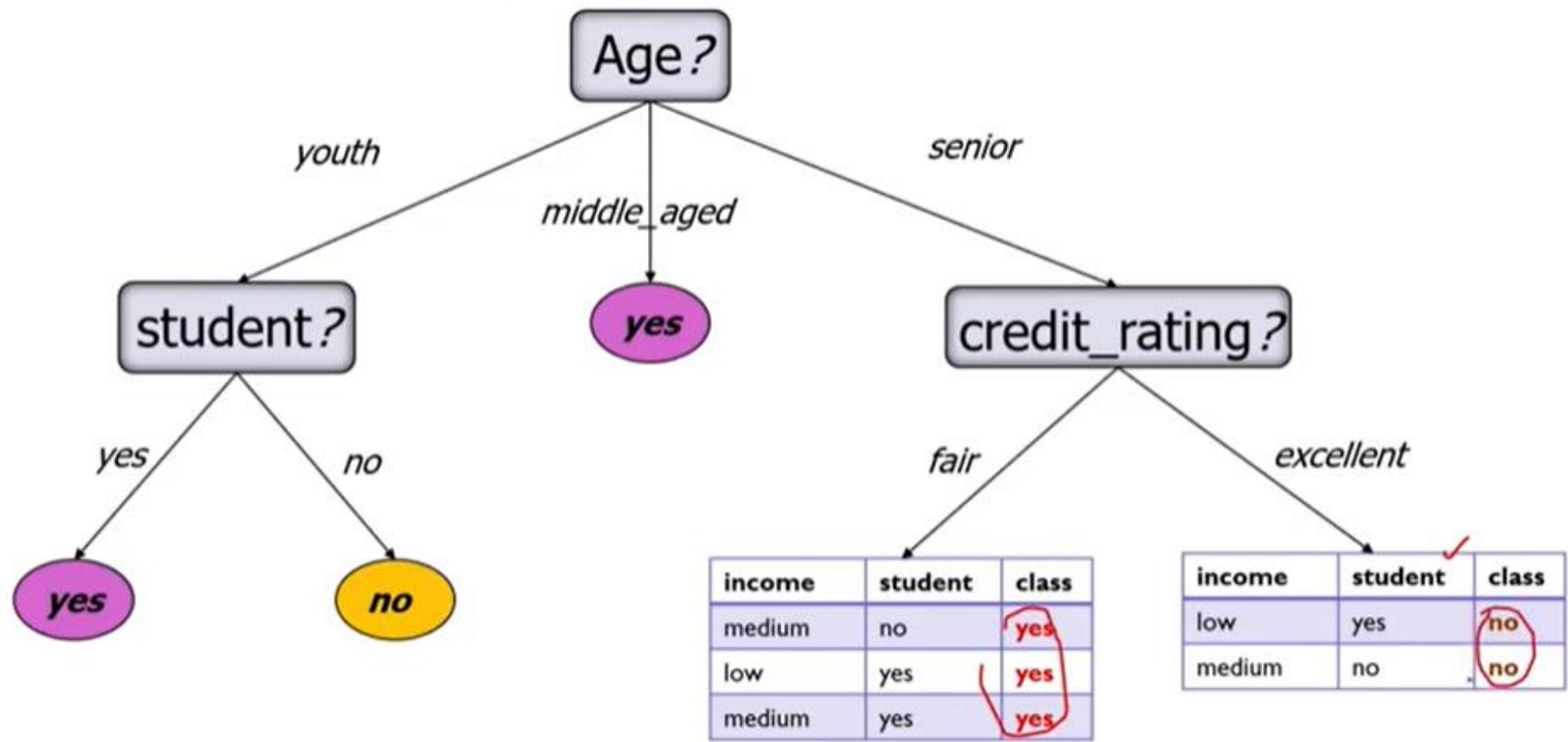
$$Info_{student}(D) = \frac{3}{5} I(2,1) + \frac{2}{5} I(1,1)$$

$$= \frac{3}{5} * 0.9183 + \frac{2}{5} * 1 = 0.551 + 0.4 = 0.951$$

$$Gain(student) = 0.971 - 0.951 = 0.02$$

$$SplitInfo_{student}(D) = -\frac{3}{5} * \log_2 \left(\frac{3}{5} \right) - \frac{2}{5} * \log_2 \left(\frac{2}{5} \right) = 0.9709$$

$$GainRatio(student) = \frac{0.02}{0.9709} = 0.0205$$



What is the decision for

- $X=[\text{age, income, student, credit}]=[15,\text{low},\text{no},\text{excellent}]$
- $X=[\text{age, income, student, credit}]=[40,\text{low},\text{no},\text{excellent}]$



Comparing Attribute Selection Measures

- The two measures, in general, return good results but
 - **Information gain:**
 - biased towards multivalued attributes
 - **Gain ratio:**
 - tends to prefer unbalanced splits in which one partition is much smaller than the others

