

第八讲：监督学习

- [决策树] 基于信息增益，对下述数据集进行决策树构建，描述过程一个关于配眼镜的一个决策分类所需要的数据，数据集包含 4 属性：age, astigmatism, trear-prod-rate 为输入特征，contact-lenses 为决策属性。

ID	AGE	ASTIGMATISM	TEAR-PRODUCTION-RATE	CONTACT-LENSES
1.	young	no	normal	soft
2.	young	yes	reduced	none
3.	young	yes	normal	hard
4.	pre-presbyopic	no	reduced	none
5.	pre-presbyopic	no	normal	soft
6.	pre-presbyopic	yes	normal	hard
7.	pre-presbyopic	yes	normal	none
8.	pre-presbyopic	yes	normal	none
9.	presbyopic	no	reduced	none
10.	presbyopic	no	normal	none
11.	presbyopic	yes	reduced	none
12.	presbyopic	yes	normal	hard

属性 age 的信息熵：

age = young = {1, 2, 3} soft 1/3 none 1/3 hard 1/3

age = pre_presbyopc = {4, 5, 6, 7, 8} soft 1/5 none 3/5 hard 1/5

age = byopic = {9, 10, 11, 12} soft 0/4 none 3/4 hard 1/3

分别: young, pre_presbyopc, byopic 的信息熵为: 1.585, 1.371, 0.811

$$H(p_i) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{1}{3} \log_2 \frac{1}{3} + \frac{1}{3} \log_2 \frac{1}{3} + \frac{1}{3} \log_2 \frac{1}{3} \right) = 1,585$$
$$H(p_i) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{1}{5} \log_2 \frac{1}{5} + \frac{2}{5} \log_2 \frac{3}{5} + \frac{1}{5} \log_2 \frac{1}{5} \right) = 1,371$$
$$H(p_i) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{3}{4} \log_2 \frac{3}{4} + \frac{1}{4} \log_2 \frac{1}{4} \right) = 0,811$$

属性 age 的信息增益为： 0.146

$$G(D, \text{age}) = H(D) - \sum_{v=1}^3 \frac{|D^v|}{|D|} \cdot H(D^v) = \left(\frac{3}{12} \cdot 1.585 + \frac{5}{12} \cdot 1.371 + \frac{4}{12} \cdot 0.811 \right) = 0.146$$

属性 astigmatism 的信息熵：

Astigmatism no = {1, 4, 5, 9, 10} soft 2/5 none 3/5 hard 0/5

Astigmatism yes = {2, 3, 6, 7, 8, 11, 12} soft 0/7 none 4/7 hard 3/7

分别：no, yes 的信息熵为：0.970, 0.985

$$H(D_1) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{2}{5} \log_2 \frac{2}{5} + \frac{3}{5} \log_2 \frac{3}{5} \right) = 0.970$$

$$H(D_2) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{4}{7} \log_2 \frac{4}{7} + \frac{3}{7} \log_2 \frac{3}{7} \right) = 0.985$$

属性 astigmatism 的信息增益为： 0.405

$$G(D, \text{astigmatism}) = H(D) - \sum_{v=1}^2 \frac{|D^v|}{|D|} \cdot H(D^v) = \left(\frac{5}{12} \cdot 0.970 + \frac{7}{12} \cdot 0.985 \right) = 0.405$$

属性 tear-production-rate (TPR) 的信息熵：

TPR reduced = {2, 4, 9, 11} soft 0/4 none 4/4 hard 0/4

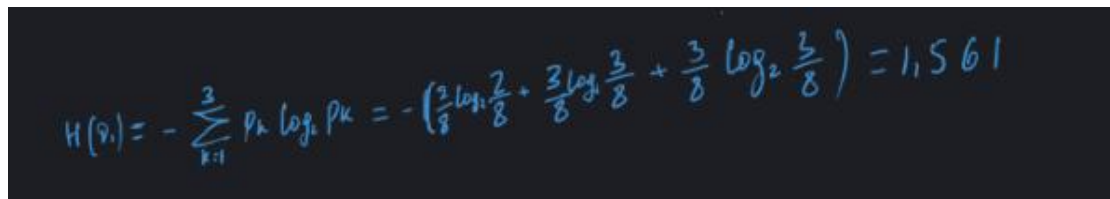
TPR normal = {1, 3, 5, 6, 7, 8, 10, 12} soft 2/8 none 3/8 hard 3/8

分别：reduced, normal 的信息熵为：0, 1.561

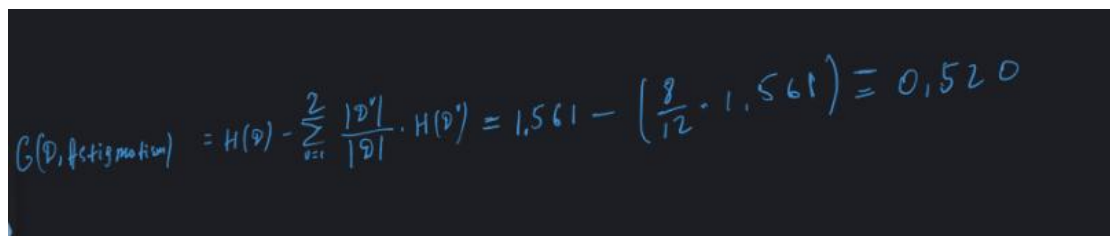
Reduced 因为所有数据集中到一个类所以 $H(D_{\text{reduced}}) = 0$

$H(D_{\text{normal}})$

=


$$H(D) = - \sum_{k=1}^3 p_k \log_2 p_k = - \left(\frac{2}{8} \log_2 \frac{2}{8} + \frac{3}{8} \log_2 \frac{3}{8} + \frac{3}{8} \log_2 \frac{3}{8} \right) = 1.561$$

属性 TPR 的信息增益为：0.520


$$G(D, \text{tear-production-rate}) = H(D) - \sum_{i=1}^2 \frac{|D_i|}{|D|} \cdot H(D_i) = 1.561 - \left(\frac{4}{12} \cdot 0 + \frac{8}{12} \cdot 1.561 \right) = 0.520$$

Tear-product-rate 的增益最大则把它选为划分属性：

