

Part I: Pen and paper

1. Complete the given decision tree using Shannon entropy (\log_2) and considering that: (i) a minimum of 4 observations is required to split an internal node, and (ii) decisions by ascending alphabetic should be placed in case of ties.

$$\begin{aligned} H(y_{\text{out}}|y_1 > 0.3) &= -P(y_{\text{out}} = A|y_1 > 0.3) \log_2(P(y_{\text{out}} = A|y_1 > 0.3)) \\ H(y_{\text{out}}|y_1 > 0.3) &= -P(y_{\text{out}} = B|y_1 > 0.3) \log_2(P(y_{\text{out}} = B|y_1 > 0.3)) \\ H(y_{\text{out}}|y_1 > 0.3) &= -P(y_{\text{out}} = C|y_1 > 0.3) \log_2(P(y_{\text{out}} = C|y_1 > 0.3)) \end{aligned}$$

$$H(y_{\text{out}}|y_1 > 0.3) = -\left(\frac{3}{9} \log_2\left(\frac{3}{9}\right) + \frac{2}{9} \log_2\left(\frac{2}{9}\right) + \frac{4}{9} \log_2\left(\frac{4}{9}\right)\right) \approx 1,53049$$

$$\begin{aligned} H(y_{\text{out}}|y_1 > 0.3, y_x) &= P(y_x = 0|y_1 > 0.3)H(y_{\text{out}}|y_1 > 0.3, y_x = 0) \\ &\quad + P(y_x = 1|y_1 > 0.3)H(y_{\text{out}}|y_1 > 0.3, y_x = 1) \\ &\quad + P(y_x = 2|y_1 > 0.3)H(y_{\text{out}}|y_1 > 0.3, y_x = 2) \end{aligned} \tag{1}$$

$$IG(y_{\text{out}}|y_1 > 0.3, y_x) = H(y_{\text{out}}|y_1 > 0.3) - H(y_{\text{out}}|y_1 > 0.3, y_x) \tag{2}$$

x=2:

$$P(y_2 = 0|y_1 > 0.3) = \frac{3}{9} \quad P(y_2 = 1|y_1 > 0.3) = \frac{2}{9} \quad P(y_2 = 2|y_1 > 0.3) = \frac{4}{9}$$

$$H(y_{\text{out}}|y_1 > 0.3, y_2 = 0) = -\left(\frac{1}{5} \log_2\left(\frac{1}{5}\right) + \frac{1}{5} \log_2\left(\frac{1}{5}\right) + \frac{3}{5} \log_2\left(\frac{3}{5}\right)\right) \approx 1.37095$$

$$H(y_{\text{out}}|y_1 > 0.3, y_2 = 1) = -\left(\frac{0}{2} \log_2\left(\frac{0}{2}\right) + \frac{1}{2} \log_2\left(\frac{1}{2}\right) + \frac{1}{2} \log_2\left(\frac{1}{2}\right)\right) = 1$$

$$H(y_{\text{out}}|y_1 > 0.3, y_2 = 2) = -\left(\frac{2}{2} \log_2\left(\frac{2}{2}\right) + \frac{0}{2} \log_2\left(\frac{0}{2}\right) + \frac{0}{2} \log_2\left(\frac{0}{2}\right)\right) = 0$$

$$H(y_{\text{out}}|y_1 > 0.3, y_2) = \frac{5}{9} \times 1.37095 + \frac{2}{9} \times 1 + \frac{2}{9} \times 0 \approx 0.98386$$

$$IG(y_{\text{out}}|y_1 > 0.3, y_2) = 1.53049 - 0.98386 = \mathbf{0.54663}$$

x=3:

$$P(y_3 = 0|y_1 > 0.3) = \frac{2}{9} \quad P(y_3 = 1|y_1 > 0.3) = \frac{2}{9} \quad P(y_3 = 2|y_1 > 0.3) = \frac{5}{9}$$

$$E(y_{\text{out}}|y_1 > 0.3, y_3 = 0) = -\left(\frac{0}{2}\log_2\left(\frac{0}{2}\right) + \frac{1}{2}\log_2\left(\frac{1}{2}\right) + \frac{1}{2}\log_2\left(\frac{1}{2}\right)\right) = 1$$

$$E(y_{\text{out}}|y_1 > 0.3, y_3 = 1) = -\left(\frac{1}{2}\log_2\left(\frac{1}{2}\right) + \frac{1}{2}\log_2\left(\frac{1}{2}\right) + \frac{0}{2}\log_2\left(\frac{0}{2}\right)\right) = 1$$

$$E(y_{\text{out}}|y_1 > 0.3, y_3 = 2) = -\left(\frac{2}{5}\log_2\left(\frac{2}{5}\right) + \frac{0}{5}\log_2\left(\frac{0}{5}\right) + \frac{3}{5}\log_2\left(\frac{3}{5}\right)\right) \approx 0.97095$$

$$E(y_{\text{out}}|y_1 > 0.3, y_3) = \frac{2}{9} \times 1 + \frac{2}{9} \times 1 + \frac{5}{9} \times 0.97095 \approx 0.98386$$

$$IG(y_{\text{out}}|y_1 > 0.3, y_3) = 1.53049 - 0.98386 = \mathbf{0.54663}$$

x=4:

$$P(y_4 = 0|y_1 > 0.3) = \frac{2}{9} \quad P(y_4 = 1|y_1 > 0.3) = \frac{4}{9} \quad P(y_4 = 2|y_1 > 0.3) = \frac{3}{9}$$

$$E(y_{\text{out}}|y_1 > 0.3, y_4 = 0) = -\left(\frac{1}{2}\log_2\left(\frac{1}{2}\right) + \frac{0}{2}\log_2\left(\frac{0}{2}\right) + \frac{1}{2}\log_2\left(\frac{1}{2}\right)\right) = 1$$

$$E(y_{\text{out}}|y_1 > 0.3, y_4 = 1) = -\left(\frac{1}{4}\log_2\left(\frac{1}{4}\right) + \frac{2}{4}\log_2\left(\frac{2}{4}\right) + \frac{1}{4}\log_2\left(\frac{1}{4}\right)\right) = 1.5$$

$$E(y_{\text{out}}|y_1 > 0.3, y_4 = 2) = -\left(\frac{1}{3}\log_2\left(\frac{1}{3}\right) + \frac{0}{3}\log_2\left(\frac{0}{3}\right) + \frac{2}{3}\log_2\left(\frac{2}{3}\right)\right) = 0.918295$$

$$E(y_{\text{out}}|y_1 > 0.3, y_4) = \frac{2}{9} \times 1 + \frac{4}{9} \times 1.5 + \frac{3}{9} \times 0.918295 \approx 1.19499$$

$$IG(y_{\text{out}}|y_1 > 0.3, y_4) = 1.53049 - 1.19499 = 0.3355$$

After calculating the information gains for each attribute, we can observe that both attributes y_2 and y_3 have the highest value of 0.54663.

Since we are faced with a tie, we choose y_2 as the next node, following the ascending alphabetical order (mentioned in point (ii) of the question summary). Considering that there are at least four observations with $y_1 > 0.3$, we split the new node.

2. Draw the training confusion matrix for the learnt decision tree.

		x_1	x_2	x_3	x_4	x_5	x_6	x_7	x_8	x_9	x_{10}	x_{11}	x_{12}
real	=	A	B	B	C	C	A	A	A	B	B	C	C
predicted	=	A	B	C	C	C	A	A	A	A	B	C	C

3. For more details on putting math into L^AT_EX documents you can see
4. When you get to the next problem, you can end the enumerate for the parts of the previous problem and then add another item.
 1. Use a nested enumerate environment to label the parts of the next problem.
 2. For a quick and broad overview of how to create documents in L^AT_EX see

Part II: Programming

3. Solution to the programming questions here.

End note: do not forget to also submit your Jupyter notebook