

CS2109S Tutorial 3

AY 25/26 Sem 1 — github/omgeta

A. 1. At root:

$$\begin{aligned}
 H(Y) &= 1 \\
 H(Y \mid \text{Education}) &= \frac{4}{10}H\left(\frac{1}{4}, \frac{3}{4}\right) + \frac{3}{10}H\left(\frac{1}{3}, \frac{2}{3}\right) + \frac{3}{10}H\left(\frac{3}{3}, 0\right) = 0.6 \\
 \implies IG(Y; \text{Education}) &= H(Y) - H(Y \mid \text{Education}) = 1 - 0.4 = 0.6 \\
 H(Y \mid \text{Age}) &= \frac{5}{10}H\left(\frac{3}{5}, \frac{2}{5}\right) + \frac{5}{10}H\left(\frac{2}{5}, \frac{3}{5}\right) = 0.971 \\
 \implies IG(Y; \text{Age}) &= H(Y) - H(Y \mid \text{Age}) = 1 - 0.971 = 0.029 \\
 H(Y \mid \text{Experience}) &= \frac{6}{10}H\left(\frac{3}{6}, \frac{3}{6}\right) + \frac{4}{10}H\left(\frac{2}{4}, \frac{2}{4}\right) = 1 \\
 \implies IG(\text{Experience}) &= H(Y) - H(Y \mid \text{Experience}) = 0
 \end{aligned}$$

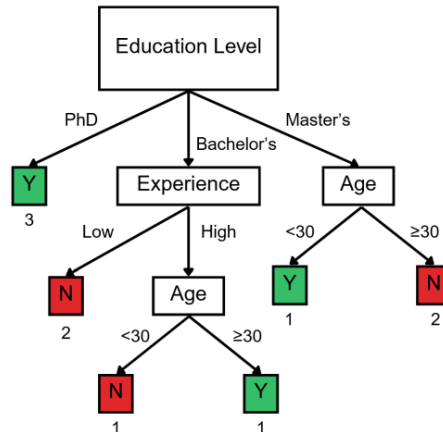
Since Education has the highest information gain, we choose it as root. Since all $\text{Education} = \text{PhD}$ have same classification, we only build a new subtree for $\text{Education} = \text{Masters}, \text{Bachelors}$. For $\text{Education} = \text{Masters}$:

$$\begin{aligned}
 H(Y \mid \text{Education} = \text{Masters}) &= H\left(\frac{1}{3}, \frac{2}{3}\right) = 0.918 \\
 H(Y \mid \text{Education} = \text{Masters}, \text{Experience}) &= \frac{2}{3}H\left(\frac{1}{2}, \frac{1}{2}\right) + \frac{1}{3}H(1, 0) = 0.667 \\
 \implies IG(Y \mid \text{Education} = \text{Masters}; \text{Experience}) &= 0.918 - 0.667 = 0.251 \\
 H(Y \mid \text{Education} = \text{Masters}, \text{Age}) &= \frac{1}{3}H\left(\frac{1}{1}, \frac{0}{1}\right) + \frac{2}{3}H\left(\frac{0}{2}, \frac{2}{2}\right) = 0 \\
 \implies IG(Y \mid \text{Education} = \text{Masters}; \text{Age}) &= 0.918 - 0 = 0.918
 \end{aligned}$$

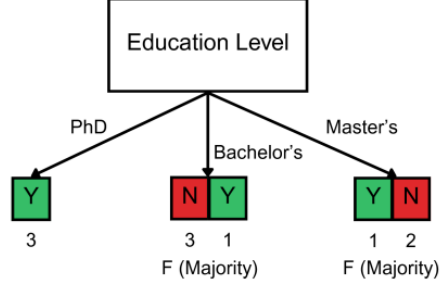
Since Age has the highest information gain, we choose it as subroot. For $\text{Education} = \text{Bachelors}$:

$$\begin{aligned}
 H(Y \mid \text{Education} = \text{Bachelors}) &= H\left(\frac{1}{4}, \frac{3}{4}\right) = 0.811 \\
 H(Y \mid \text{Education} = \text{Bachelors}, \text{Experience}) &= \frac{2}{4}H\left(\frac{2}{2}, \frac{0}{2}\right) + \frac{2}{4}H\left(\frac{1}{2}, \frac{1}{2}\right) = 0.5 \\
 \implies IG(Y \mid \text{Education} = \text{Bachelors}; \text{Experience}) &= 0.811 - 0.5 = 0.311 \\
 H(Y \mid \text{Education} = \text{Bachelors}, \text{Age}) &= \frac{2}{4}H\left(\frac{0}{2}, \frac{2}{2}\right) + \frac{2}{4}H\left(\frac{1}{2}, \frac{1}{2}\right) = 0.5 \\
 \implies IG(Y \mid \text{Education} = \text{Bachelors}; \text{Age}) &= 0.811 - 0.5 = 0.311
 \end{aligned}$$

With equal information gain, we tiebreak choosing Experience as subroot. We are left with splitting by Age in $\text{Experience} = \text{High}$



2. Outliers would be the applicant with Bachelors, High Experience and Age ≥ 30 , as well as the applicant with Masters and age < 3



3. Based on the predicted and actual decisions, $TP = 3, TN = 1, FP = 1, FN = 2$ so that $Accuracy = \frac{4}{7}, Precision = \frac{3}{4}, Recall = \frac{3}{5}, F1 = 0.667$ where $F1 = 0.667 > 0.6$ which indicates the model achieves a stronger balance than expected providing more reliable predictions for the positive class.

B. 1. Construct $X = \begin{pmatrix} 1 & 2 & 1 & 4 & 2 & 1 \\ 1 & 3 & 2 & 9 & 6 & 4 \\ 1 & 5 & 3 & 25 & 15 & 9 \\ 1 & 7 & 4 & 49 & 28 & 16 \\ 1 & 8 & 5 & 64 & 40 & 25 \\ 1 & 9 & 6 & 81 & 54 & 36 \end{pmatrix}$ then $w = \begin{pmatrix} 7.5 \\ -4 \\ 6.5 \\ -9.5 \\ 33.5 \\ -28 \end{pmatrix}$, giving

$$\hat{y} = 7.5 - 4x_1 + 6.5x_2 - 9.5x_1^2 + 33.5x_1x_2 - 28x_2^2$$

2. Column $x_3 = x_1^2 + 2x_1x_2 + x_2^2$ is a linear combination of the existing columns $\implies X$ loses full column rank $\implies X^T X$ is singular. We can drop one of the dependent columns.

- C. 1. For $\hat{y} = 2$, $L_{MSE} = 0.005, L_{MAE} = 0.05$;
For $\hat{y} = 4$, $L_{MSE} = 0.405, L_{MAE} = 0.45$
2. MSE magnifies large outliers

- D. 1. Given $y = x^2$, $\frac{dy}{dx} = 2x$, then $x_{t+1} = x_t - \gamma 2x_t = (1 - 2\gamma)x_t$

γ	t	x_t	$y_t = x_t^2$
all	0	5	25
10	1	-95	9025
	2	1805	3258025
	3	-34295	1176147025
	4	651605	424589076025
	5	-12380495	153276656445025
1	1	-5	25
	2	5	25
	3	-5	25
	4	5	25
	5	-5	25
0.1	1	4	16
	2	3.2	10.24
	3	2.56	6.5536
	4	2.048	4.1943
	5	1.6384	2.6844
0.01	1	4.9	24.01
	2	4.802	23.0592
	3	4.706	22.1461
	4	4.6118	21.2691
	5	4.5196	20.4268

$\therefore \gamma = 0.1$ converges the fastest

2. Add learning rate decay or an adaptive learning rate