

## Question 1a

Let  $p_1 = Pr(MPG = good)$  and  $p_2 = Pr(MPG = bad)$  then we can write the entropy of MPG as follows:

$$H(MPG) = \sum_{i=1}^2 (-p_i \times \log_2 p_i)$$

Since  $p_1 = \frac{3}{8}$  and  $p_2 = \frac{5}{8}$ ,

$$H(MPG) = \left(\frac{3}{8} \times \log_2 \frac{3}{8}\right) + \left(\frac{5}{8} \times \log_2 \frac{5}{8}\right) \approx 0.95$$

## Question 1b

Let  $X$  and  $Y$  be the random variables representing the Displacement and MPG respectively.

$$\begin{aligned} H(Y|X) &= Pr(X = low)H(Y|X = low) + Pr(X = medium)H(Y|X = medium) + Pr(X = high)H(Y|X = high) \\ &= \frac{4}{8}H\left(\frac{1}{4}, \frac{3}{4}\right) + \frac{2}{8}H\left(\frac{1}{2}, \frac{1}{2}\right) + \frac{2}{8}H\left(\frac{1}{2}, \frac{1}{2}\right) \\ &\approx \frac{4}{8}0.81 + \frac{2}{8} + \frac{2}{8} \approx 0.9 \end{aligned}$$

Mutual information between Displacement and MPG can be written as:

$$I(MPG; Displacement) = H(MPG) - H(MPG|Displacement) \approx 0.95 - 0.9 = 0.05$$

## Question 1c

Let  $X$  and  $Y$  be the random variables representing the Horsepower and MPG respectively.

$$\begin{aligned} H(Y|X) &= Pr(X = low)H(Y|X = low) + Pr(X = medium)H(Y|X = medium) + Pr(X = high)H(Y|X = high) \\ &= \frac{2}{8}H(1, 0) + \frac{4}{8}H\left(\frac{1}{4}, \frac{3}{4}\right) + \frac{2}{8}H(0, 1) \\ &\approx \frac{4}{8}0.81 \approx 0.4 \end{aligned}$$

Mutual information between Displacement and MPG can be written as:

$$I(MPG; Horsepower) = H(MPG) - H(MPG|Horsepower) \approx 0.95 - 0.4 = 0.55$$

## Question 1d

Full decision tree is shown in Figure 1. The numbers on the left and right represents the good and bad examples respectively.

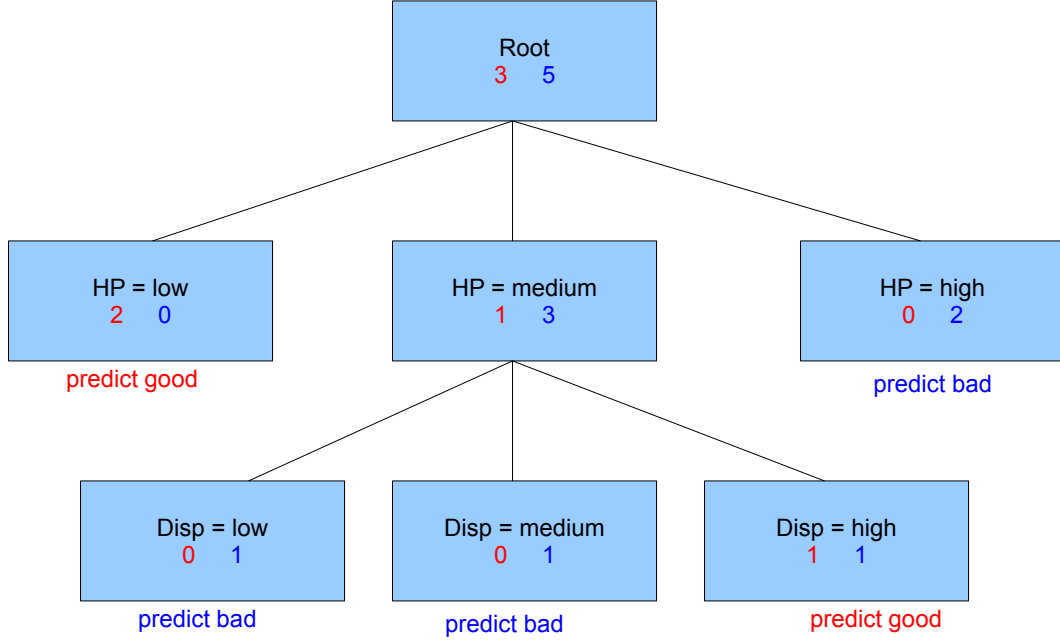


Figure 1: Full decision tree.

## Question 2

We can construct a dataset as follows:

A	B	C
0	1	0
1	0	0
0	0	1
1	1	1

$I(Y;A) = I(Y;B) = 0$ . But we can predict the full decision tree perfectly as in Figure 2. The  $I(Y;A)$  and  $I(Y;B)$  values are verified using the program of Question 4.

## Question 3

Since we are in 1D,  $w$  will be a scalar value and we can compute the norm of  $w$  as follows:

$$||w|| = \sqrt{w'w} = \sqrt{w^2} = w$$

Then we can rewrite the optimization problem as follows:

$$\begin{aligned} & \min_{w,b} \\ & \text{subject to } y_i(wx_i + b) \geq 1, \quad i = 1, 2 \end{aligned}$$

We can plug the  $x$  and  $y$  values into the problem as follows:

$$\min_{w,b}$$

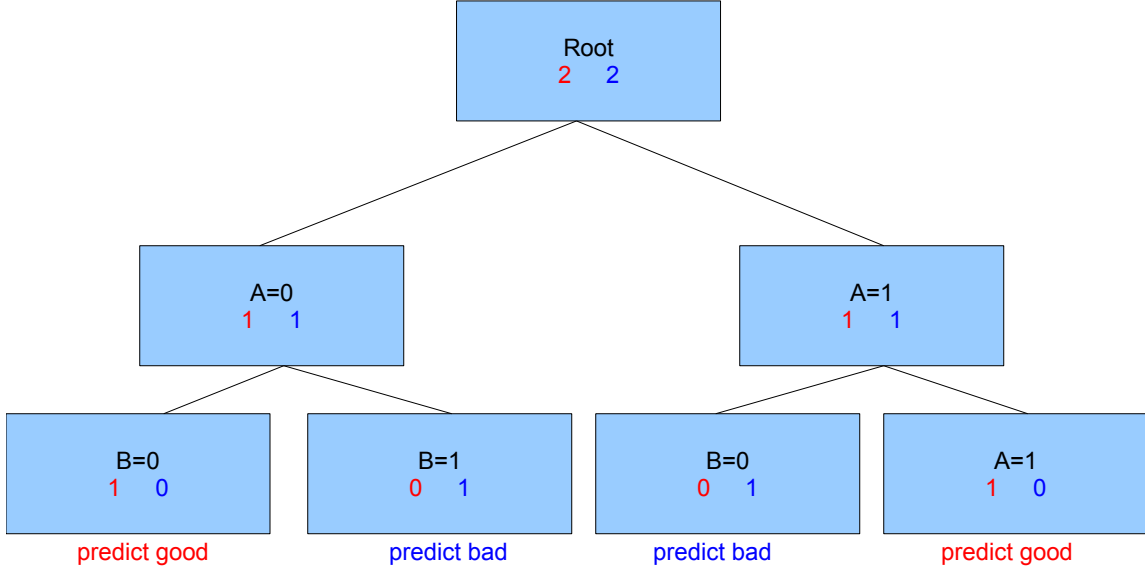


Figure 2: Full decision tree.

*subject to*  $-1(w0 + b) \geq 1$

*subject to*  $1(w1 + b) \geq 1$

Which is equal to:

$\min_{w,b} w$

*subject to*  $b \leq -1$

*subject to*  $w + b \geq 1$

We can rewrite the constraints as follows:

*subject to*  $1 - w \leq b \leq -1$

The lower bound for  $w$  is obviously 2 therefore the objective value of the optimization problem is  $w = 2$ . At this case,  $b = -1$ .