

1.

① misclassification rates

$$A = \frac{100 + 100}{(300 + 100) + (100 + 300)} = 0.25$$

$$B = \frac{0 + 200}{(200 + 0) + (200 + 400)} = 0.25$$

② Entropy

$$\begin{aligned} A &= \frac{1}{2} \times 2 \left(-\frac{3}{4} \log_2 \frac{3}{4} - \frac{1}{4} \log_2 \frac{1}{4} \right) \\ &= \left[-\frac{3}{4} (\log_2 3 - 2) - \frac{1}{4} \cdot (-2) \right] \\ &= 2 - \frac{3}{4} \log_2 3 = 0.8113 \end{aligned}$$

$$\begin{aligned} B &= \frac{1}{4} \cdot 0 + \frac{3}{4} \left(-\frac{1}{3} \log_2 \frac{1}{3} - \frac{2}{3} \log_2 \frac{2}{3} \right) \\ &= \frac{3}{4} \left[\frac{1}{3} \log_2 3 - \frac{2}{3} (1 - \log_2 3) \right] \\ &= \frac{3}{4} \left(\log_2 3 - \frac{2}{3} \right) = \frac{3}{4} \log_2 3 - \frac{1}{2} = 0.6887 \end{aligned}$$

③ Gini

$$\begin{aligned} A &= \frac{1}{2} \left[1 - \left(\frac{3}{4} \right)^2 - \left(\frac{1}{4} \right)^2 \right] + \frac{1}{2} \left[1 - \left(\frac{1}{4} \right)^2 - \left(\frac{3}{4} \right)^2 \right] \\ &= 1 - \frac{10}{16} = \frac{3}{8} = 0.3750 \end{aligned}$$

$$\begin{aligned} B &= \frac{1}{4} \left[1 - (1)^2 - (0)^2 \right] + \frac{3}{4} \left[1 - \left(\frac{1}{3} \right)^2 - \left(\frac{2}{3} \right)^2 \right] \\ &= \frac{3}{4} \left(1 - \frac{5}{9} \right) = \frac{3}{4} \left(\frac{4}{9} \right) = \frac{1}{3} = 0.3333 \end{aligned}$$

2.

$$\text{Let } J = E[e^{-t y(x)} | x] = P(t=1|x) e^{-y(x)} + P(t=-1|x) e^{y(x)}$$

$$\begin{aligned} \text{assume that } \frac{\delta J}{\delta y(x)} &= -P(t=1|x) e^{-y(x)} + P(t=-1|x) e^{y(x)} = 0 \\ \Rightarrow P(t=1|x) e^{-y(x)} &= P(t=-1|x) e^{y(x)} \end{aligned}$$

$$\text{Taking } \ln \text{ on both sides } \Rightarrow \ln P(t=1|x) - y(x) = \ln P(t=-1|x) + y(x) = 0$$

$$\Rightarrow \ln \frac{P(t=1|x)}{P(t=-1|x)} = 2y(x)$$

$$\therefore y(x) = \frac{1}{2} \ln \frac{P(t=1|x)}{P(t=-1|x)}$$

Alternative :

$$\text{Let } J(y) = E[e^{-t y(x)}]$$

Suppose we have a current estimate $y(x)$

and seek an improved estimate $y(x) + f(x)$

by minimizing $J(e^{-t y(x)} e^{-t f(x)})$ at each x .

$$J(y(x) + f(x)) = E[e^{-t y(x)} e^{-t f(x)} | x]$$

$$= e^{-f(x)} E[e^{-t y(x)} 1_{\{t=1\}} | x] + e^{f(x)} E[e^{-t y(x)} 1_{\{t=-1\}} | x],$$

$$\text{Where } E[1_{\{t=j\}} | x] = P(t=j|x)$$

Dividing this through by $E[e^{-t y(x)} | x]$ and setting its derivative w.r.t. $f(x)$ to zero

$$\text{, we get } f(x) = \frac{1}{2} \ln \frac{E_w[1_{\{t=1\}} | x]}{E_w[1_{\{t=-1\}} | x]}$$

$$= \frac{1}{2} \ln \frac{P_w[1_{\{t=1\}} | x]}{P_w[1_{\{t=-1\}} | x]}, \text{ where } w(x, t) = e^{-t y(x)}$$

$$\text{Therefore, } y(x) = \frac{1}{2} \ln \frac{P(t=1|x)}{P(t=-1|x)}$$