

11 2.

$$P(\text{predict win} | \text{win}) = 0.766$$

$$P(\text{predict loss} | \text{loss}) = 0.892$$

$$P(\text{win} | \text{predict win}) = \frac{P(\text{predict win} | \text{win}) P(\text{win})}{P(\text{predict win})} = \frac{(0.766)(0.045)}{P(\text{predict win})}$$

$$P(\text{loss} | \text{predict win}) = \frac{P(\text{predict win} | \text{loss}) P(\text{loss})}{P(\text{predict win})} = \frac{(1 - 0.892)(1 - 0.045)}{P(\text{predict win})}$$

$$1 = P(\text{win} | \text{predict win}) + P(\text{loss} | \text{predict win})$$

$$1 = \frac{0.03447 + 0.10314}{P(\text{predict win})}$$

$$\Rightarrow P(\text{predict win}) = 0.13761$$

$$\Rightarrow P(\text{win} | \text{predict win}) = \frac{0.766 (0.045)}{0.13761} = \boxed{0.25}$$

$$P(\text{win} | \text{predict loss}) = \frac{P(\text{predict loss} | \text{win}) P(\text{win})}{P(\text{predict loss})} = \frac{(1 - 0.766)(0.045)}{P(\text{predict loss})}$$

$$P(\text{loss} | \text{predict loss}) = \frac{P(\text{predict loss} | \text{loss}) P(\text{loss})}{P(\text{predict loss})} = \frac{(0.892)(1 - 0.045)}{P(\text{predict loss})}$$

$$P(\text{win} | \text{predict loss}) + P(\text{loss} | \text{predict loss}) = \frac{0.01053 + 0.85186}{P(\text{predict loss})} = 1 \Rightarrow P(\text{predict loss}) = 0.86239$$

$$\Rightarrow P(\text{win} | \text{predict loss}) = \frac{(1 - 0.766)(0.045)}{0.86239}$$

$$= \boxed{0.0122}$$

$$b) P(\text{win} | \text{predict loss}) = \frac{P(\text{predict loss} | \text{win}) P(\text{win})}{P(\text{predict loss})} = \frac{[1 - P(\text{predict win} | \text{win})] P(\text{win})}{P(\text{predict loss})}$$

$$P(\text{loss} | \text{predict loss}) = \frac{P(\text{predict loss} | \text{loss}) P(\text{loss})}{P(\text{predict loss})} = \frac{[0.892] [1 - P(\text{win})]}{P(\text{predict loss})}$$

$$P(\text{win} | \text{predict loss}) + P(\text{loss} | \text{predict loss})$$

$$\Rightarrow \frac{[1 - P(\text{predict win} | \text{win})] P(\text{win})}{P(\text{predict loss})} + \frac{(0.892)(1 - P(\text{win}))}{P(\text{predict loss})} = 1$$

$$0.234 P(\text{win}) + 0.892 - 0.892 P(\text{win}) = P(\text{predict loss})$$

$$- 0.658 P(\text{win}) + 0.892 = P(\text{predict loss})$$

$$\Rightarrow P(\text{win} | \text{predict loss}) = \frac{0.234 P(\text{win})}{0.892 - 0.658 P(\text{win})} = 0.5$$

$$\Rightarrow 0.234 P(\text{win}) = 0.446 - 0.329 P(\text{win})$$

$$P(\text{win}) = \frac{0.446}{0.234 + 0.329} = \boxed{0.79}$$

\*\* Plot on Word Doc \*\*

→ See ipynb code & notebook pdf showing 0.79 \*\*

$$c) P(\text{win} | \text{predict win}) = \frac{P(\text{predict win} | \text{win}) P(\text{win})}{P(\text{predict win})} = \frac{0.766 (0.15)}{P(\text{predict win})}$$

$$P(\text{loss} | \text{predict win}) = \frac{P(\text{predict win} | \text{loss}) P(\text{loss})}{P(\text{predict win})} = \frac{(1 - 0.892)(1 - 0.15)}{P(\text{predict win})}$$

$$\Rightarrow P(\text{predict win}) = (0.766)(0.15) + (1 - 0.892)(1 - 0.15) \\ = 0.2067$$

$$\Rightarrow P(\text{win} | \text{predict win}) = \frac{0.766 (0.15)}{0.2067} = \boxed{0.556}$$

$$d) \text{Accuracy} = P(\text{predict win} | \text{win}) \cdot P(\text{win}) + P(\text{predict loss} | \text{loss}) \cdot P(\text{loss})$$

$$= (0.766)(0.15) + (0.892)(0.85)$$

$$= \boxed{0.8731 \text{ or } 87.31\%}$$

$$e) P(\text{win} | \text{predict win}) = \frac{P(\text{predict win} | \text{win}) P(\text{win})}{P(\text{predict win})} = \frac{(0.766)(0.15)}{P(\text{predict win})}$$

$$P(\text{loss} | \text{predict win}) = \frac{P(\text{predict win} | \text{loss}) P(\text{loss})}{P(\text{predict win})} = \frac{(1 - \text{Specificity})(1 - 0.15)}{P(\text{predict win})}$$

$$\Rightarrow \frac{0.1149 + 0.85 - 0.85 \text{ Specificity}}{P(\text{predict win})} = 1$$

$$\Rightarrow P(\text{predict win}) = 0.9649 - 0.85 \text{ Specificity}$$

$$\Rightarrow P(\text{win} | \text{predict win}) = \frac{0.1149}{0.9649 - 0.85 \text{ Specificity}} \rightarrow \text{Plot to Python}$$

\*\* See Plot on Doc \*\*

f) \*\* See Doc \*\*

2) 2. Set  $MSE = \sum_{i=1}^N (x_i - a)^2$  as the mean squared error

$$= \sum_{i=1}^N x_i^2 - 2ax_i + a^2$$

$$\frac{d}{da}(MSE) = \sum_{i=1}^N -2x_i + 2a$$

$$= 2 \sum_{i=1}^N (a - x_i)$$

Setting  $\frac{dMSE}{da} = 0$  to find critical points.

$$\Rightarrow 2 \sum_{i=1}^N (a - x_i) = 0$$

$$\sum_{i=1}^N a - \sum_{i=1}^N x_i = 0$$

$$aN - \sum_{i=1}^N x_i = 0$$

$$\Rightarrow a_{\text{critical}}^* = \frac{\sum_{i=1}^N x_i}{N} = \bar{x} \text{ (Mean)}$$

To check whether or not this is a value that provides  $MSE_{\min}$ ,

take  $\frac{d^2 MSE}{da^2} = \frac{d}{da} \left( 2 \sum_{i=1}^N (a - x_i) \right)$

$$= 2 \sum_{i=1}^N 1$$

$$= 2N > 0$$

for  $N > 0$

↳ Given since  $N = \#$  of terms

$\Rightarrow a_{\text{critical}}$  gives  $MSE_{\min}$

QED

$$\begin{aligned}
 2) \text{ b. } \frac{d}{da} (\log p(x|a, \sigma^2)) &= \frac{d}{da} \left( -\frac{1}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} (x-a)^2 \right) \\
 &= -\frac{1}{2\sigma^2} (2)(-1)(x-a) \\
 &= \frac{1}{\sigma^2} (x-a) = 0 \\
 \Rightarrow \underline{x_{\text{opt}} = a} &= \text{Mean (Same to } \frac{\partial}{\partial} \text{)}
 \end{aligned}$$

$$\begin{aligned}
 2) \text{ c. Bernoulli } (y; \sigma(z)) &= \sigma(z)^y (1-\sigma(z))^{(1-y)} \\
 -\log \text{ Bernoulli } (y; \sigma(z)) &= -\log (\sigma(z)^y (1-\sigma(z))^{(1-y)}) \\
 &= -\log (\sigma(z)^y) - \log (1-\sigma(z))^{(1-y)} \\
 &= -y \log(\sigma(z)) - (1-y) \log(1-\sigma(z)) \\
 &= \ell(y, \sigma(z)) \\
 &\quad \underline{\text{QED}}
 \end{aligned}$$