



AML5202 | Deep Learning | In-class Quiz-1

Consider a dataset in which a sample has 5 features:

(1) *heart rate* (2) *blood pressure* (3) *temperature* (4) *age* (5) *weight*,

and a target variable *in hospital death* that has two levels *yes* and *no*. Suppose we build a logistic regression model using the dataset. Choose the correct answer for the following questions:

1. $x_2^{(1)}$ represents
 - A. 1st patients heart rate
 - B. 2nd patients heart rate
 - C. 2nd patients blood pressure
 - D. 1st patients blood pressure
2. The component w_3 of the weight vector is the weight applied to
 - A. Weight
 - B. Age
 - C. Blood pressure
 - D. Temperature
3. Which feature gets most weighted (ignoring sign) using the weight vector

$$w = [10^{-1} \quad 10^{-2} \quad -10^{-2} \quad -10^2 \quad 10] ?$$

- A. Age
 - B. Weight
 - C. Temperature
 - D. Blood pressure
4. $y^{(1)}$ represents the 1st patient's
 - A. predicted output label
 - B. correct output label
 - C. predicted probability that it belongs to label 1
 - D. predicted probability that it belongs to label 0
 5. Suppose the 1st patient survived. Then, $\hat{y}^{(1)}$ represents the 1st patient's
 - A. predicted output label
 - B. correct output label
 - C. predicted probability that they belong to label 1
 - D. predicted probability that they belong to label 0

6. Suppose the 1st patient did not survive. Then, $\hat{y}^{(1)}$ is equal to
- $1 - \sigma(w \cdot x^{(1)})$
 - $\sigma(w \cdot x^{(1)})$
 - $\sigma(w \cdot x^{(1)}) \times (1 - \sigma(w \cdot x^{(1)}))$
 - $\sigma(w \cdot x^{(1)})^{(1-y^{(1)})} \times (1 - \sigma(w \cdot x^{(1)}))^{y^{(1)}}$
7. Suppose there are 4 patients in the dataset in which first two survived and the last two did not. Using a particular weight vector w , we get:

$$\begin{aligned}\sigma(w \cdot x^{(1)}) &= 0.1, \\ 1 - \sigma(w \cdot x^{(2)}) &= 0.2, \\ \sigma(w \cdot x^{(3)}) &= 0.9, \\ 1 - \sigma(w \cdot x^{(4)}) &= 0.15.\end{aligned}$$

Without calculation, we can say that the loss is the highest for patient

- 1
 - 2
 - 3
 - 4
8. The bias feature value for all samples is equal to
- 1
 - 0
 - 1
 - any fixed positive number
9. If there are 100 patients' information in the dataset, then the size of the data matrix after the bias trick is
- 100×5
 - 6×100
 - 101×5
 - 100×6
10. Suppose we have 100 patients of which 10 survived and the remaining did not. Which of the following is a suitable loss definition for the i th sample to address the output label imbalance?
- $L = [-y^{(i)} \log(\sigma(w \cdot x^{(i)}))] + [-(1 - y^{(i)}) \log(1 - \sigma(w \cdot x^{(i)}))]$
 - $L = 0.9 [-y^{(i)} \log(\sigma(w \cdot x^{(i)}))] + 0.1 [-(1 - y^{(i)}) \log(1 - \sigma(w \cdot x^{(i)}))]$
 - $L = 0.1 [-y^{(i)} \log(\sigma(w \cdot x^{(i)}))] + 0.9 [-(1 - y^{(i)}) \log(1 - \sigma(w \cdot x^{(i)}))]$
 - $L = 0.5 [-y^{(i)} \log(\sigma(w \cdot x^{(i)}))] + 0.5 [-(1 - y^{(i)}) \log(1 - \sigma(w \cdot x^{(i)}))]$