機器學習概論作業

範圍: Backprogation Step by Step

銘傳大學電腦與通訊工程系

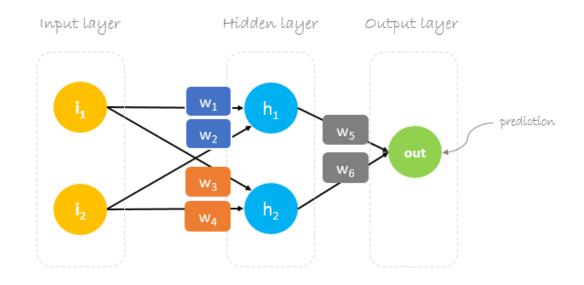
班	級	電通三乙
姓	名	李柏賢
學	號	07050862
作業	成果	應繳作業共 2 題,每題 50 分
		我共完成 <u>2</u> 題,應得 <u>100</u> 分
授課	教師	陳慶逸

■ 請確實填寫自己寫完成題數,填寫不實者(如上傳與作業明顯無關的答案,或是計算題數有誤者),本次作業先扣 50 分。

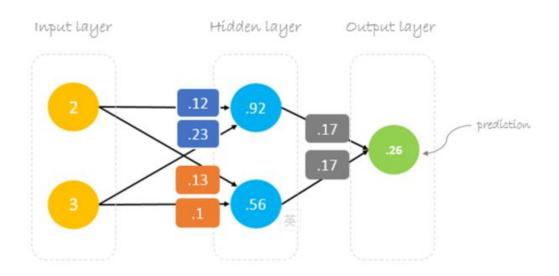
一、針對下面的 3 layer neural network,

其中
$$h_1 = \mathbf{w}_1 * \mathbf{i}_1 + \mathbf{w}_2 * \mathbf{i}_2, h_2 = \mathbf{w}_3 * \mathbf{i}_1 + \mathbf{w}_4 * \mathbf{i}_2,$$

$$prediction = \mathbf{w}_5 * h_1 + \mathbf{w}_6 * h_2$$



若權重 $w_1 = 0.12$, $w_2 = 0.23$, $w_3 = 0.13$, $w_4 = 0.1$, $w_5 = 0.17$, $w_6 = 0.17$;



1. 當 inputs=[2, 3], 期望輸出是 1 (但此時 neural network 的預測輸出 是 0.26)



2. Error =
$$\frac{1}{2}(prediction - actual)^2$$

試手算再經過一次梯度修正後,得到所有權重及輸出為(learning rate = 0.05):

$$w_1 = \underline{0.13258}$$
, $w_2 = \underline{0.24887}$, $w_3 = \underline{0.1428}$, $w_4 = \underline{0.11887}$, $w_5 = \underline{0.20404}$, $w_6 = \underline{0.19072}$

$$h_1 = \underline{0.69356}$$
, $h_2 = \underline{0.85435}$, $prediction = \underline{0.30446}$

$$\begin{bmatrix} w_1 & w_3 \\ w_2 & w_4 \end{bmatrix}$$

$$= \begin{bmatrix} w_1 & w_3 \\ w_2 & w_4 \end{bmatrix} - a * (prediction - actual) \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} [w_5 w_6]$$

$$= \begin{bmatrix} 0.12 & 0.13 \\ 0.23 & 0.1 \end{bmatrix} - 0.05 * (-0.74) \begin{bmatrix} 2 \\ 3 \end{bmatrix} [0.17 & 0.17]$$

$$= \begin{bmatrix} 0.12 & 0.13 \\ 0.23 & 0.1 \end{bmatrix} - \begin{bmatrix} -0.01258 & -0.01258 \\ -0.01887 & -0.01887 \end{bmatrix}$$

$$= \begin{bmatrix} 0.13258 & 0.1428 \\ 0.24887 & 0.11887 \end{bmatrix}$$

$$\begin{bmatrix} w_5 \\ w_6 \end{bmatrix}$$
= $\begin{bmatrix} w_5 \\ w_6 \end{bmatrix} - a * (prediction - actual) \begin{bmatrix} h_1 \\ h_2 \end{bmatrix}$
= $\begin{bmatrix} 0.17 \\ 0.17 \end{bmatrix} - 0.05 * (0.26 - 1) \begin{bmatrix} 0.92 \\ 0.56 \end{bmatrix}$
= $\begin{bmatrix} 0.17 \\ 0.17 \end{bmatrix} - \begin{bmatrix} -0.03404 \\ -0.02072 \end{bmatrix}$
= $\begin{bmatrix} 0.20404 \\ 0.19072 \end{bmatrix}$

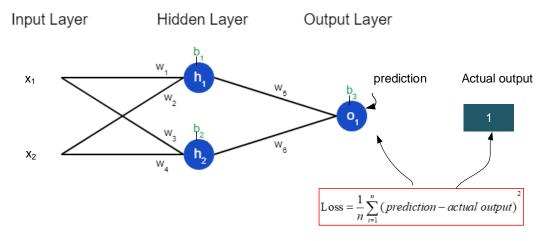
$$\begin{bmatrix} h_1 \\ h_2 \end{bmatrix} = \begin{bmatrix} w_1 & w_3 \\ w_2 & w_4 \end{bmatrix} \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} = \begin{bmatrix} 0.13258 & 0.14280 \\ 0.24887 & 0.11887 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 0.69356 \\ 0.85435 \end{bmatrix}$$

$$prediction = \begin{bmatrix} h_1 \ h_2 \end{bmatrix} \begin{bmatrix} W_5 \\ W_6 \end{bmatrix} = \begin{bmatrix} 0.69356 \ 0.85435 \end{bmatrix} \begin{bmatrix} 0.20404 \\ 0.19072 \end{bmatrix} = 0.30446$$

二、針對下面的 3 layer neural network,

其中
$$h_1 = f(w_1 * x_1 + w_2 * x_2 + b_1)$$
, $h_2 = f(w_3 * x_1 + w_4 * x_2 + b_2)$,

prediction = $f(w_5 * h_1 + w_6 * h_2 + b_3) \circ f$ 為 sigmoid function \circ



- 1. 當 inputs=[-2, -1], 期望輸出是 1 (但此時 neural network 的預測輸出 是?)
- 2. 誤差函數:

Loss =
$$\frac{1}{1} \sum_{i=1}^{1} (prediction - actual output)^{2}$$

= $(actual output - prediction)^{2}$
= $(1 - prediction)^{2}$

試推導

$$\frac{\partial Loss}{\partial w_{6}} = \frac{\partial Loss}{\partial w_{2}} = \frac{\partial Loss}{\partial w_{3}} = \frac{\partial Loss}{\partial b_{4}} = \frac{\partial Loss}{\partial b_{2}} = \frac{\partial Los$$

$$\frac{\partial Loss}{\partial b_3} =$$

$$\begin{split} &\frac{\partial Loss}{\partial w_{6}} \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial w_{6}} \\ &= \frac{\partial (1 - prediction)^{2}}{\partial prediction} * \frac{\partial f(w_{5} * h_{1} + w_{6} * h_{2} + b_{3})}{\partial w_{6}} \\ &= -2(1 - prediction) * h_{2} * f'(w_{5}h_{1} + w_{6}h_{2} + b_{3}) \end{split}$$

$$\frac{\partial Loss}{\partial w_{2}} = \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial h_{1}} * \frac{\partial h_{1}}{\partial w_{2}}$$

$$= \frac{\partial (1 - prediction)^{2}}{\partial prediction} * \frac{\partial f(h_{1}w_{5} + h_{2}w_{6} + b_{3})}{\partial h_{1}} * \frac{\partial f(w_{1}x_{1} + w_{2}x_{2} + b_{1})}{\partial w_{2}}$$

$$= -2(1 - prediction) * w_{5} * f'(h_{1}w_{5} + h_{2}w_{6} + b_{3}) * x_{2} * f'(w_{1}x_{1} + w_{2}x_{2} + b_{1})$$

$$\begin{split} &\frac{\partial Loss}{\partial w_3} \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial h_2} * \frac{\partial h_2}{\partial w_3} \\ &= \frac{\partial (1 - prediction)^2}{\partial prediction} * \frac{\partial f(w_5 h_1 + w_6 h_2 + b_3)}{\partial h_2} * \frac{\partial f(w_3 x_1 + w_4 x_2 + b_2)}{\partial w_3} \\ &= -2(1 - prediction) * w_6 * f'(w_5 h_1 + w_6 h_2 + b_3) * x_1 * f'(w_3 x_1 + w_4 x_2 + b_2) \end{split}$$

$$\begin{split} &\frac{\partial Loss}{\partial w_4} \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial h_2} * \frac{\partial h_2}{\partial w_4} \\ &= \frac{\partial (1 - prediction)^2}{\partial prediction} * \frac{\partial f(w_5 h_1 + w_6 h_2 + b_3)}{\partial h_2} * \frac{\partial f(w_3 x_1 + w_4 x_2 + b_2)}{\partial w_4} \\ &= -2(1 - prediction) * w_6 * f'(w_5 h_1 + w_6 h_2 + b_3) * x_2 * f'(w_3 x_1 + w_4 x_2 + b_2) \end{split}$$

$$\begin{split} &\frac{\partial Loss}{\partial b_{1}} = \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial h_{1}} * \frac{\partial h_{1}}{\partial b_{1}} \\ &= \frac{\partial (1 - prediction)^{2}}{\partial prediction} * \frac{\partial f(w_{5}h_{1} + w_{6}h_{2} + b_{3})}{\partial h_{1}} * \frac{\partial f(w_{1}x_{1} + w_{2}x_{2} + b_{1})}{\partial b_{1}} \\ &= -2(1 - prediction) * w_{5} * f'(w_{5}h_{1} + w_{6}h_{2} + b_{3}) * 1 * f'(w_{1}x_{1} + w_{2}x_{2} + b_{1}) \end{split}$$

$$\begin{split} &\frac{\partial Loss}{\partial b_2} \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial h_2} * \frac{\partial h_2}{\partial b_2} \\ &= \frac{\partial (1 - prediction)^2}{\partial prediction} * \frac{\partial f(h_1 w_5 + h_2 w_6 + b_3)}{\partial h_2} * \frac{\partial f(w_3 x_1 + w_4 x_2 + b_2)}{\partial b_2} \\ &= -2(1 - prediction) * w_6 * f'(h_1 w_5 + h_2 w_6 + b_3) * 1 * f'(w_3 x_1 + w_4 x_2 + b_2) \end{split}$$

$$\begin{split} &\frac{\partial Loss}{\partial b_3} \\ &= \frac{\partial Loss}{\partial prediction} * \frac{\partial prediction}{\partial b_3} \\ &= \frac{\partial (1 - prediction)^2}{\partial prediction} * \frac{\partial f(w_5 h_1 + w_6 h_2 + b_3)}{\partial b_3} \\ &= -2(1 - prediction) * 1 * f'(w_5 h_1 + w_6 h_2 + b_3) \end{split}$$