

Neural Network Basic Assignment

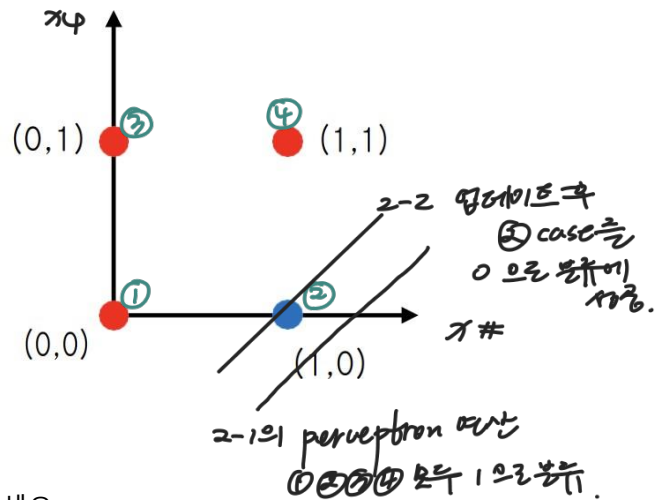
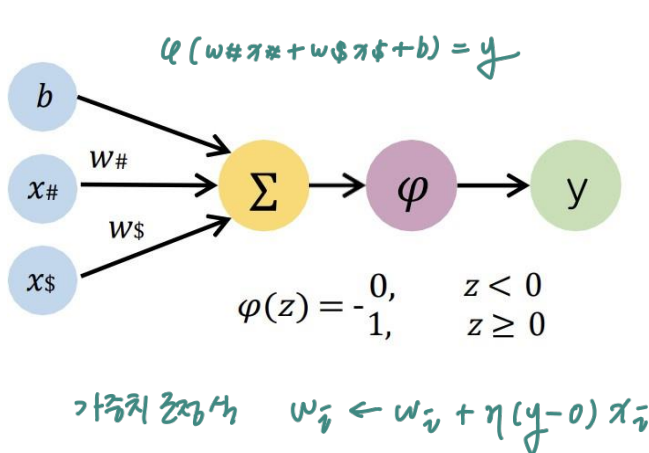
이름: 강호은

1. Sigmoid Function을 z 에 대해 미분하세요.

$$\sigma(z) = \frac{1}{1 + e^{-z}}$$

$$\begin{aligned} \frac{d}{dz} \sigma(z) &= \frac{-(-1) \times e^{-z}}{(1 + e^{-z})^2} = \frac{e^{-z}}{(1 + e^{-z})^2} = \frac{1 - 1 + e^{-z}}{(1 + e^{-z})^2} \\ &= \frac{(1 + e^{-z})}{(1 + e^{-z})^2} - \frac{1}{(1 + e^{-z})^2} = \frac{1}{1 + e^{-z}} - \frac{1}{(1 + e^{-z})^2} = \frac{1}{1 + e^{-z}} \left(1 - \frac{1}{1 + e^{-z}} \right) = \sigma(z) (1 - \sigma(z)) \end{aligned}$$

2. 다음과 같은 구조의 Perceptron과 ● (=1), ● (=0)을 평면좌표상에 나타낸 그림이 있습니다.



2-1. ●, ●을 분류하는 임의의 b, w 를 선정하고 분류해보세요.

$W_{\#} = -1.0$ $W_{\$} = 1.0$ $b = 1.2$ 라는 임의의 b, w 값을 설정한다.

$$\textcircled{1} (-1.0 \times 0) + (1.0 \times 0) + 1.2 = 1.2$$

$$\varphi(1.2) = 1$$

$$\textcircled{2} (-1.0 \times 1) + (1.0 \times 0) + 1.2 = 0.2$$

$$\varphi(0.2) = 1$$

$$\textcircled{3} (-1.0 \times 0) + (1.0 \times 1) + 1.2 = 2.2$$

$$\varphi(2.2) = 1$$

$$\textcircled{4} (-1.0 \times 1) + (1.0 \times 1) + 1.2 = 1.2$$

$$\varphi(1.2) = 1$$

2-2. Perceptron 학습 규칙에 따라 임의의 학습률을 정하고 b, w 를 1회 업데이트 해주세요.

임의의 학습률 0.15 설정. 잘못분류된 2번 케이스에 대하여 업데이트 진행.

$$b \leftarrow b + 0.15(0 - 1) \times 1.0$$

$$W_{\#} \leftarrow W_{\#} + 0.15(0 - 1) \times 1.0$$

$$W_{\$} \leftarrow W_{\$} + 0.15(0 - 1) \times 0$$

* 업데이트 후 분류

$$\textcircled{1} (-1.15 \times 0) + (1.0 \times 0) + 1.05 = 1.05, \varphi(1.05) = 1$$

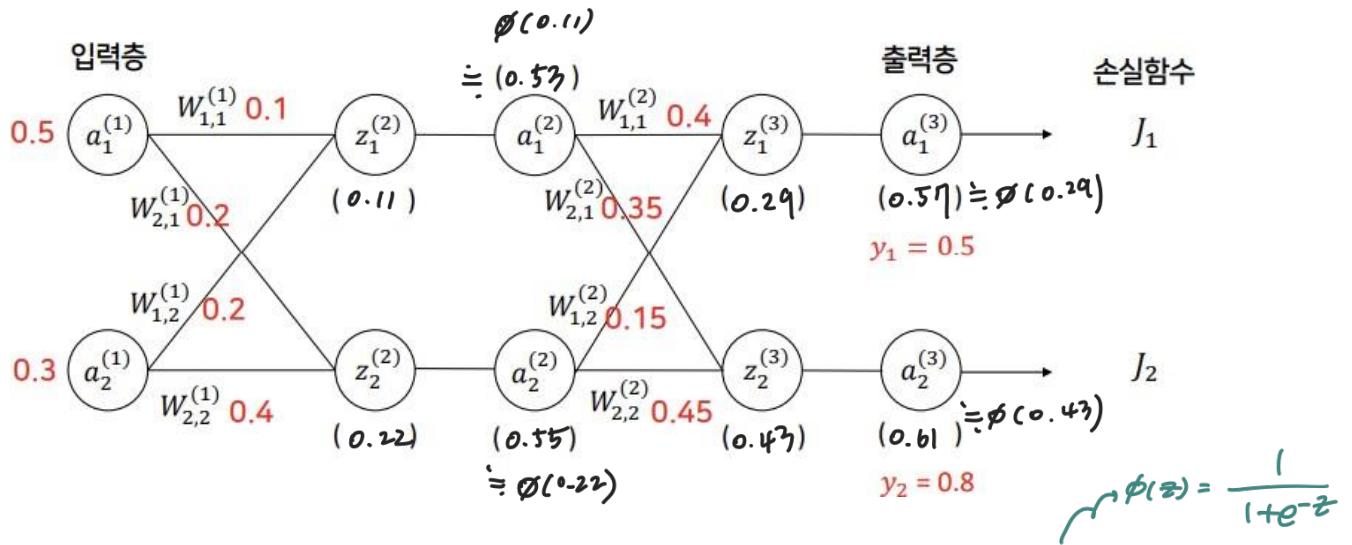
$$\textcircled{2} (-1.15 \times 1) + (1.0 \times 0) + 1.05 = -0.1, \varphi(-0.1) = 0$$

$$\textcircled{3} (-1.15 \times 0) + (1.0 \times 1) + 1.05 = 2.05, \varphi(2.05) = 1$$

$$\textcircled{4} (-1.15 \times 1) + (1.0 \times 1) + 1.05 = 0.9, \varphi(0.9) = 1$$

$\Rightarrow b = 1.05, W_{\#} = -1.15, W_{\$} = 1.0$ 으로 업데이트. ▶ 성공적으로 분류되는 것 확인할 수 있다.

3. 다음과 같이 입력과 가중치가 주어진 퍼셉트론이 있을 때, 아래의 물음에 답해주세요. 모든 문제는 풀이과정을 자세하게 적어주세요! (3-3까지 있습니다.)



- 3-1. FeedForward가 일어날 때, 각 노드가 갖는 값을 빈칸에 써주세요. 단, 활성화함수는 sigmoid 함수입니다. (모든 계산의 결과는 소수점 셋째자리에서 반올림하여 둘째자리까지만 써주세요.)

$$z_1^{(2)} = w_{1,1}^{(1)} a_1^{(1)} + w_{2,1}^{(1)} a_2^{(1)} = 0.1 \times 0.5 + 0.2 \times 0.3 = 0.11$$

$$a_1^{(2)} = \phi(z_1^{(2)}) = \frac{1}{1+e^{-0.11}}$$

$$z_2^{(2)} = w_{2,1}^{(1)} a_1^{(1)} + w_{2,2}^{(1)} a_2^{(1)} = 0.2 \times 0.5 + 0.4 \times 0.3 = 0.22$$

$$a_2^{(2)} = \phi(z_2^{(2)}) = \frac{1}{1+e^{-0.22}}$$

$$z_1^{(3)} = w_{1,1}^{(2)} a_1^{(2)} + w_{1,2}^{(2)} a_2^{(2)} = 0.4 \times 0.57 + 0.15 \times 0.55 \approx 0.29$$

$$a_1^{(3)} = \phi(z_1^{(3)}) = \frac{1}{1+e^{-0.29}}$$

$$z_2^{(3)} = w_{2,1}^{(2)} a_1^{(2)} + w_{2,2}^{(2)} a_2^{(2)} = 0.35 \times 0.57 + 0.45 \times 0.55 \approx 0.43$$

$$a_2^{(3)} = \phi(z_2^{(3)}) = \frac{1}{1+e^{-0.43}}$$

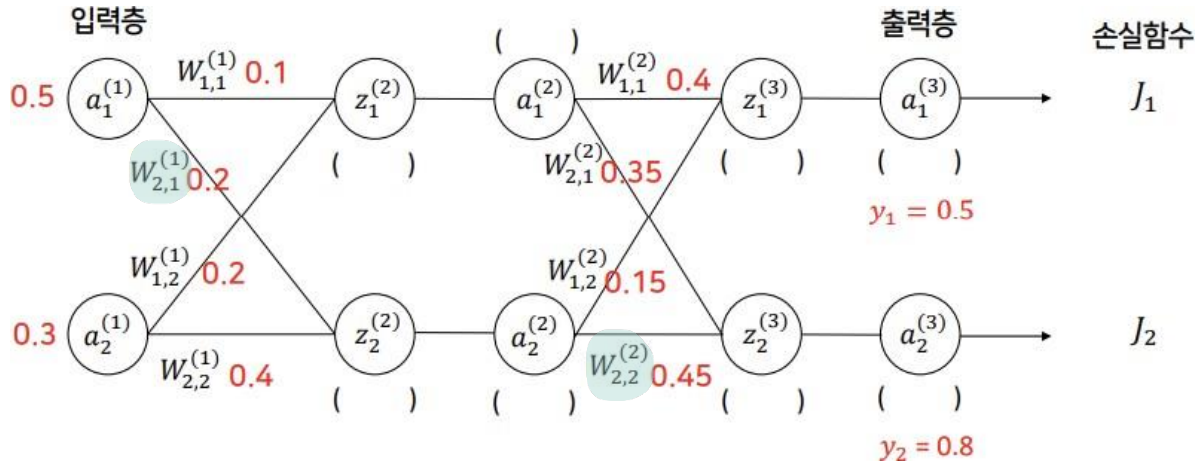
- 3-2. 3-1에서 구한 값을 이용하여 손실함수 J_1 과 J_2 의 값을 구해주세요. (J_1 과 J_2 는 반올림하지 말고 써주세요.)

손실함수 $MSE = \frac{1}{2N} \sum_{i=1}^N (y_i - \hat{y}_i)^2$

$$J_1 = \frac{1}{2} (a_1^{(3)} - y_1)^2 = \frac{1}{2} (0.57 - 0.5)^2 = 0.00245$$

$$J_2 = \frac{1}{2} (a_2^{(3)} - y_2)^2 = \frac{1}{2} (0.61 - 0.8)^2 = 0.01805$$

가중치 업데이트식 $w_j = w_j - \eta \frac{\partial J_{total}}{\partial w_j}$



3-3. 위에서 구한 값을 토대로, BackPropagation이 일어날 때 $w_{2,2}^{(2)}$ 와 $w_{2,1}^{(1)}$ 의 조정된 값을 구해주세요.

단, learning rate는 0.1입니다. (계산 과정에서 소수점 넷째자리에서 반올림하여 셋째자리까지만 써 주시고, 마지막 결과인 $w_{2,1}^{(1)}$ 과 $w_{2,2}^{(2)}$ 의 값만 반올림하지 말고 써주세요.)

$$w_{2,2}^{(2)} = w_{2,2}^{(2)} - \eta \frac{\partial J_{total}}{\partial w_{2,2}^{(2)}}$$

$$\frac{\partial J_{total}}{\partial w_{2,2}^{(2)}} = \frac{\partial J_2}{\partial a_2^{(3)}} \times \frac{\partial a_2^{(3)}}{\partial z_2^{(3)}} \times \frac{\partial z_2^{(3)}}{\partial w_{2,2}^{(2)}} \quad (\because \text{chain rule})$$

$$\therefore \frac{\partial J_2}{\partial a_2^{(3)}} = \frac{1}{2} \frac{\partial}{\partial a_2^{(3)}} (a_2^{(3)} - y_1)^2 = a_2^{(3)} - y_1 = 0.61 - 0.8 = -0.19$$

$$\therefore \frac{\partial a_2^{(3)}}{\partial z_2^{(3)}} = \frac{\partial \phi(z_2^{(3)})}{\partial z_2^{(3)}} = \phi(z_2^{(3)}) (1 - \phi(z_2^{(3)})) = a_2^{(3)} (1 - a_2^{(3)}) = 0.61 \times 0.39 \doteq 0.238$$

$$(\because \frac{d}{dz} \phi(z) = \phi(z) (1 - \phi(z)), \text{ *10117 306})$$

$$\therefore \frac{\partial z_2^{(3)}}{\partial w_{2,2}^{(2)}} = \frac{\partial (w_{2,1}^{(2)} a_1^{(2)} + w_{2,2}^{(2)} a_2^{(2)})}{\partial w_{2,2}^{(2)}} = a_2^{(2)} = 0.55$$

$$\Rightarrow w_{2,2}^{(2)} = w_{2,2}^{(2)} - 0.1 \times (-0.19) \times 0.238 \times 0.55$$

$$= 0.45 + 0.024871 = 0.474871$$

$$\therefore w_{2,2}^{(2)} = 0.474871$$

$$w_{2,1}^{(1)} = w_{2,1}^{(1)} - \eta \frac{\partial J_{total}}{\partial w_{2,1}^{(1)}}, \text{ or equivalently } J_{total} = J_1 + J_2$$

$$\frac{\partial J_{total}}{\partial w_{2,1}^{(1)}} = \frac{\partial J_{total}}{\partial a_2^{(2)}} \times \frac{\partial a_2^{(2)}}{\partial z_2^{(2)}} \times \frac{\partial z_2^{(2)}}{\partial w_{2,1}^{(1)}}$$

$$\begin{aligned} i) \frac{\partial J_{total}}{\partial a_2^{(2)}} &= \frac{\partial J_1}{\partial a_2^{(2)}} + \frac{\partial J_2}{\partial a_2^{(2)}} = \frac{\partial J_1}{\partial z_1^{(3)}} \times \frac{\partial z_1^{(3)}}{\partial a_2^{(2)}} + \frac{\partial J_2}{\partial z_2^{(3)}} \times \frac{\partial z_2^{(3)}}{\partial a_2^{(2)}} \\ &= \delta_1^{(3)} \times \frac{\partial (w_{1,1}^{(2)} a_1^{(1)} + w_{1,2}^{(2)} a_2^{(1)})}{\partial a_2^{(2)}} + \delta_2^{(3)} \times \frac{\partial (w_{2,1}^{(2)} a_1^{(2)} + w_{2,2}^{(2)} a_2^{(2)})}{\partial a_2^{(2)}} \\ &= \delta_1^{(3)} \times w_{1,2}^{(2)} + \delta_2^{(3)} \times w_{2,2}^{(2)} \end{aligned}$$

$$\text{or equivalently, } \delta_1^{(3)} = \frac{\partial J_1}{\partial z_1^{(3)}} = (a_1^{(3)} - y_1) \times a_1^{(3)} (1 - a_1^{(3)}) = 0.07 \times 0.57 \times 0.43 = 0.017$$

$$\delta_2^{(3)} = \frac{\partial J_2}{\partial z_2^{(3)}} = (a_2^{(3)} - y_2) \times a_2^{(3)} (1 - a_2^{(3)}) = 0.19 \times 0.43 \times 0.57 = 0.047$$

$$ii) \frac{\partial a_2^{(2)}}{\partial z_2^{(2)}} = a_2^{(2)} (1 - a_2^{(2)}) = 0.57 \times 0.43 = 0.249$$

$$iii) \frac{\partial z_2^{(2)}}{\partial w_{2,1}^{(1)}} = \frac{\partial (w_{2,1}^{(1)} a_1^{(1)} + w_{2,2}^{(1)} a_2^{(1)})}{\partial w_{2,1}^{(1)}} = a_1^{(1)} = 0.5$$

$$\begin{aligned} \Rightarrow \frac{\partial J_{total}}{\partial w_{2,1}^{(1)}} &= (\delta_1^{(3)} \times w_{1,2}^{(2)} + \delta_2^{(3)} \times w_{2,2}^{(2)}) \times a_2^{(2)} (1 - a_2^{(2)}) \times a_1^{(1)} \\ &= (0.017 \times 0.15 + 0.047 \times 0.45) \times 0.249 \times 0.5 \\ &= 0.00295065 \approx 0.003 \end{aligned}$$

$$\Rightarrow w_{2,1}^{(1)} = w_{2,1}^{(1)} - 0.1 \times 0.003 = 0.2 - 0.0003 = 0.1997$$

$$\therefore w_{2,1}^{(1)} = 1.997$$