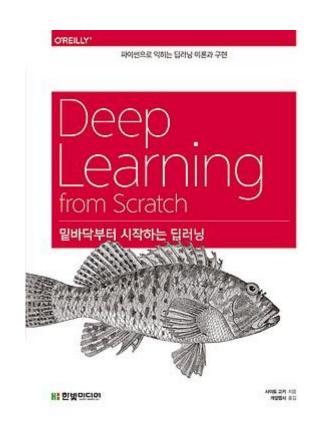
2021.07.28 AAI Lab. 세미나

김민준

#### 참고 자료

• 유튜브 NeoWizard 머신러닝 강의

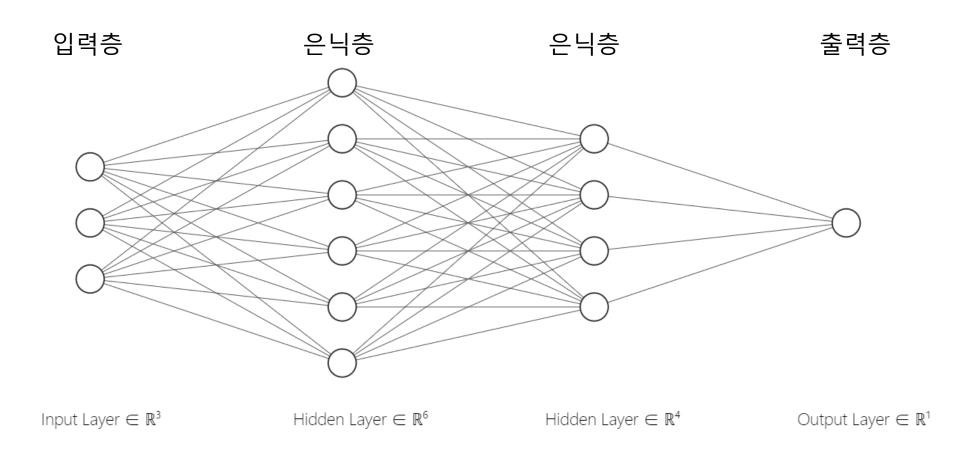
• 밑바닥부터 시작하는 딥러닝 – 사이토 고키



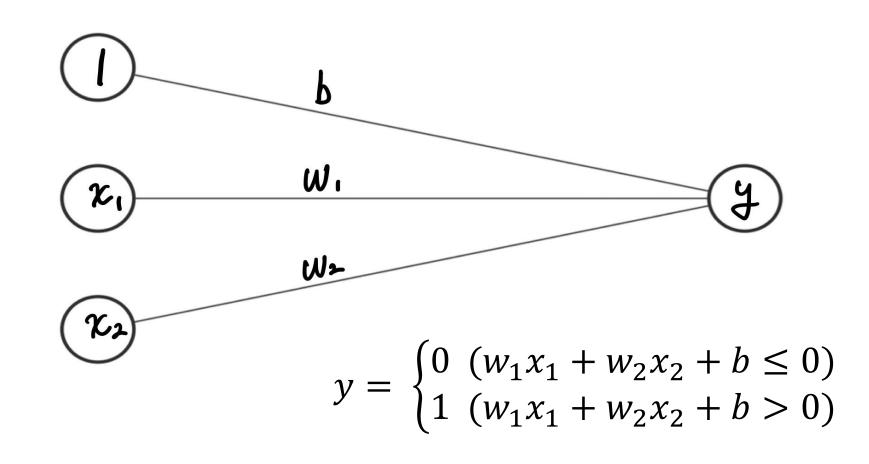
#### 신경망

• 가중치 매개변수의 적절한 값을 데이터로부터 자동으로 학습

# 신경망

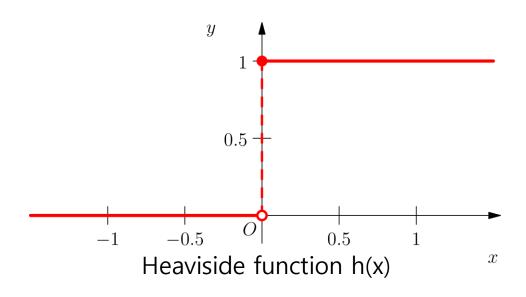


### 퍼셉트론



#### 퍼셉트론

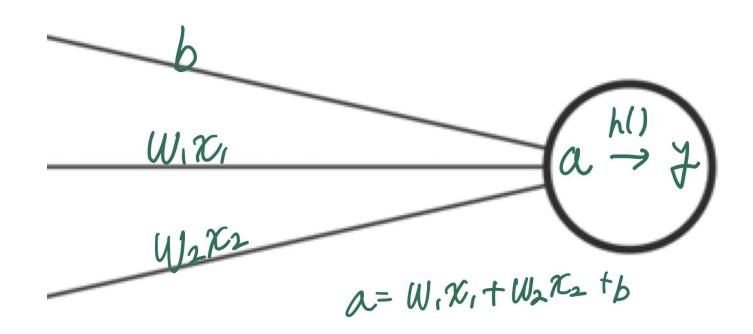
$$h(x) = \begin{cases} 0 & (x \le 0) \\ 1 & (x > 0) \end{cases}$$



$$y = \begin{cases} 0 & (w_1 x_1 + w_2 x_2 + b \le 0) \\ 1 & (w_1 x_1 + w_2 x_2 + b > 0) \end{cases}$$

$$\rightarrow y = h(w_1x_1 + w_2x_2 + b)$$

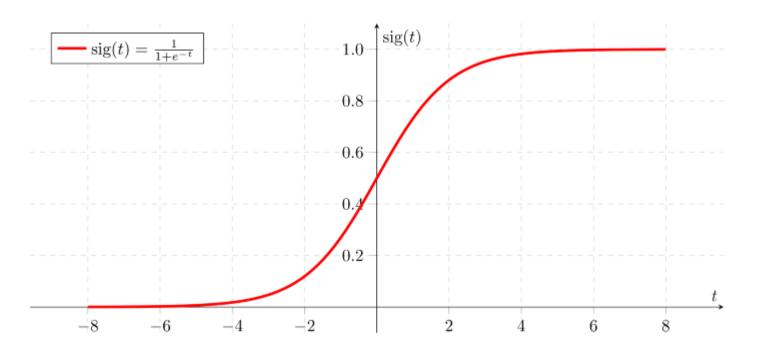
## 퍼셉트론



Sigmoid

• 
$$sigmoid(x) = \frac{1}{1+e^{-x}}$$

•  $0 \le sigmoid(x) \le 1$ 

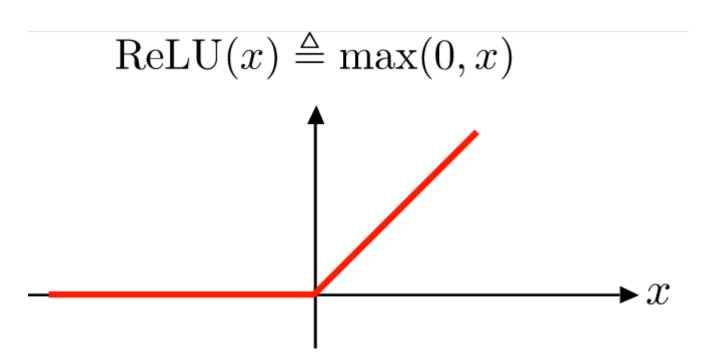


계단함수와 유사한 모양

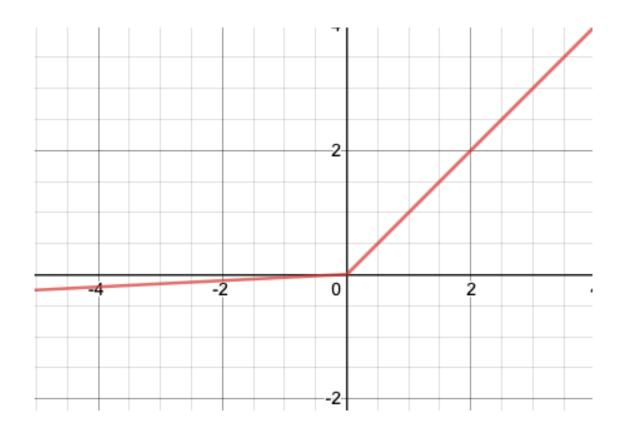
sigmoid'(x) = sigmoid(x)(1 - sigmoid(x))

ReLU

Solve vanishing gradient



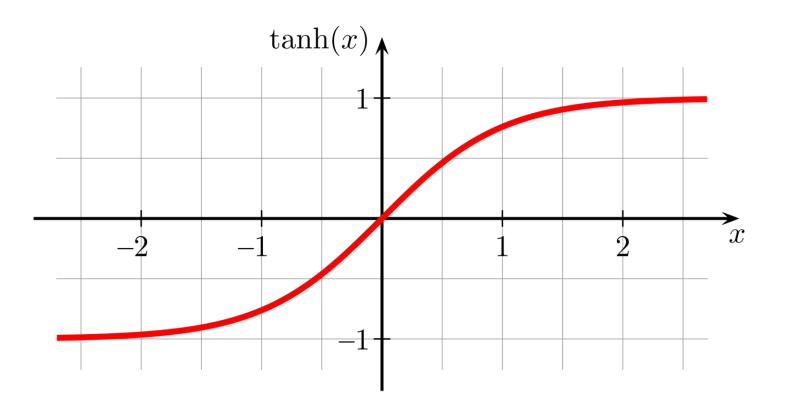
- Leaky ReLU
- Solve vanishing gradient



• Tanh

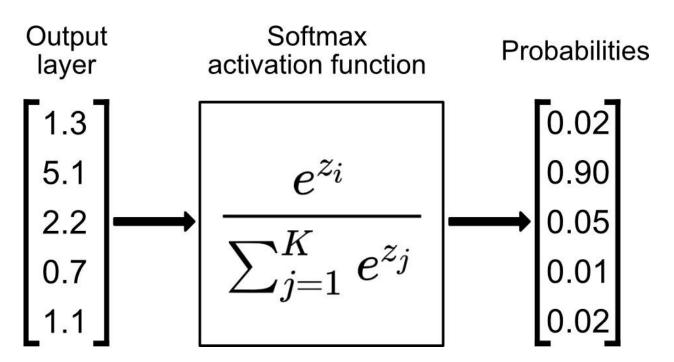
• 
$$tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

•  $-1 \le \tanh(x) \le 1$ 



Softmax

• 
$$softmax(x) = \frac{e^{x_i}}{\sum e^{x_i}}$$



• 출력층에서 확률 값을 표현하기 위해 사용

# 활성화 함수, why?

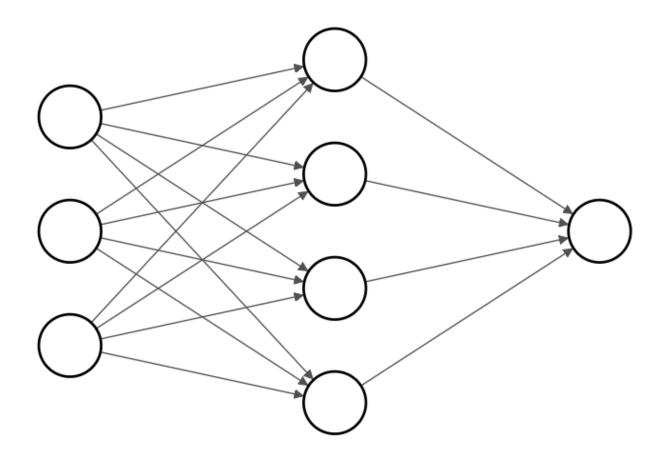
• Linear → non-Linear

• Deep neural network에서 모든 정보를 전달하지 않고 선별적으로 전달 (ReLU의 x<0 부분)

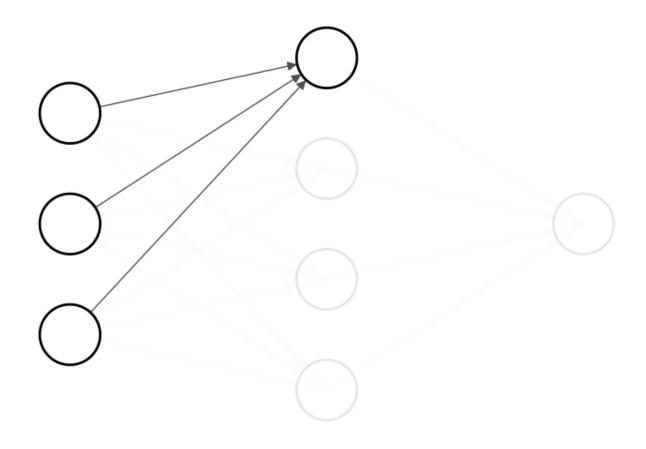
• Vanishing gradient, gradient overflow 방지

• 확률 값 출력

- 신경망 구조를 바탕으로 입력층, 은닉층, 출력층으로 구성
- 은닉층이 1개 이상일 때 deep neural network
- 은닉층이 깊을수록(많을수록) 정확도가 높아짐



Input Layer ∈ ℝ³ Hidden Layer ∈ ℝ⁴ Output Layer ∈ ℝ¹



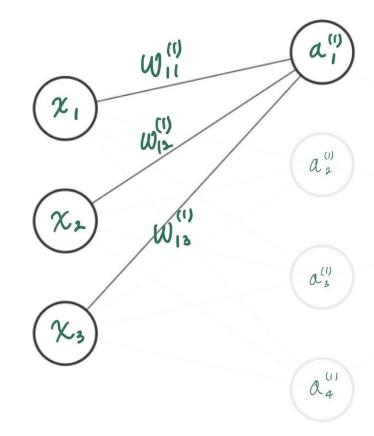
Input Layer  $\in \mathbb{R}^3$  Hidden Layer  $\in \mathbb{R}^4$  Output Layer  $\in \mathbb{R}^1$ 

$$X = (x_1, x_2, x_3)$$

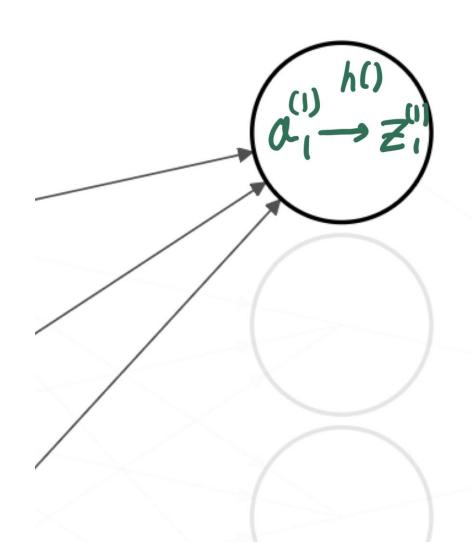
$$W^{(1)} = \begin{pmatrix} w_{11}^{(1)} & w_{21}^{(1)} & w_{31}^{(1)} & w_{41}^{(1)} \\ w_{12}^{(1)} & w_{22}^{(1)} & w_{32}^{(1)} & w_{42}^{(1)} \\ w_{13}^{(1)} & w_{23}^{(1)} & w_{33}^{(1)} & w_{43}^{(1)} \end{pmatrix}$$

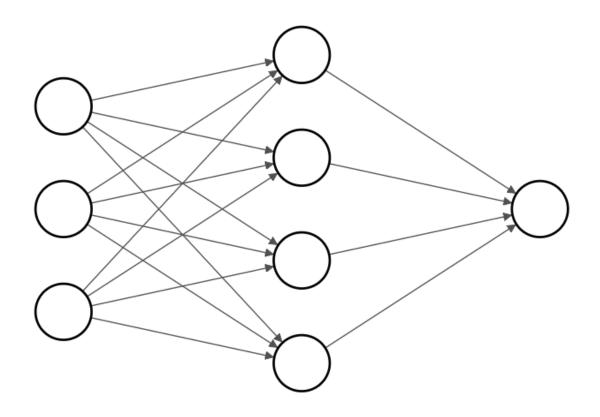
$$A^{(1)}=(a_1^{(1)},a_2^{(1)},a_3^{(1)},\ a_4^{(1)})$$

$$XW = A$$
$$M_{13} \times M_{34} = M_{14}$$



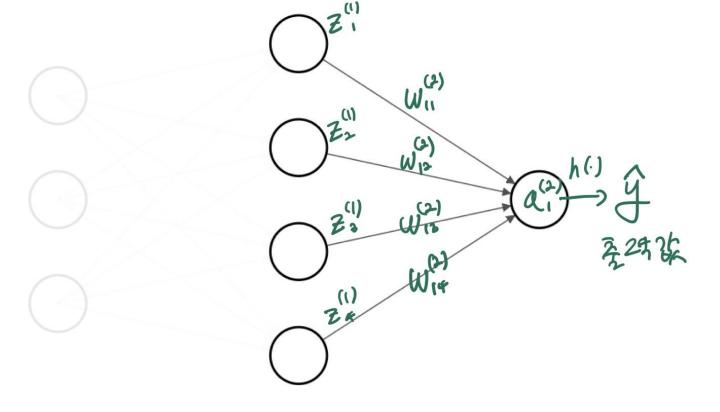
$$a_1^{(1)} = x_1 w_{11}^{(1)} + x_2 w_{12}^{(1)} + x_3 w_{13}^{(1)}$$





Input Layer  $\in \mathbb{R}^3$  Hidden Layer  $\in \mathbb{R}^4$  Output Layer  $\in \mathbb{R}^1$ 

$$a_1^{(2)} = z_1^{(1)} w_{11}^{(2)} + z_2^{(1)} w_{12}^{(2)} + z_3^{(1)} w_{13}^{(2)} + z_4^{(1)} w_{14}^{(2)}$$



Input Layer € R³

Hidden Layer ∈ R<sup>4</sup> Output Layer ∈ R<sup>1</sup>

- 가중치 값들은 층과 층 사이의 모든 노드에 초기화 되어있음.
- Random initialization
- Xavier initialization
- He initialization

• 오차가 최소값이 될 때의 가중치가 최적의 가중치

# 감사합니다