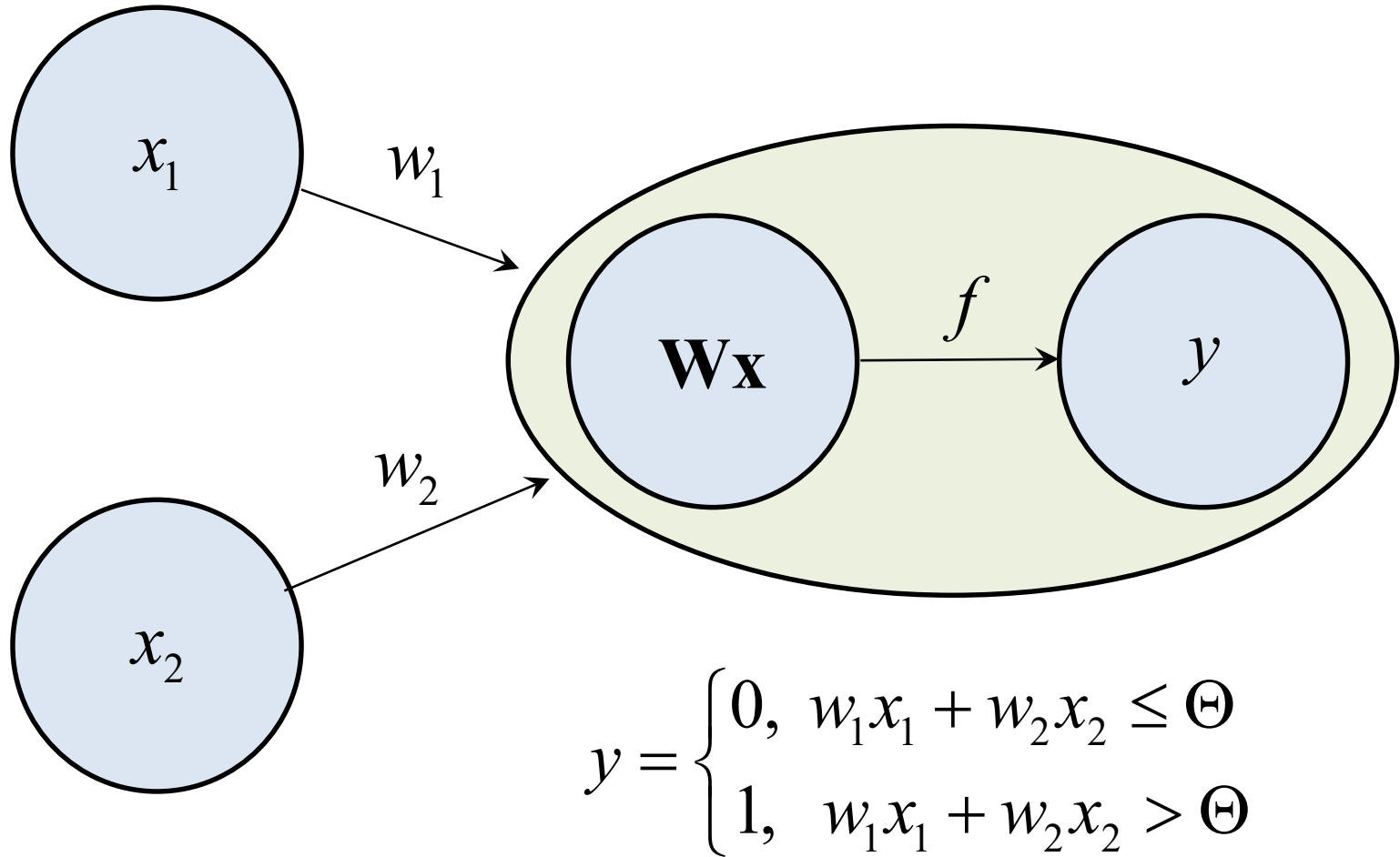
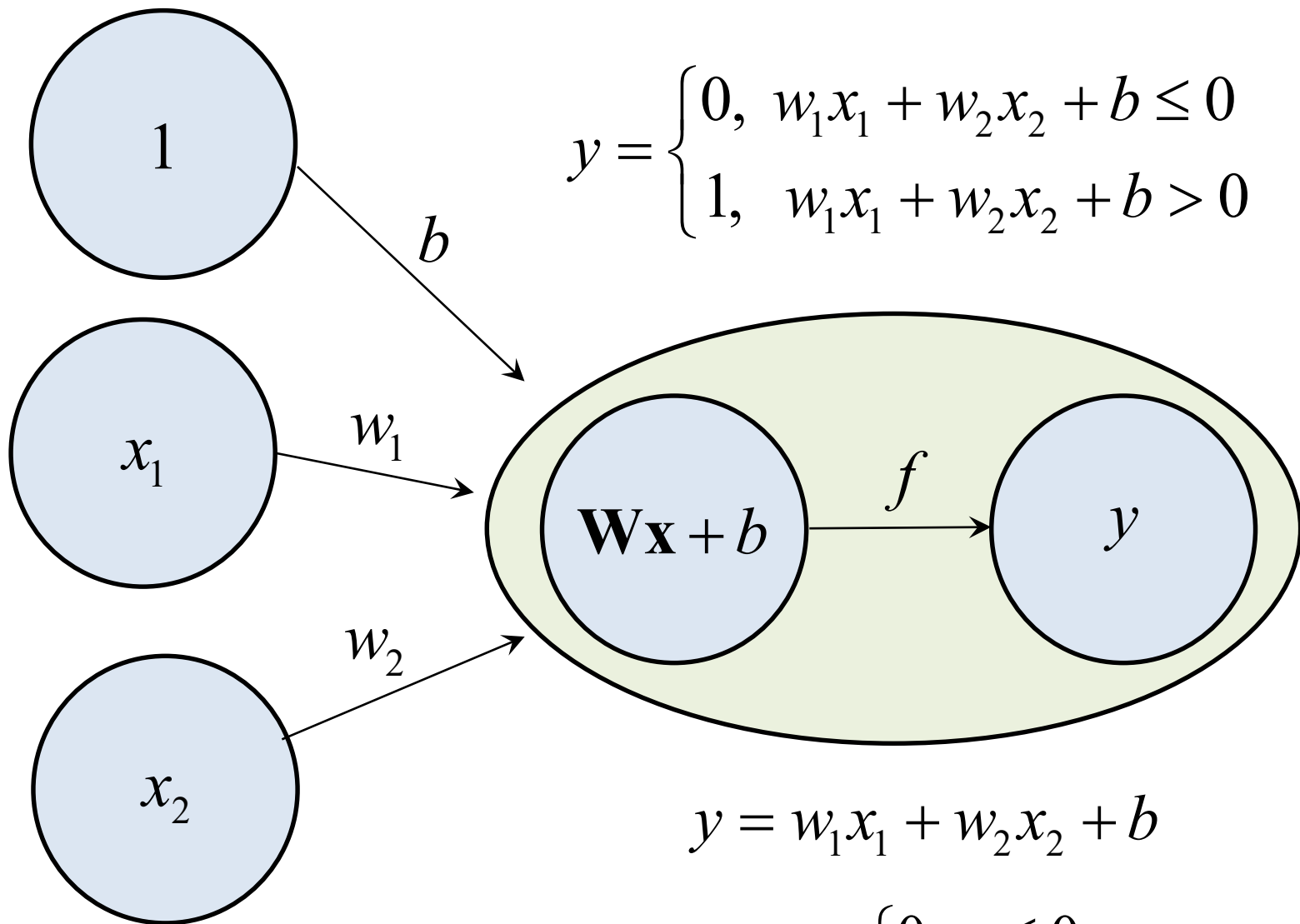


# Neural Network

## 3. Neural Network

# Perceptron





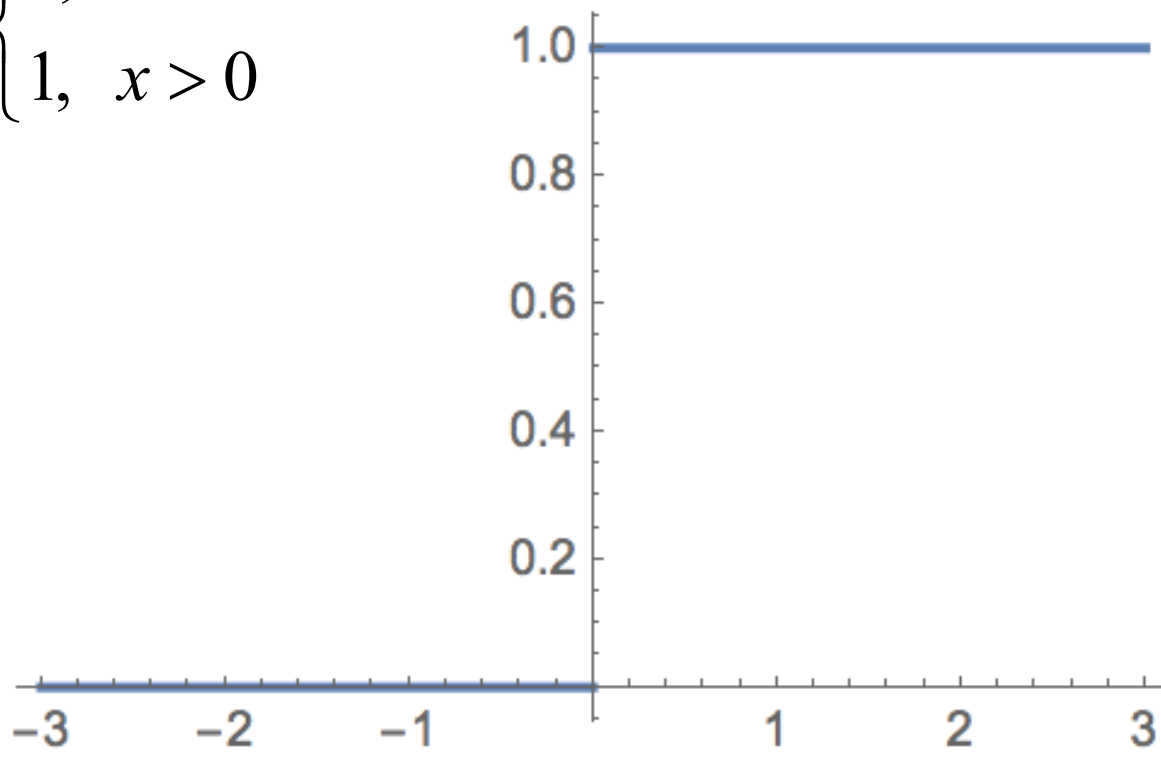
$$y = \begin{cases} 0, & w_1x_1 + w_2x_2 + b \leq 0 \\ 1, & w_1x_1 + w_2x_2 + b > 0 \end{cases}$$

$$y = w_1x_1 + w_2x_2 + b$$

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

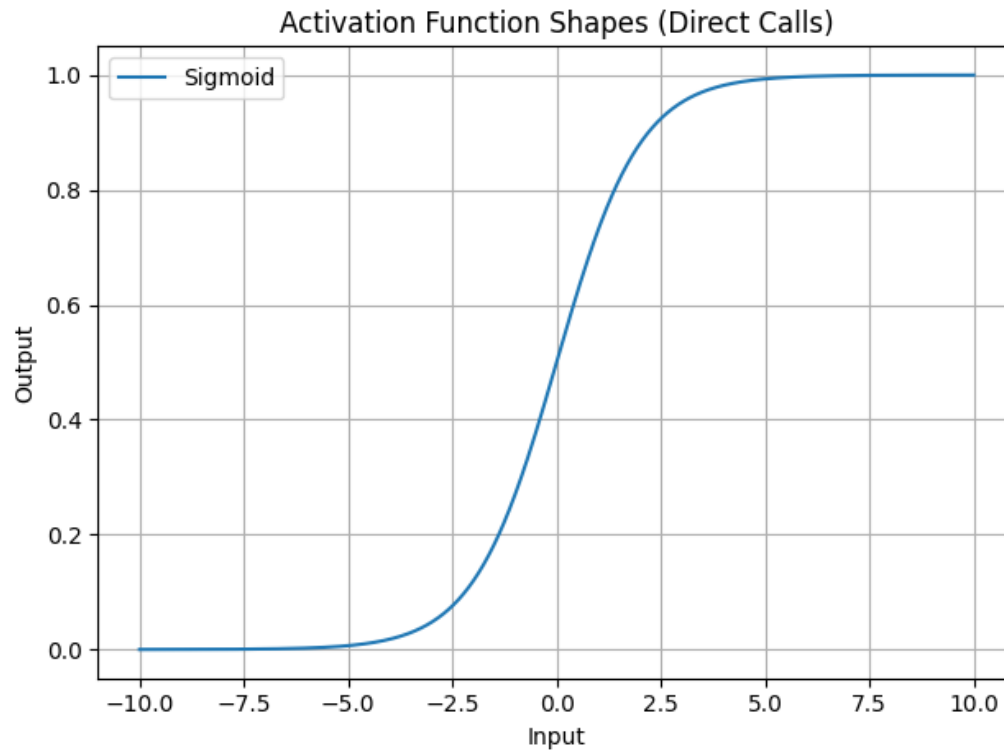
# Activation Function

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



# Sigmoid

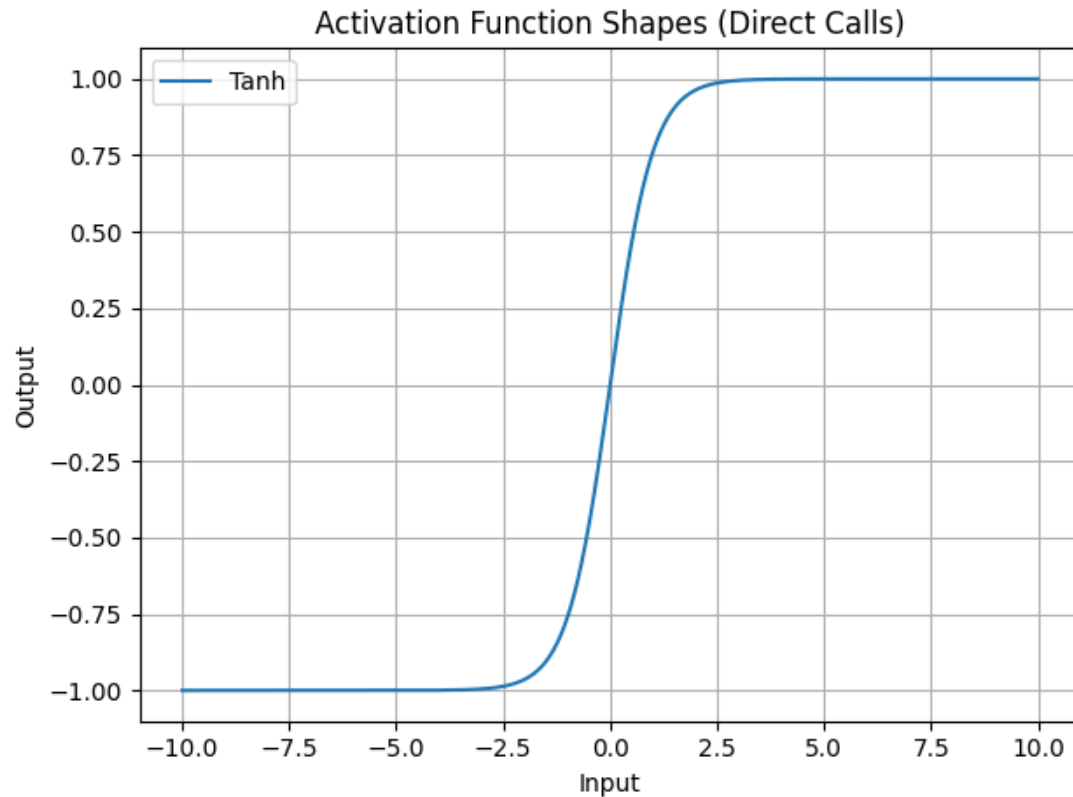
$$f(x) = \frac{1}{1 + e^{-x}}$$



# Tanh

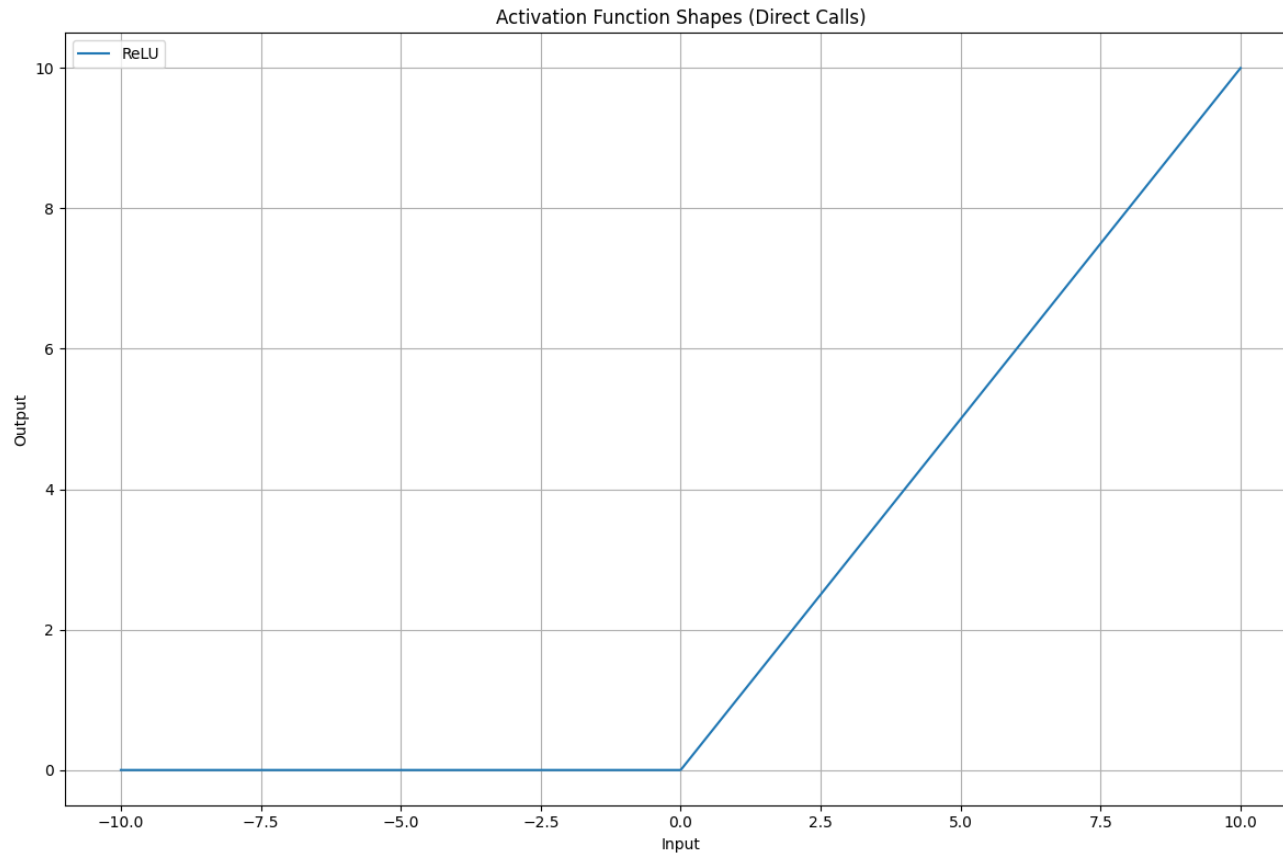
$$f(x) = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

$$\text{sigmoid} = \frac{1}{2} \left\{ 1 + \tanh\left(\frac{x}{2}\right) \right\}$$



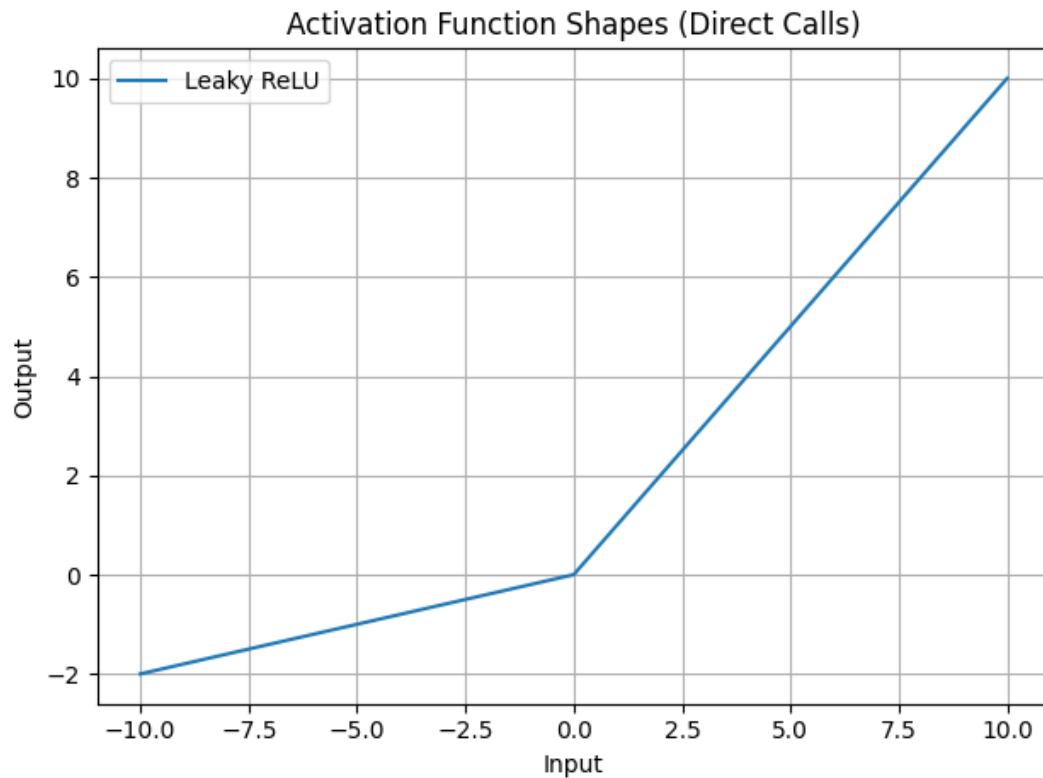
# ReLU (Rectified Linear Unit)

$$f(x) = \max(0, x)$$



# Leaky ReLU

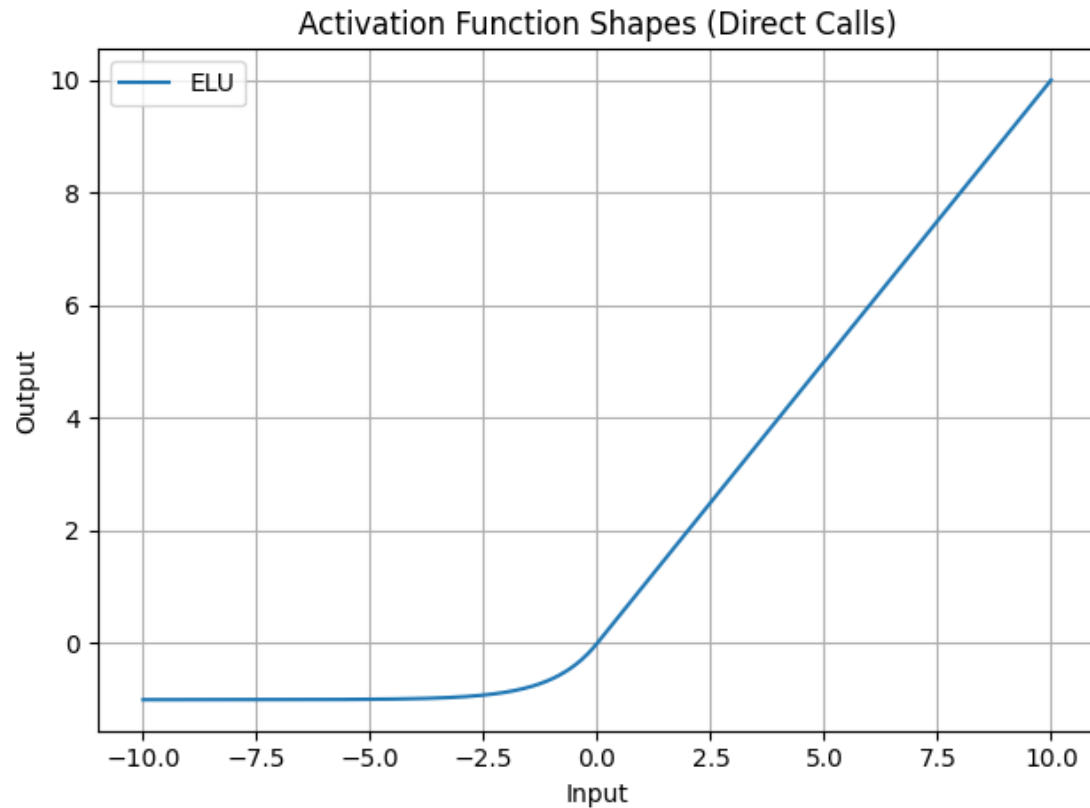
$$f(x) = \begin{cases} x, & x \geq 0 \\ \alpha x, & x < 0 \end{cases}$$





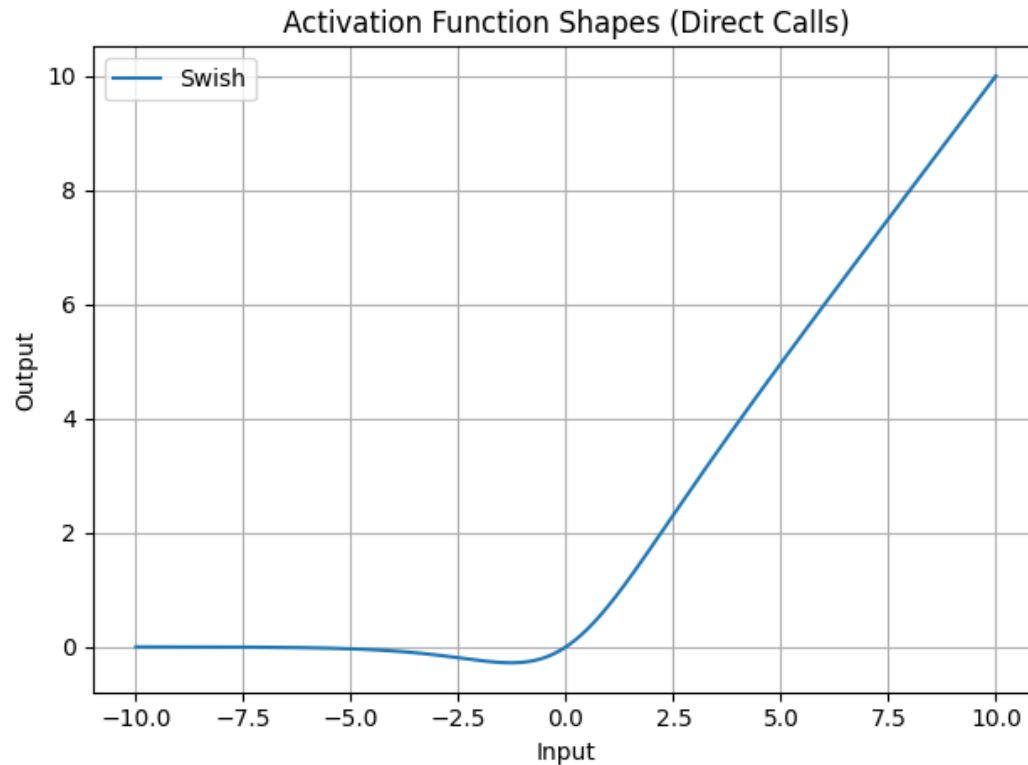
# ELU

$$f(x) = \begin{cases} x, & x \geq 0 \\ \alpha(e^x - 1), & x < 0 \end{cases}$$



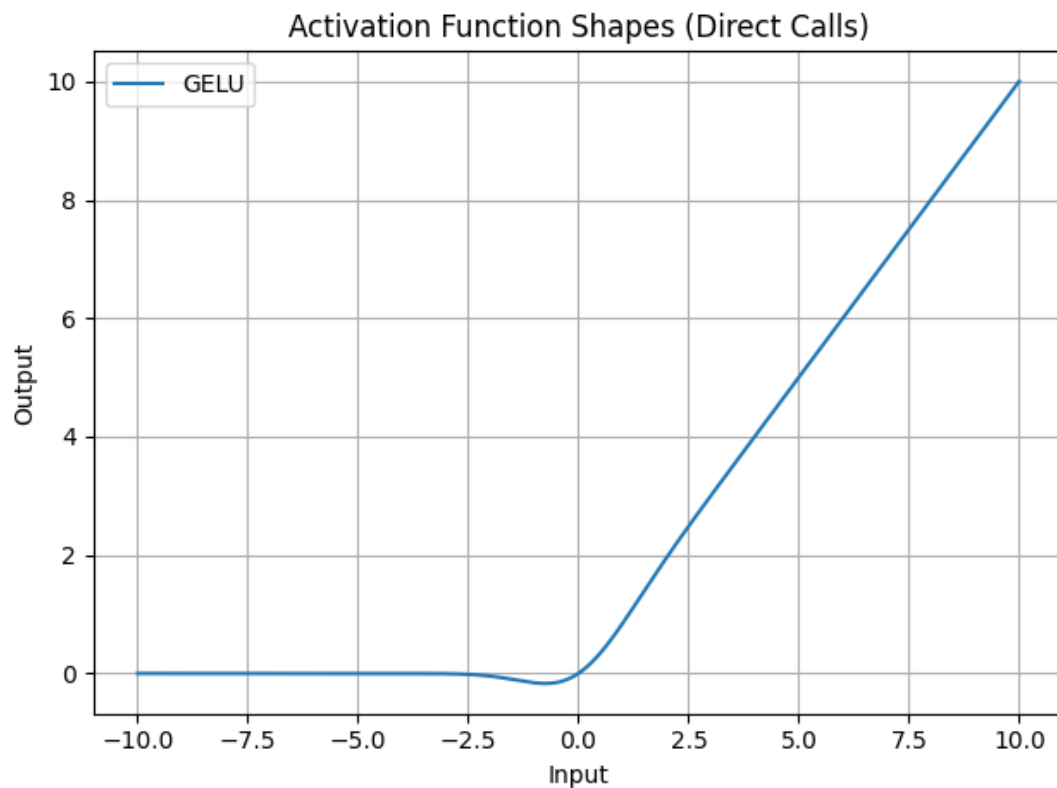
# SWISH

$$f(x) = x \cdot \text{sigmoid}(x)$$

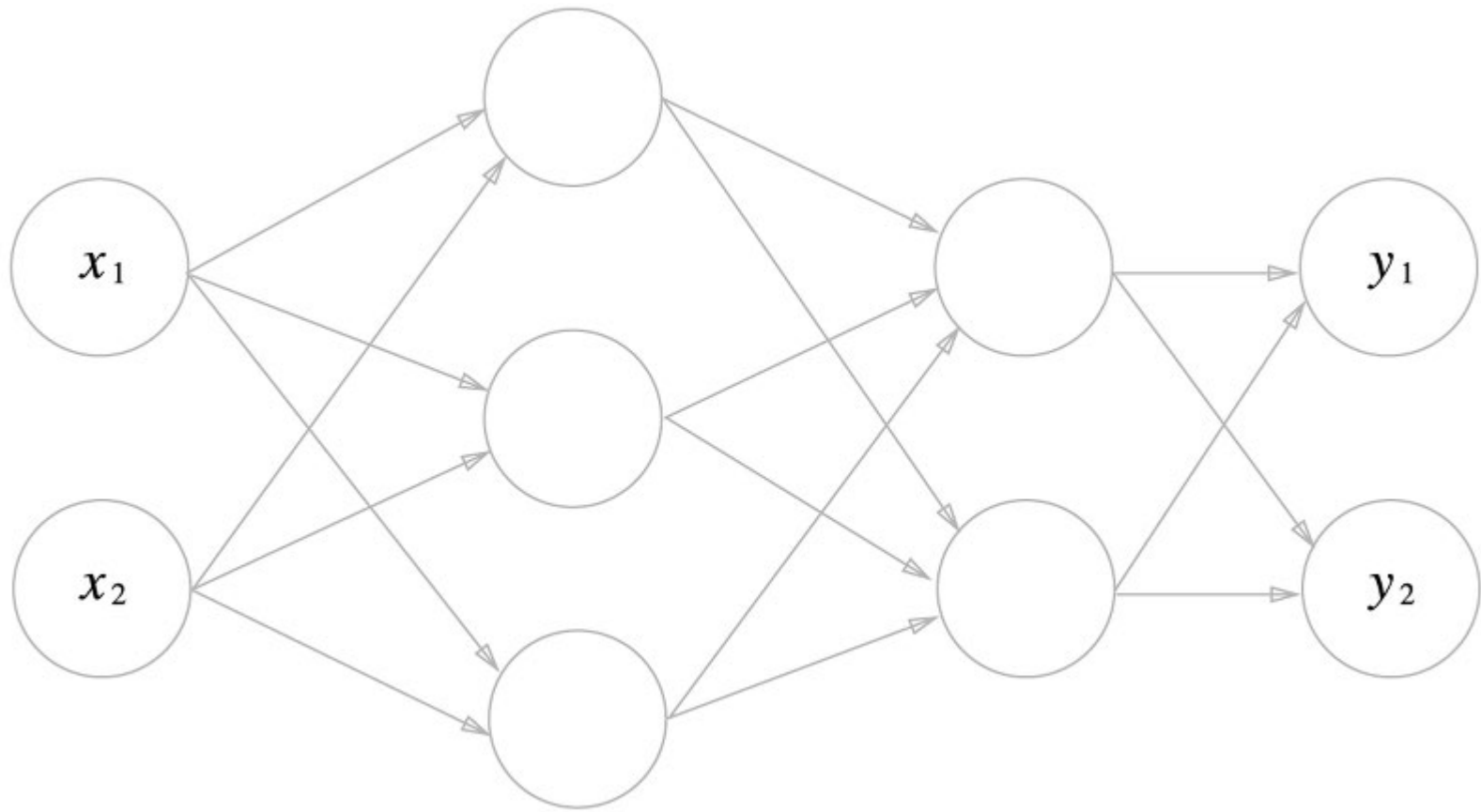


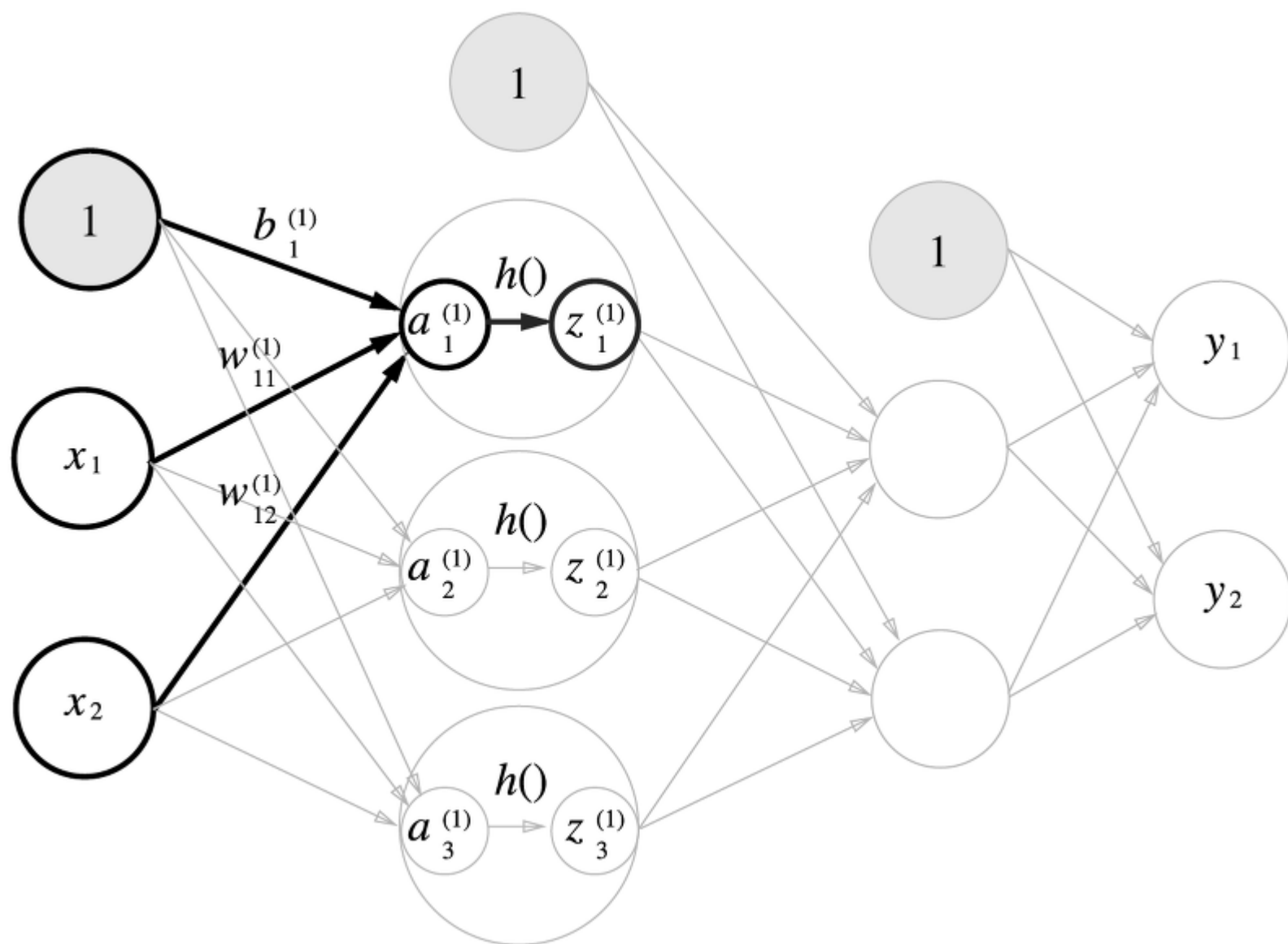
# GELU (Gaussian Error LU)

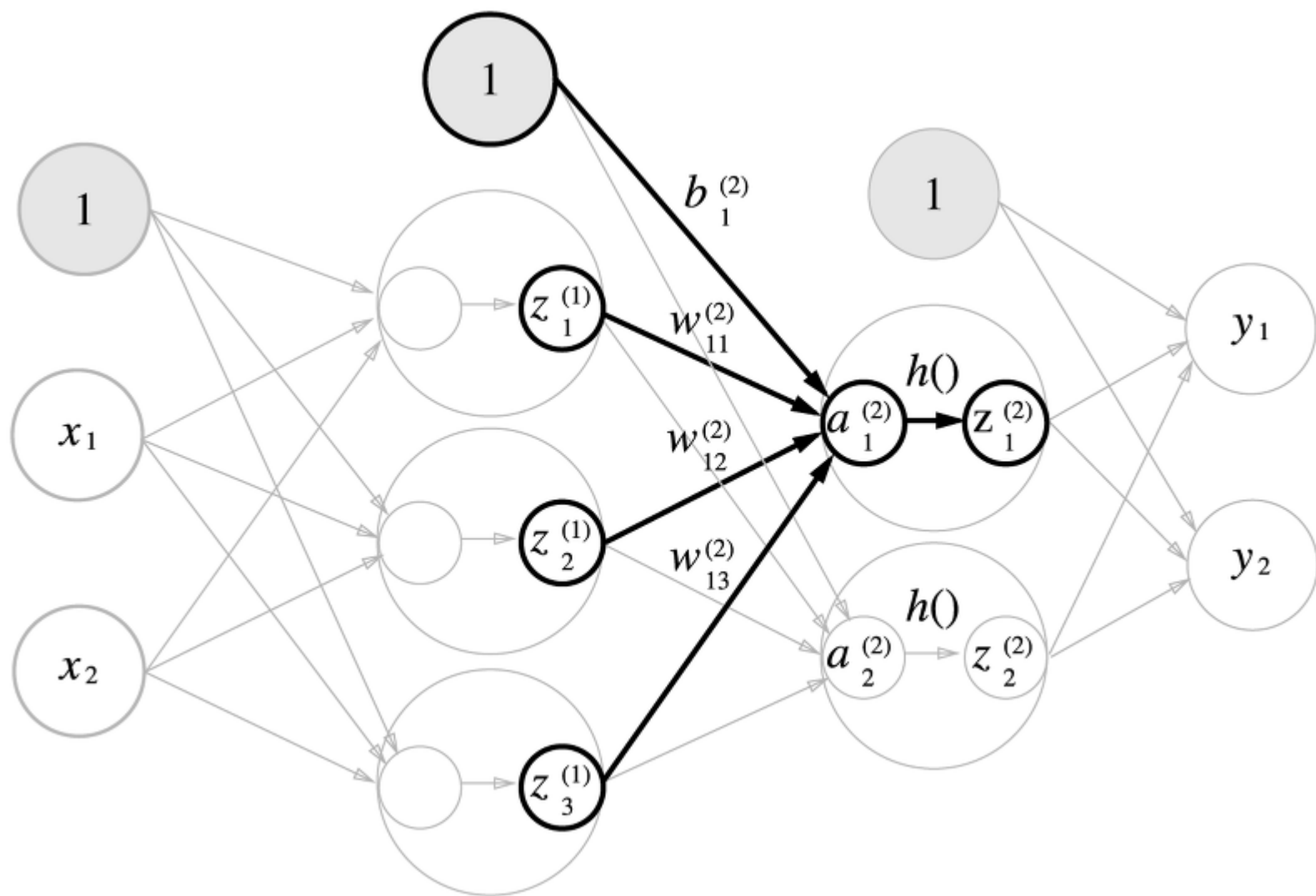
$$f(x) = x \cdot \Phi(x)$$

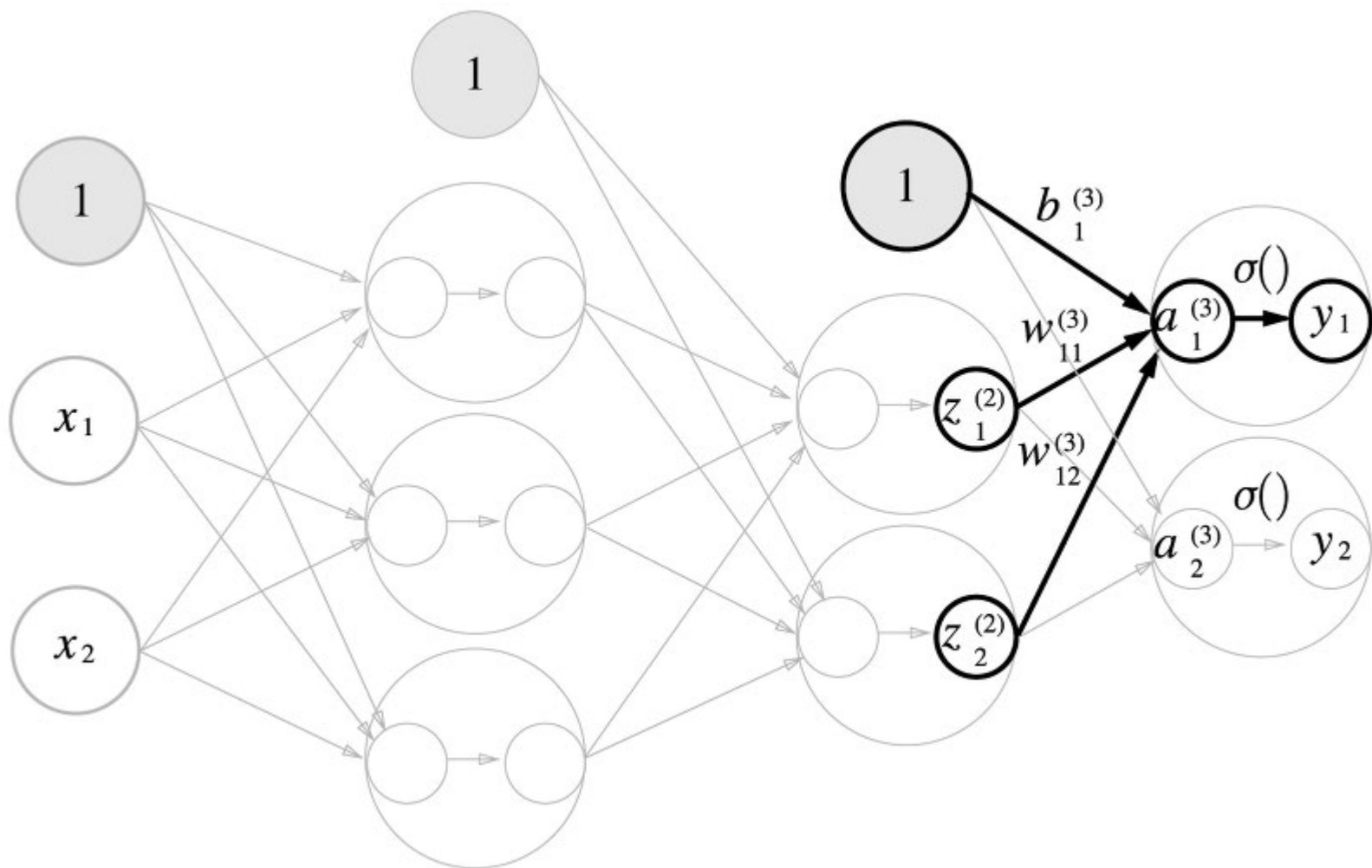


# 3층 신경망 구조









# Matrix

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} = (a_{ij})$$

$m \times n$  matrix



# Operation

If  $\mathbf{A} = (a_{ij})$  and  $\mathbf{B} = (b_{ij})$  then

$$\mathbf{A} + \mathbf{B} = (a_{ij} + b_{ij}) \qquad \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} + \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} 6 & 8 \\ 10 & 12 \end{pmatrix}$$

$$\mathbf{A} - \mathbf{B} = (a_{ij} - b_{ij}) \qquad \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} - \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} -4 & -4 \\ -4 & -4 \end{pmatrix}$$

$$c\mathbf{A} = (ca_{ij}) \qquad 2 \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} = \begin{pmatrix} 2 & 4 \\ 6 & 8 \end{pmatrix}$$

# Product

$$\mathbf{A} = (a_{ij}) \quad m \times p \text{ matrix}$$

$$\mathbf{B} = (b_{ij}) \quad p \times n \text{ matrix}$$

$$\mathbf{AB} = (c_{ij}) \quad m \times n \text{ matrix}$$

$$c_{ij} = \sum_{k=1}^p a_{ik} b_{kj}$$

# Product

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} =$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} =$$

$$\mathbf{AB} \neq \mathbf{BA}$$

# Transpose

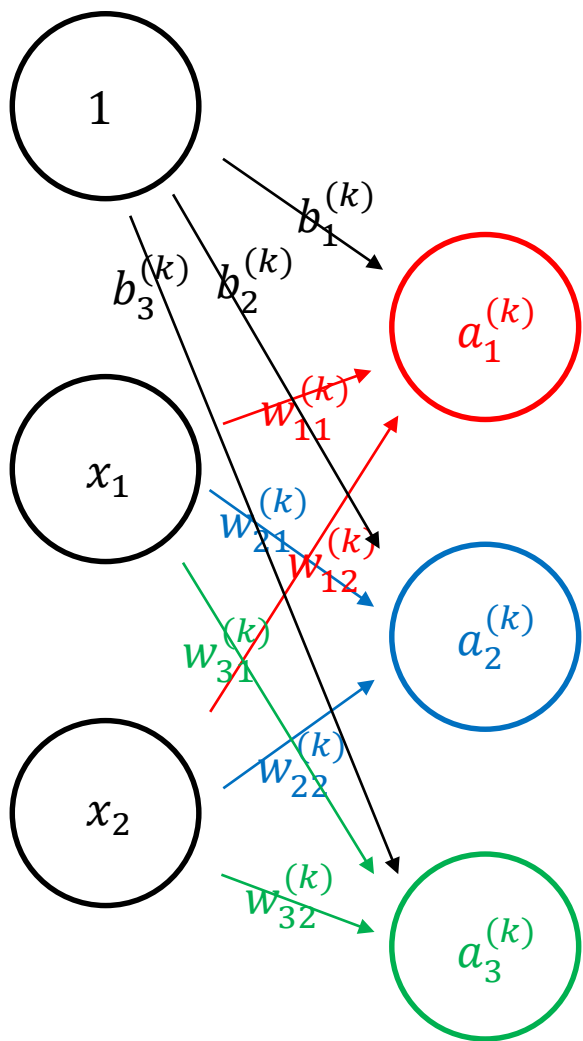
$$\mathbf{A}^T = (a_{ji})$$

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}^T = \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}^T = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}$$

$$(\mathbf{A}^T)^T = \mathbf{A}$$

$$(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$$



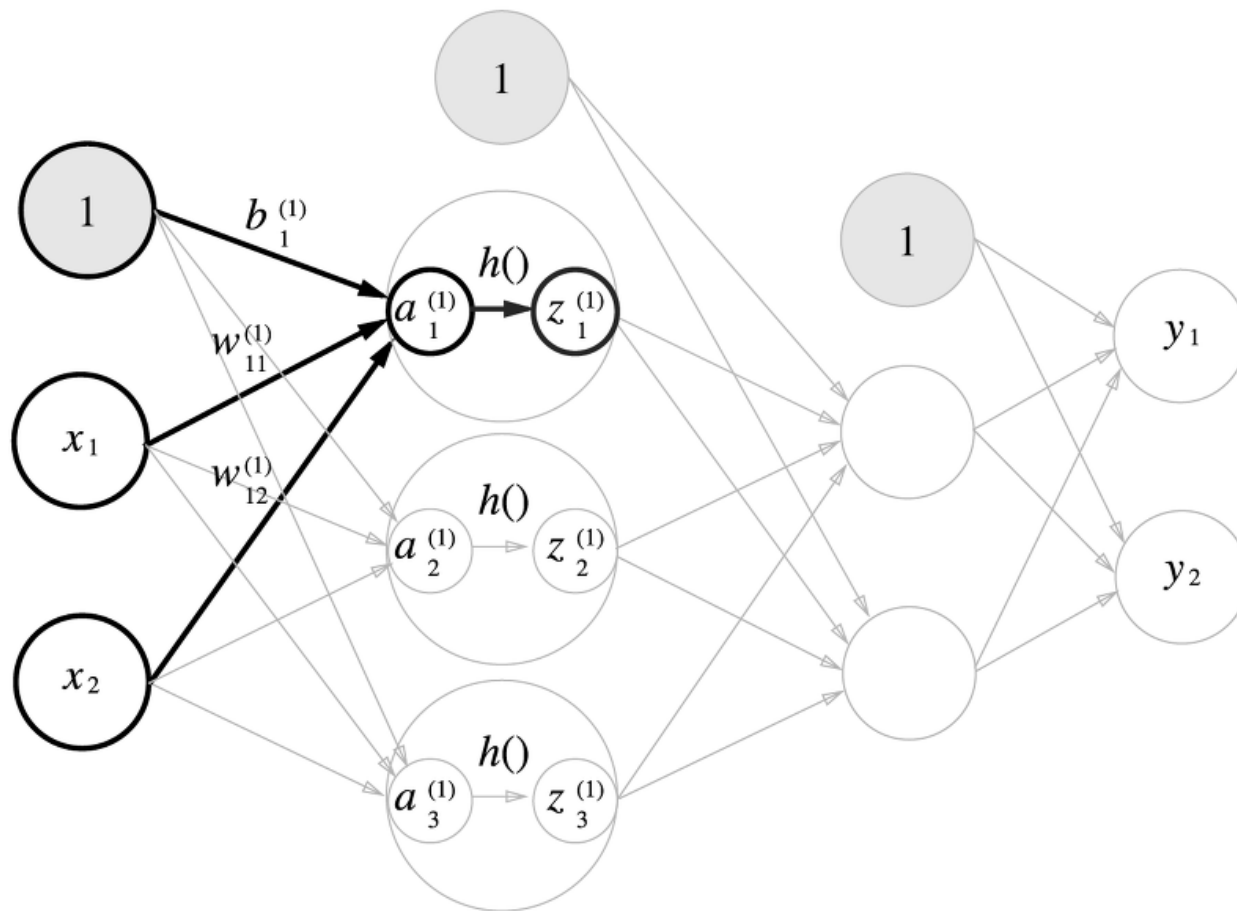
$$\mathbf{W}_{\text{out, in}}^{(\text{layer})}$$

$$a_1^{(k)} = w_{11}^{(k)} x_1 + w_{12}^{(k)} x_2 + b_1^{(k)}$$

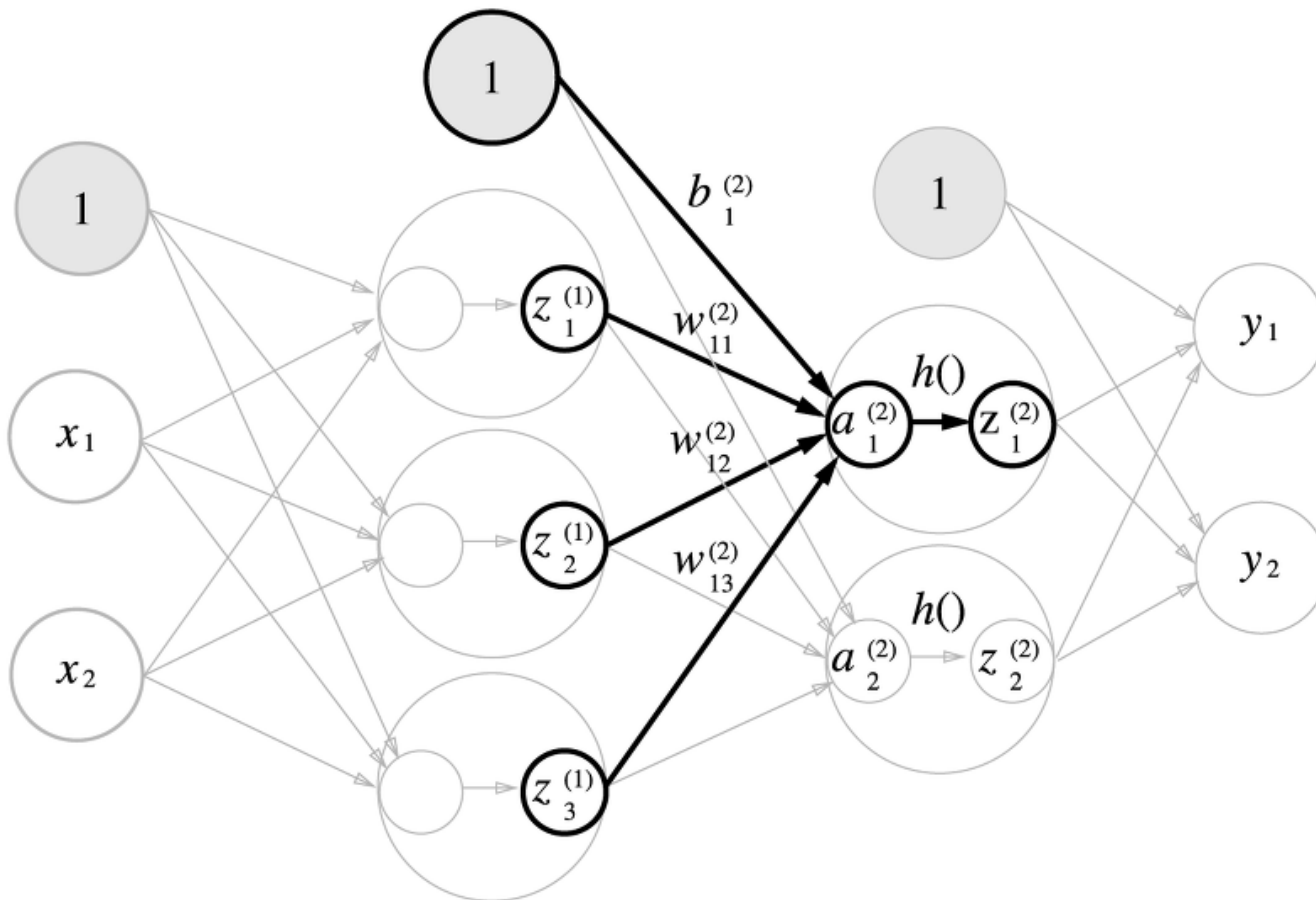
$$a_2^{(k)} = w_{21}^{(k)} x_1 + w_{22}^{(k)} x_2 + b_2^{(k)}$$

$$a_3^{(k)} = w_{31}^{(k)} x_1 + w_{32}^{(k)} x_2 + b_3^{(k)}$$

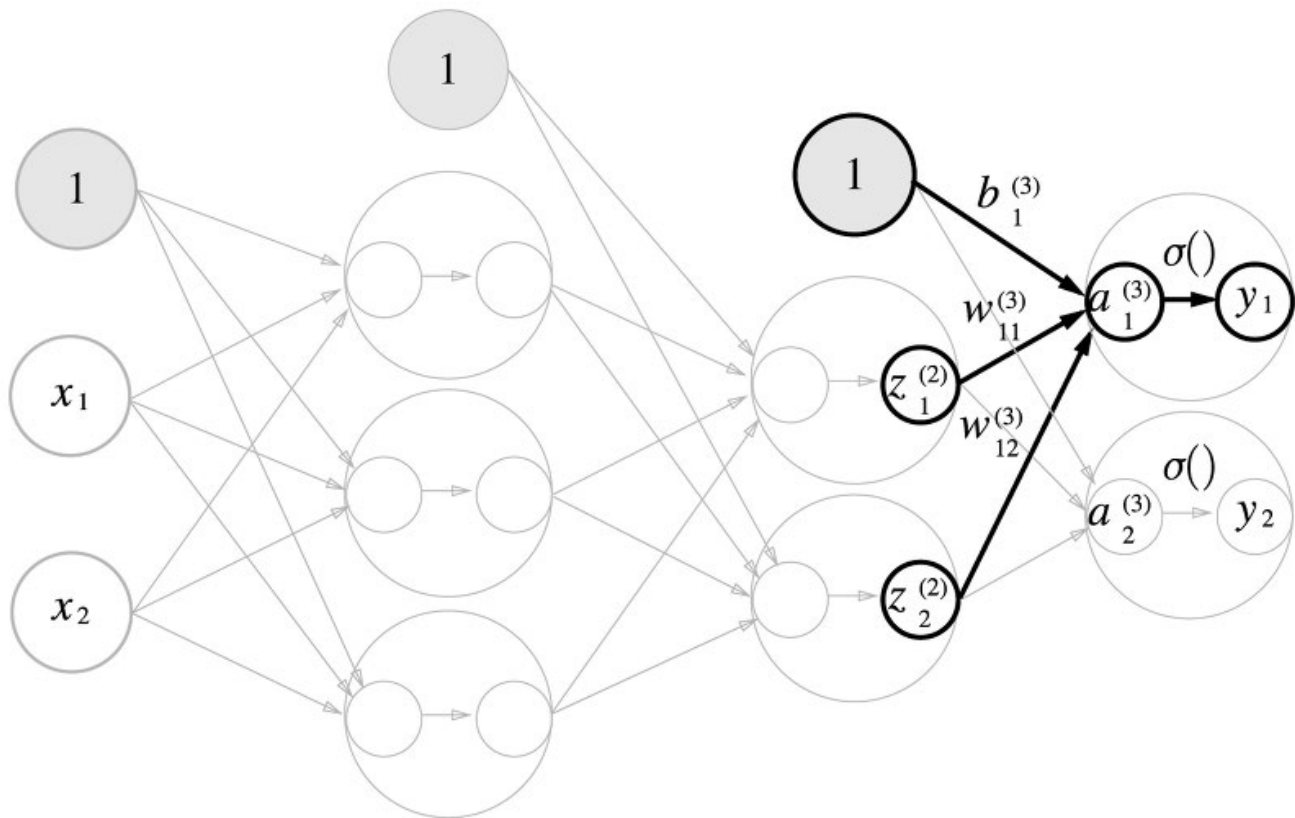
$$\begin{pmatrix} a_1^{(k)} \\ a_2^{(k)} \\ a_3^{(k)} \end{pmatrix} = \begin{pmatrix} w_{11}^{(k)} & w_{12}^{(k)} \\ w_{21}^{(k)} & w_{22}^{(k)} \\ w_{31}^{(k)} & w_{32}^{(k)} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} b_1^{(k)} \\ b_2^{(k)} \\ b_3^{(k)} \end{pmatrix}$$



$$\begin{pmatrix} a_1^{(1)} \\ a_2^{(1)} \\ a_3^{(1)} \end{pmatrix} = \begin{pmatrix} w_{11}^{(1)} & w_{12}^{(1)} \\ w_{21}^{(1)} & w_{22}^{(1)} \\ w_{31}^{(1)} & w_{32}^{(1)} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} b_1^{(1)} \\ b_2^{(1)} \\ b_3^{(1)} \end{pmatrix}$$



$$\begin{pmatrix} a_1^{(2)} \\ a_2^{(2)} \end{pmatrix} = \begin{pmatrix} w_{11}^{(2)} & w_{12}^{(2)} & w_{13}^{(2)} \\ w_{21}^{(2)} & w_{22}^{(2)} & w_{23}^{(2)} \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \\ z_3 \end{pmatrix} + \begin{pmatrix} b_1^{(2)} \\ b_2^{(2)} \end{pmatrix}$$



$$\begin{pmatrix} a_1^{(3)} \\ a_2^{(3)} \end{pmatrix} = \begin{pmatrix} w_{11}^{(3)} & w_{12}^{(3)} \\ w_{21}^{(3)} & w_{22}^{(3)} \end{pmatrix} \begin{pmatrix} z_1^{(2)} \\ z_2^{(2)} \end{pmatrix} + \begin{pmatrix} b_1^{(3)} \\ b_2^{(3)} \end{pmatrix}$$



# Softmax

$(p_1, p_2, p_3, \dots, p_n)$  : Probability vector

$$p_i \geq 0 \quad \sum p_i = 1$$

$$(a_1, a_2, a_3, \dots, a_n)$$

$$(0, -x)$$

$$(e^{a_1}, e^{a_2}, e^{a_3}, \dots, e^{a_n})$$

$$(1, e^{-x})$$

$$\left( \frac{e^{a_1}}{\sum e^{a_i}}, \frac{e^{a_2}}{\sum e^{a_i}}, \frac{e^{a_3}}{\sum e^{a_i}}, \dots, \frac{e^{a_n}}{\sum e^{a_i}} \right)$$

$$\left( \frac{1}{1 + e^{-x}}, \frac{e^{-x}}{1 + e^{-x}} \right)$$

# Softmax

$$(0, -x)$$

$$(0, x)$$

$$(1, e^{-x})$$

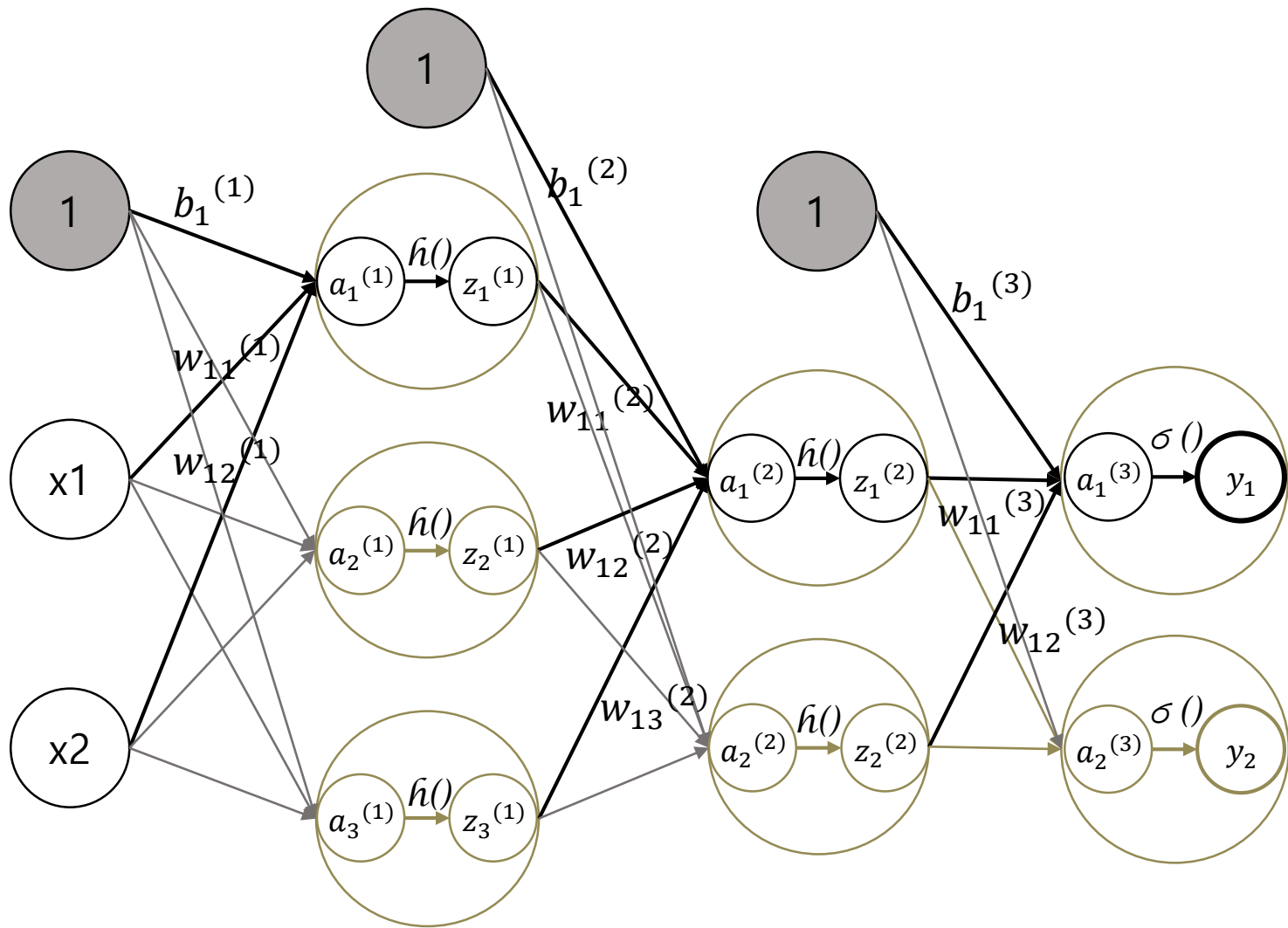
$$(1, e^x)$$

$$\left(\frac{1}{1+e^{-x}}, \frac{e^{-x}}{1+e^{-x}}\right)$$

$$\left(\frac{1}{1+e^x}, \frac{e^x}{1+e^x}\right)$$

$$\text{softmax}(a_1, a_2, a_3, \dots, a_n)$$

$$= \text{softmax}(a_1 + c, a_2 + c, a_3 + c, \dots, a_n + c)$$



Affine  $\rightarrow$  Sigmoid  $\rightarrow$  Affine  $\rightarrow$  Sigmoid  $\rightarrow$  Affine  $\rightarrow$  Softmax

# MNIST



0부터 9까지의 손 글씨 이미지로 구성

훈련 데이터가 6만장, 테스트 데이터가 1만장

각 데이터는 이미지와 라벨로 이루어짐

각 이미지는 28x28 해상도의 흑백 사진

각 픽셀은 0에서 255로 밝기 표현

Question?

## 자료 출처

Deep learning from scratch, 한빛미디어, 사이토고키

[https://github.com/youbeebee/deeplearning\\_from\\_scratch](https://github.com/youbeebee/deeplearning_from_scratch)