AI and Deep Learning

3. Neuron and Learning

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하지만,
연결만 되었다고 되나?
그럼, 어떻게 가능한가?

학습(Learning)

Agenda

- Artificial Intelligence
- Brain and Neurons
- Learning
- Regression
- Deep Neural Networks
- CNN
- RNN
- Unsupervised Learning
- Reinforcement Learning
- AI Applications

Supervised Learning

Learning occurs…

- while experiencing something
- the strength of connection between neurons is properly changed

신기하게도,

아기는 무엇인가를 경험할 때마다

뉴런 사이의 연결이

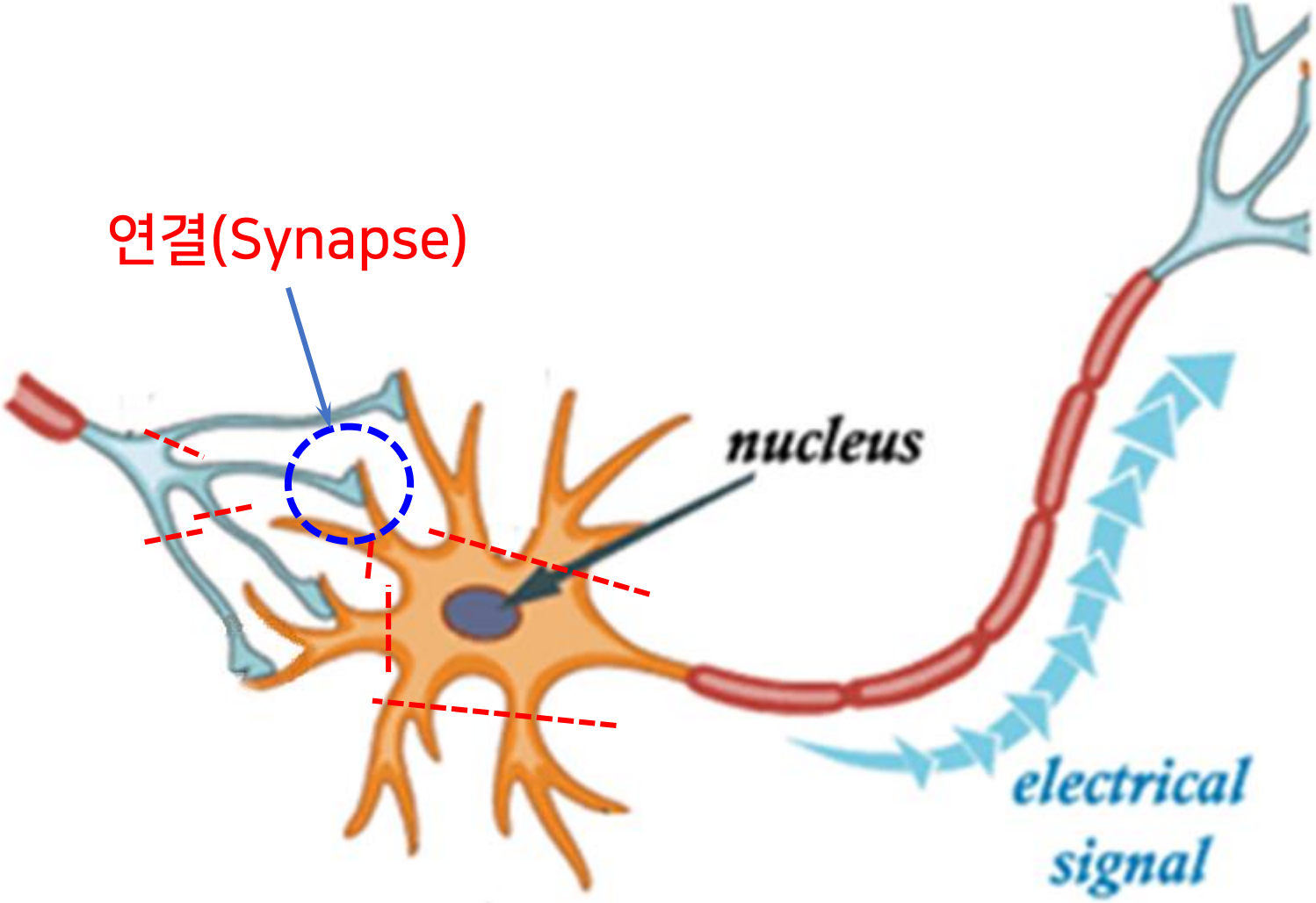
‘자동으로’ 조정된다.

이것이 학습

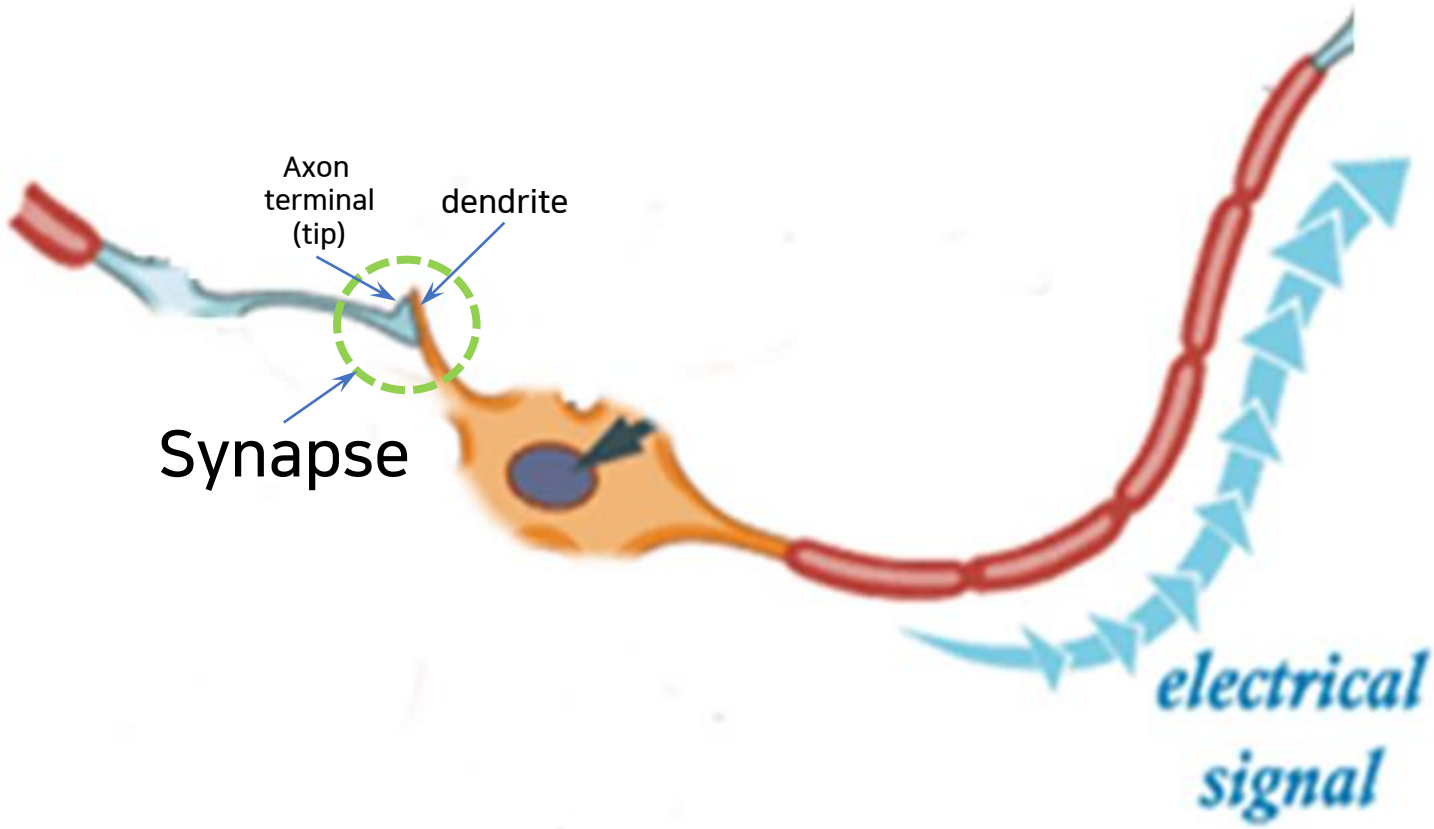
학습 = 연결 값을 조정하는 것

{강하게, 약하게}

두 뉴런의 연결

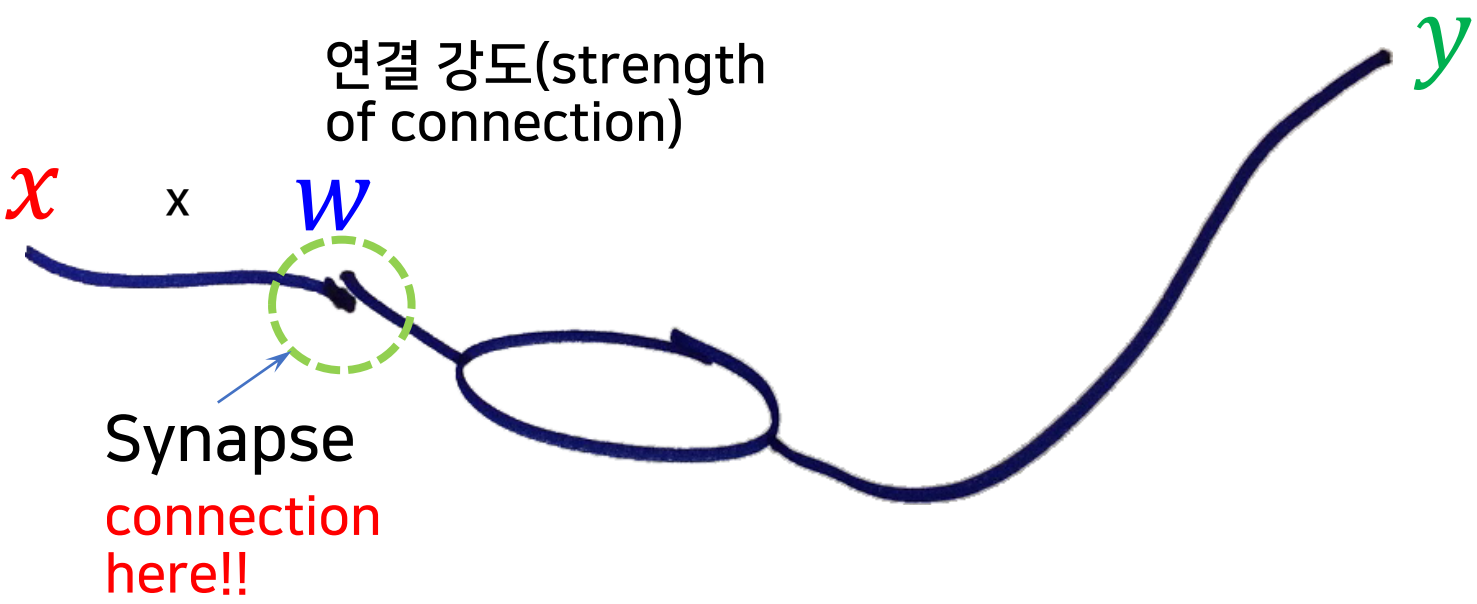


1개 입력을 갖는 뉴런



뉴런의 동작

w: $0, 7, -5$ 등 임의의 값



$$y = wx$$

뉴런의 동작은 매우 단순

입력(x) * 가중치(w)

$$y = wx$$

급료 계산기(응용의 예)

. 1시간(x) 일하면 1USD(y)를 번다고 할 때

. How much you get if work 4 hours? (prediction)

. 이를 위한 w 값을 구하라.

$$w = 1$$

$$y = 1x$$

학습이란?

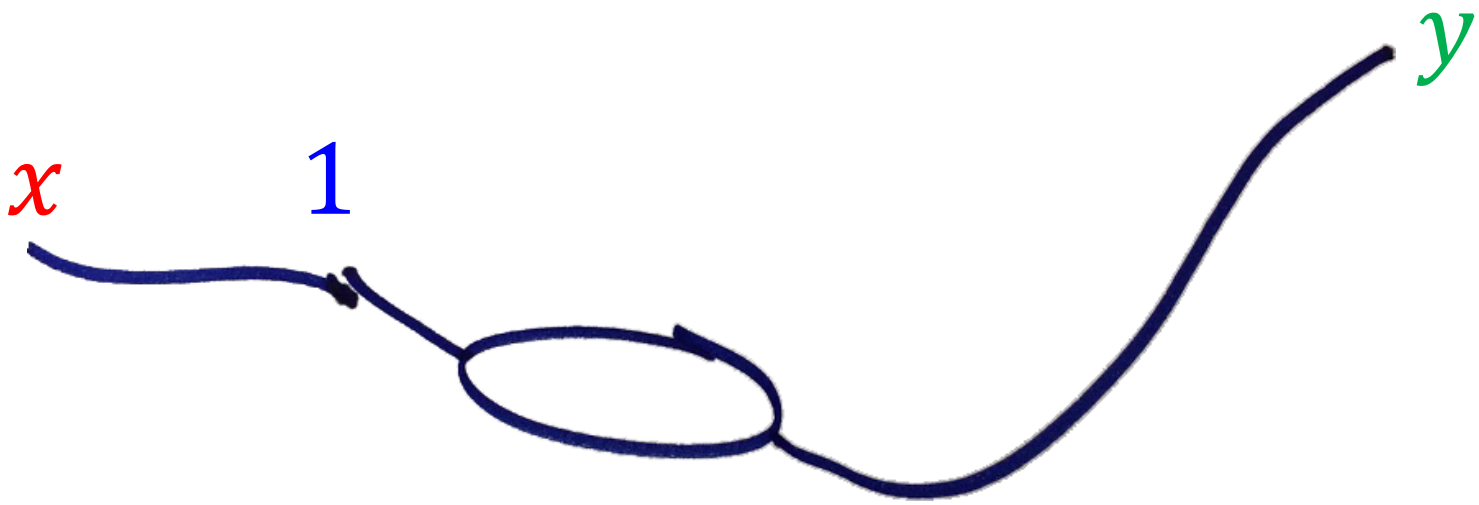
연결 값 w를 조정하는 것

{강하게, 약하게}

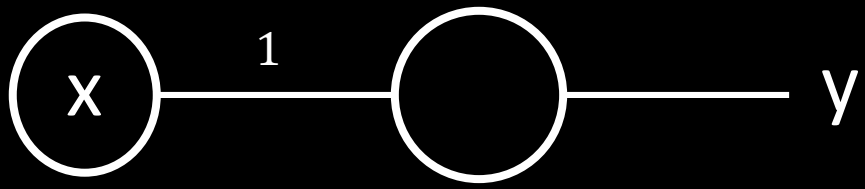
(Q) Draw a neuron

Representing the following equation:

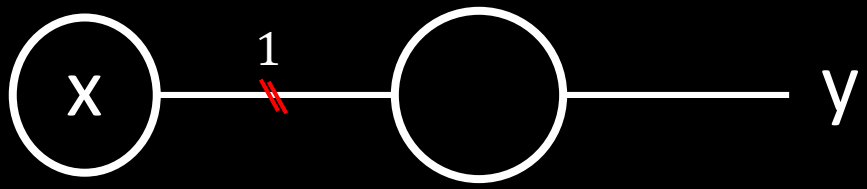
$$y = 1x$$



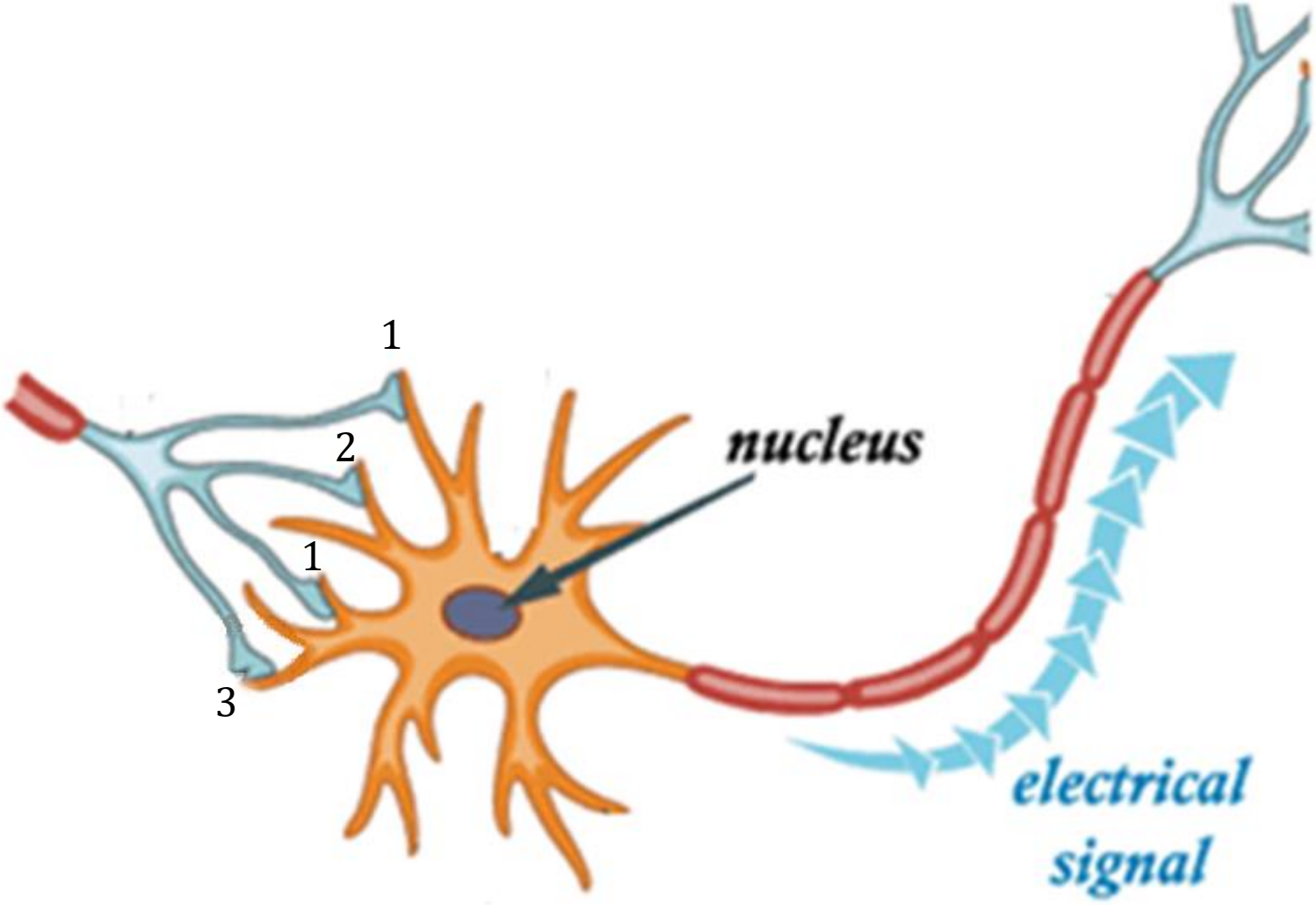
연결(시냅스)은 어디에 있을까?

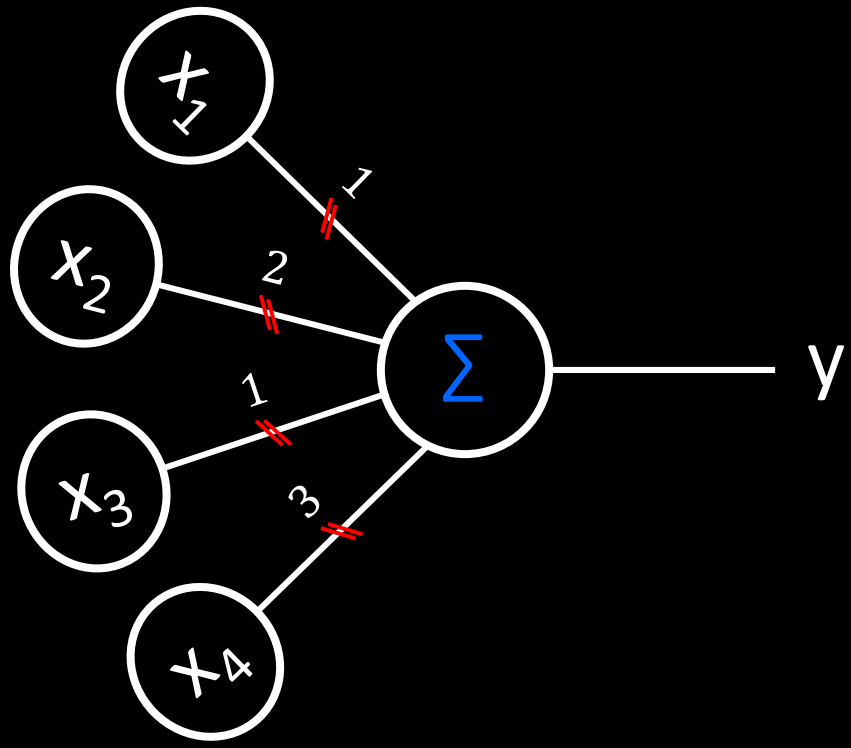


연결(시냅스)은 어디에 있을까?


x
1
y

여러 입력을 갖는 뉴런





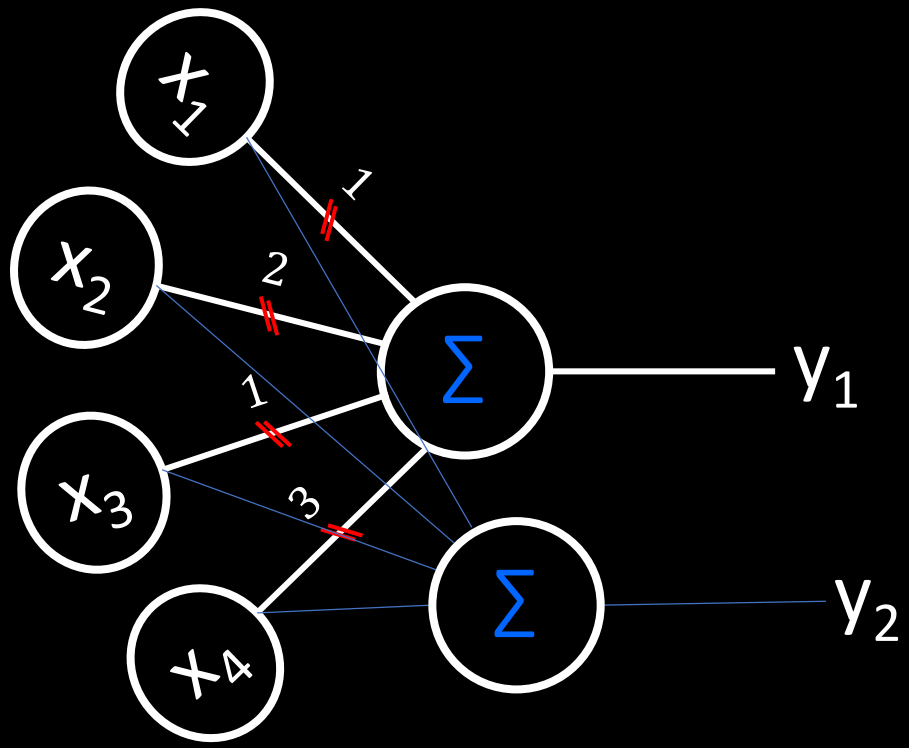
입력에 가중치를 곱하여
모두 더해서 (weighted sum) 출력
(x가 각각 1,1,1,1이면 출력 값은?)

입력의 수 = 연결의 수
(Synapses, Weights)

$$y = w1x1 + w2x2 + w3x3 + w4x4$$

$$y = [w1, w2, w3, w4] \begin{bmatrix} x1 \\ x2 \\ x3 \\ x4 \end{bmatrix}$$

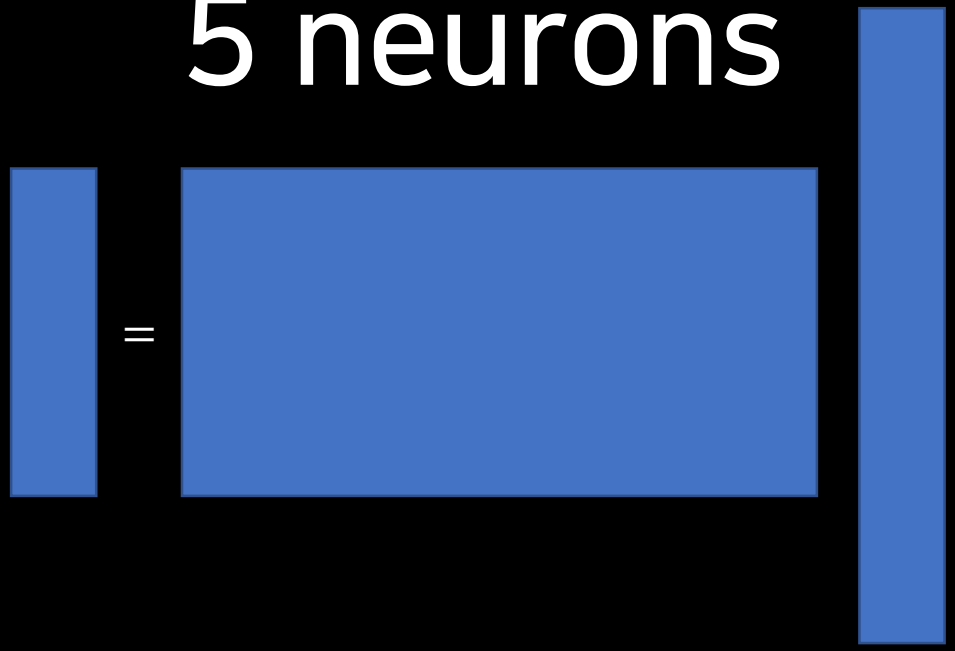
$$y = WX$$


x_1
x_2
x_3
x_4
1
2
1
3
Σ
Σ
y_1
y_2

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} {w_1}^1, {w_2}^1, {w_3}^1, {w_4}^1 \\ {w_1}^2, {w_2}^2, {w_3}^2, {w_4}^2 \end{bmatrix} \begin{bmatrix} x1 \\ x2 \\ x3 \\ x4 \end{bmatrix}$$

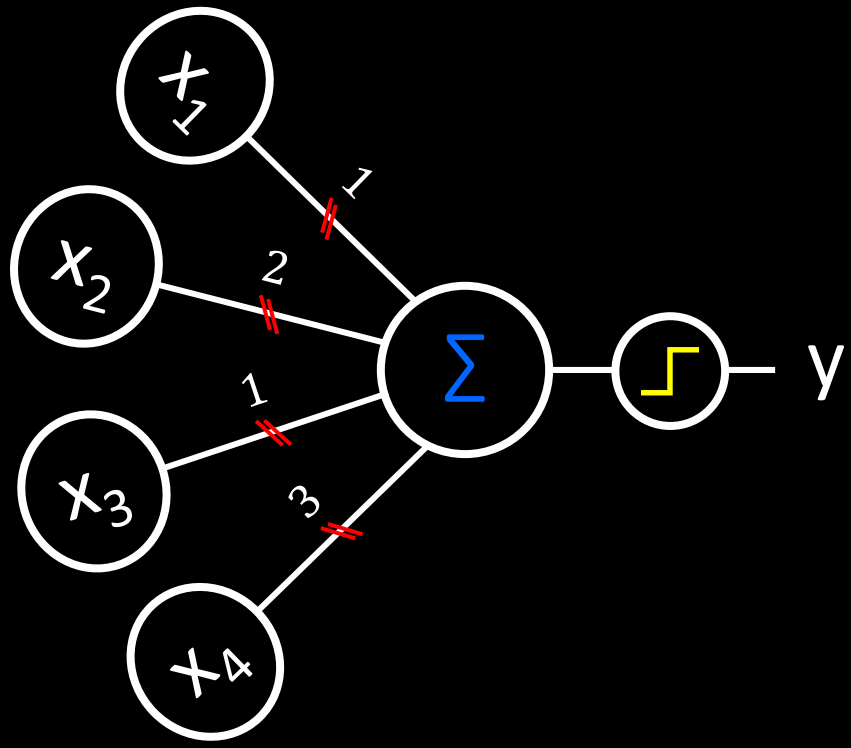
$$Y = WX$$

7 inputs
5 neurons

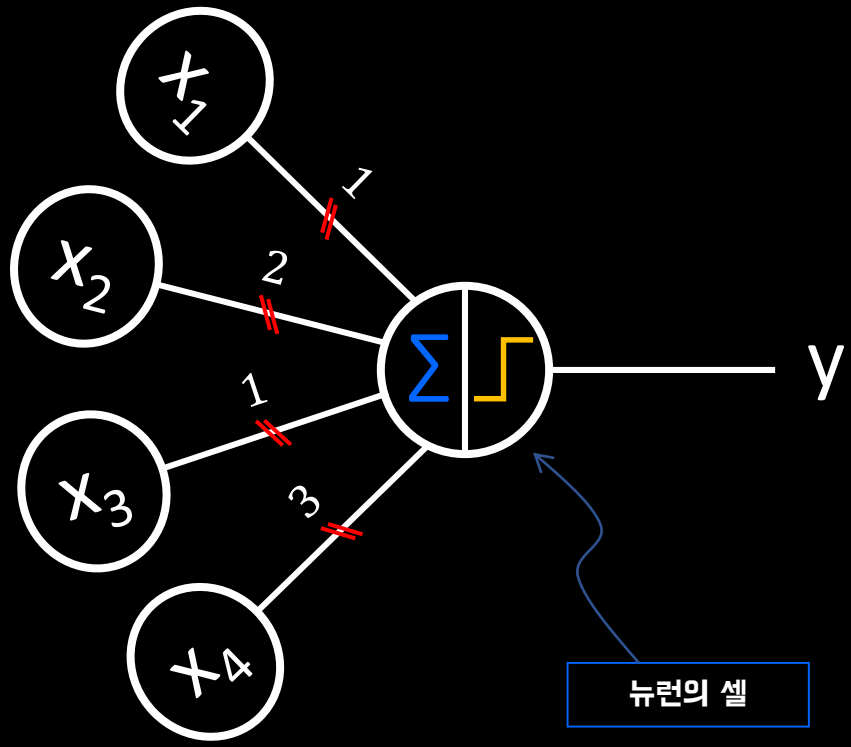


사실은..

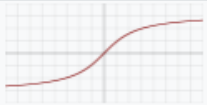
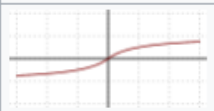
- 뉴런은 모두 더해서(weighted sum)
  일정한 값 이상일 때만 시그널 ON
- 그렇지 않으면 시그널 OFF



모두 더해서 특정 값(T) 이상이면 ON(1),
아니면 OFF(0)



모두 더해서 특정 값(T) 이상이면 ON(1),
아니면 OFF(0)

Name	Plot	Equation	Derivative (with respect to x)
Identity		$f(x) = x$	$f'(x) = 1$
Binary step		$f(x) = \begin{cases} 0 & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$	$f'(x) = \begin{cases} 0 & \text{for } x \neq 0 \\ ? & \text{for } x = 0 \end{cases}$
Logistic (a.k.a. Soft step)		$f(x) = \frac{1}{1 + e^{-x}}$	$f'(x) = f(x)(1 - f(x))$
TanH		$f(x) = \tanh(x) = \frac{2}{1 + e^{-2x}} - 1$	$f'(x) = 1 - f(x)^2$
ArcTan		$f(x) = \tan^{-1}(x)$	$f'(x) = \frac{1}{x^2 + 1}$
Softsign [7][8]		$f(x) = \frac{x}{1 + \lvert x \rvert}$	$f'(x) = \frac{1}{(1 + \lvert x \rvert)^2}$
Rectified linear unit (ReLU)[9]		$f(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \geq 0 \end{cases}$	$f'(x) = \begin{cases} 0 & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$
Leaky rectified linear unit (Leaky ReLU) [10]		$f(x) = \begin{cases} 0.01x & \text{for } x < 0 \\ x & \text{for } x \geq 0 \end{cases}$	$f'(x) = \begin{cases} 0.01 & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$
Parameteric rectified linear unit (PReLU)[11]		$f(\alpha, x) = \begin{cases} \alpha x & \text{for } x < 0 \\ x & \text{for } x \geq 0 \end{cases}$	$f'(\alpha, x) = \begin{cases} \alpha & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$
Randomized leaky rectified linear unit (RReLU)[12]		$f(\alpha, x) = \begin{cases} \alpha x & \text{for } x < 0 \text{ [1]} \\ x & \text{for } x \geq 0 \end{cases}$	$f'(\alpha, x) = \begin{cases} \alpha & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$
Exponential linear unit (ELU)[13]		$f(\alpha, x) = \begin{cases} \alpha(e^x - 1) & \text{for } x < 0 \\ x & \text{for } x \geq 0 \end{cases}$	$f'(\alpha, x) = \begin{cases} f(\alpha, x) + \alpha & \text{for } x < 0 \\ 1 & \text{for } x \geq 0 \end{cases}$

다음 뉴런을 그려보자.

(1) $y = 1x$

(2) $y = x_1 + 2x_2 + x_3 + 2x_4$

(3) $y = \begin{cases} 0 & if\ x_1 + 2x_2 + x_3 + 2x_4 > T \\ 1 & otherwise \end{cases}$

요 약

- 뉴런의 연결 부분, 시냅스
- 학습은 연결을 조정하는 것
- 파라미터(W) 튜닝
- 뉴런의 동작
- 뉴런 그리기