

# Deep Learning

from Scratch

## chapter 2

이새봄

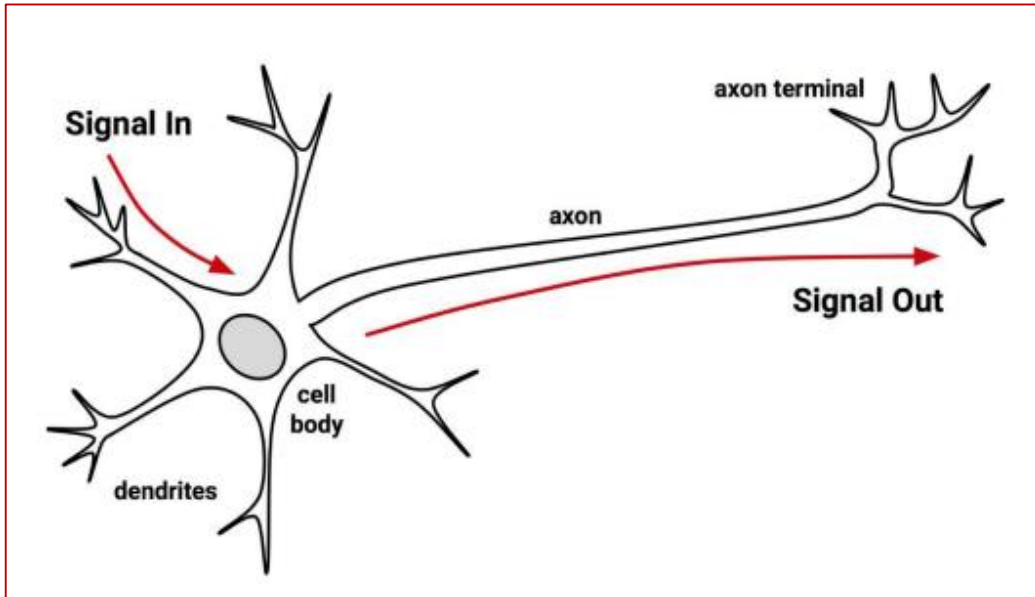
# I . Why Python?

- 간단하고 쉬움
- 오픈소스-> 무료
- 컴파일 과정 X
- 뛰어난 성능

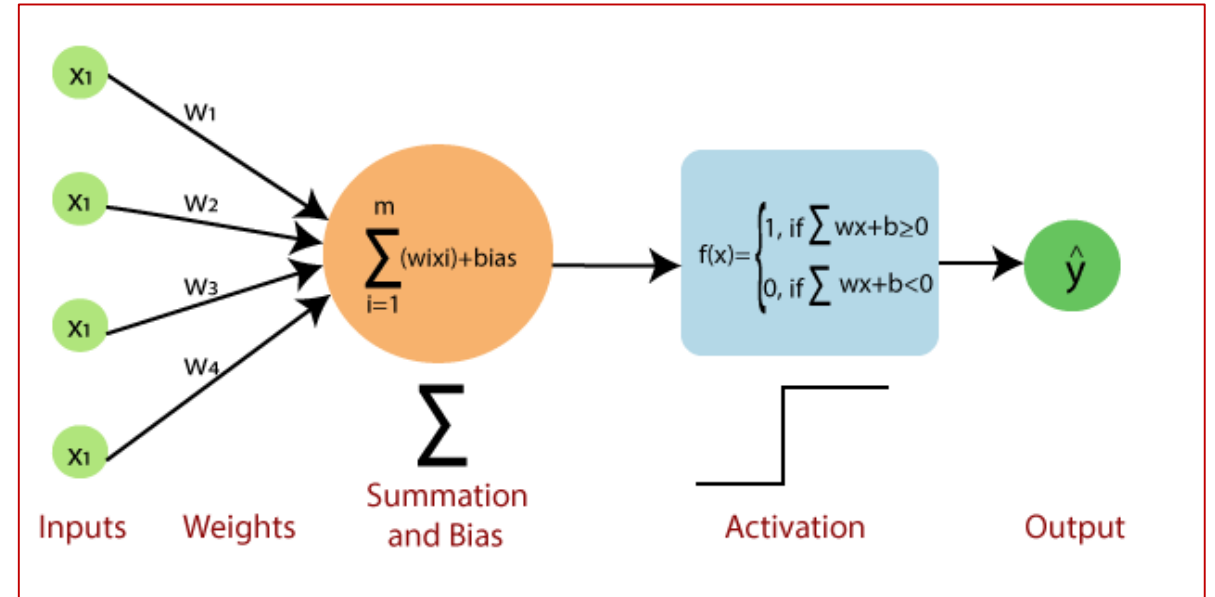


패키지 설치 및 관리  
부분 용이

## II. Perceptron



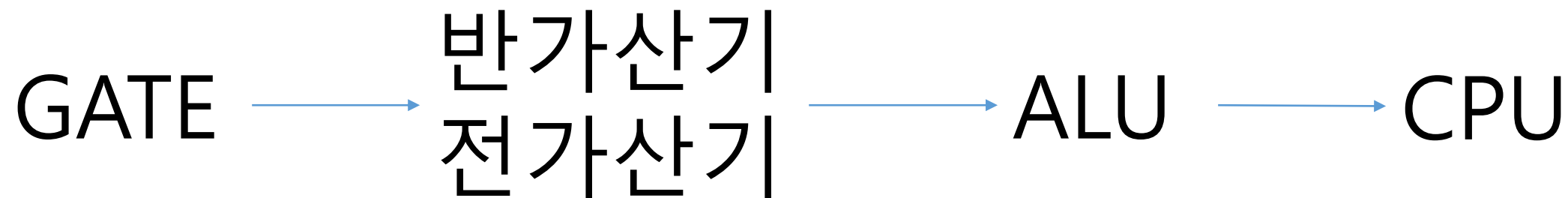
Biological Neuron



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

Artificial Neural Network

## II. Perceptron



# II. Perceptron

▼ AND 진리표

입력		연산자	출력
$x_1$	$x_2$		$y$
0	0	AND	0
0	1		0
1	0		0
1	1		1

▼ OR 진리표

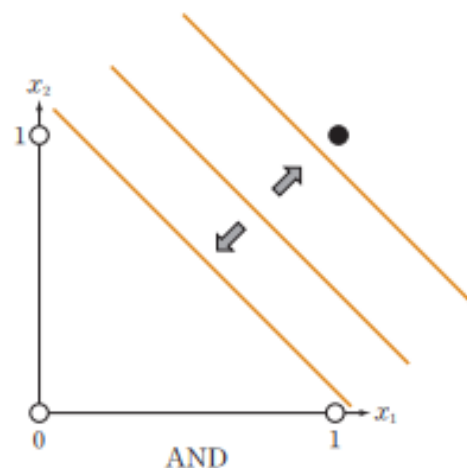
입력		연산자	출력
$x_1$	$x_2$		$y$
0	0	OR	0
0	1		1
1	0		1
1	1		1

▼ NAND 진리표

입력		연산자	출력
$x_1$	$x_2$		$y$
0	0	NAND	1
0	1		1
1	0		1
1	1		0

▼ 퍼셉트론식을 적용한 AND 진리표

입력		퍼셉트론	출력
$x_1$	$x_2$	$y = \begin{cases} 0 & (w_1x_1 + w_2x_2 \leq \text{임계값}) \\ 1 & (w_1x_1 + w_2x_2 > \text{임계값}) \end{cases}$ $(w_1, w_2, \text{임계값}) = (0.2, 0.2, 0.3)$	$y$
0	0	$0 \times 0.2 + 0 \times 0.2 \leq 0.3$	0
0	1	$0 \times 0.2 + 1 \times 0.2 \leq 0.3$	0
1	0	$1 \times 0.2 + 0 \times 0.2 \leq 0.3$	0
1	1	$1 \times 0.2 + 1 \times 0.2 > 0.3$	1



▲ 퍼셉트론의 범위

## II. Perceptron

```
In [5]: def AND(x1,x2):  
        w1,w2,theta=0.5,0.5,0.7  
        tmp = x1*w1 + x2*w2  
        if tmp<=theta:  
            return 0  
        elif tmp > theta:  
            return 1
```

```
In [6]: AND(0,0)
```

```
Out[6]: 0
```

```
In [7]: AND(1,0)
```

```
Out[7]: 0
```

```
In [8]: AND(0,1)
```

```
Out[8]: 0
```

```
In [9]: AND(1,1)
```

```
Out[9]: 1
```

```
In [23]: def AND(x1,x2):  
        x=np.array([x1,x2])  
        y=np.array([0.5,0.5])  
        b=-0.7  
        tmp = np.sum(w*x)+b  
        if tmp<=0:  
            return 0  
        else:  
            return 1
```

```
In [24]: AND(0,0)
```

```
Out[24]: 0
```

```
In [25]: AND(1,0)
```

```
Out[25]: 0
```

```
In [26]: AND(0,1)
```

```
Out[26]: 0
```

```
In [27]: AND(1,1)
```

```
Out[27]: 1
```

기존 퍼셉트론의 식

$$y = \begin{cases} 0 & (w_1x_1 + w_2x_2 \leq (\text{임계값})) \\ 1 & (w_1x_1 + w_2x_2 > (\text{임계값})) \end{cases}$$

⇒

변형된 퍼셉트론의 식

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

가중치: 입력신호가 결과에  
주는 영향력을 조절

편향: 뉴런이 얼마나 쉽게 활  
성화하느냐를 조정

## II. Perceptron

```
In [28]: def NAND(x1,x2):  
    x=np.array([x1,x2])  
    y=np.array([-0.5,-0.5])  
    b= 0.7  
    tmp = np.sum(w*x)+b  
    if tmp<=0:  
        return 0  
    else:  
        return 1
```

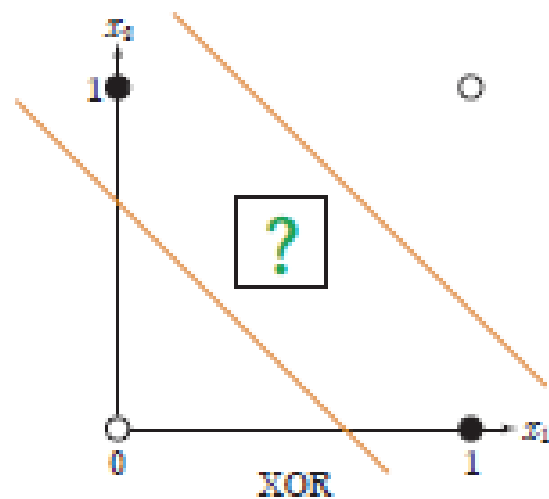
```
In [29]: def OR(x1,x2):  
    x=np.array([x1,x2])  
    y=np.array([0.5,0.5])  
    b= -0.2  
    tmp = np.sum(w*x)+b  
    if tmp<=0:  
        return 0  
    else:  
        return 1
```

가중치와 편향만 조절

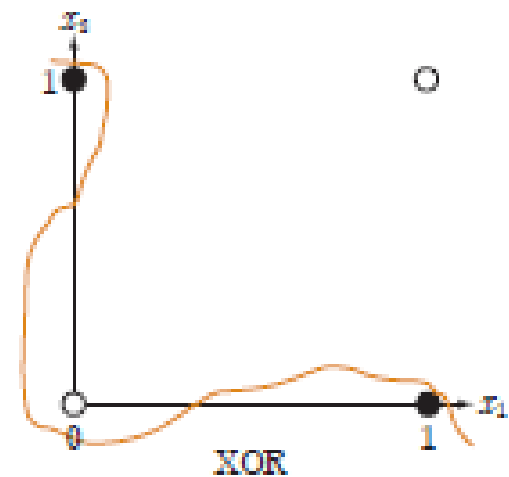
## II. Perceptron

### ▼ XOR 진리표

입력		연산자	출력
$x_1$	$x_2$		$y$
0	0	XOR	0
0	1		1
1	0		1
1	1		0



(a) XOR 연산 결과 분류에 실패한 퍼셉트론

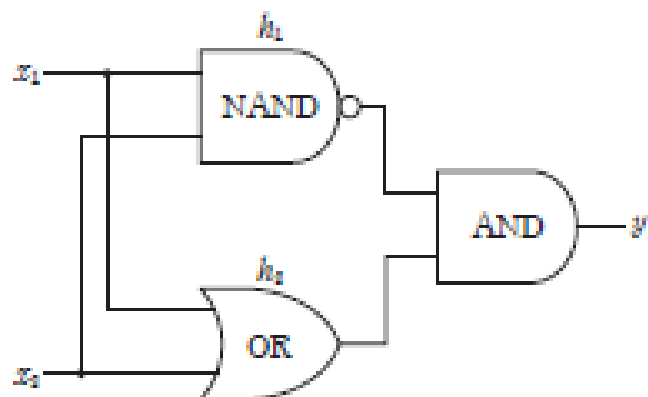


(b) XOR 연산 결과를 분류하는 곡선



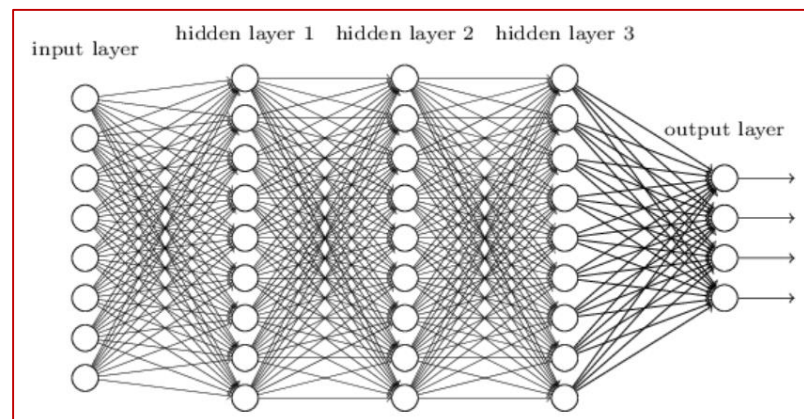
## II. Perceptron

### 다층 퍼셉트론



▼ XOR 진리표

입력		연산자		연산자	출력
		NAND	OR		
$x_1$	$x_2$	$h_1$	$h_2$	AND	$y$
0	0	1	0		0
0	1	1	1		1
1	0	1	1		1
1	1	0	1		0



```
In [71]: def XOR(x1,x2):  
         s1=NAND(x1,x2)  
         s2=OR(x1,x2)  
         y=AND(s1,s2)  
         return y
```

```
In [72]: XOR(0,0)
```

```
Out[72]: 0
```

```
In [73]: XOR(1,0)
```

```
Out[73]: 1
```

```
In [74]: XOR(0,1)
```

```
Out[74]: 1
```

```
In [75]: XOR(1,1)
```

```
Out[75]: 0
```