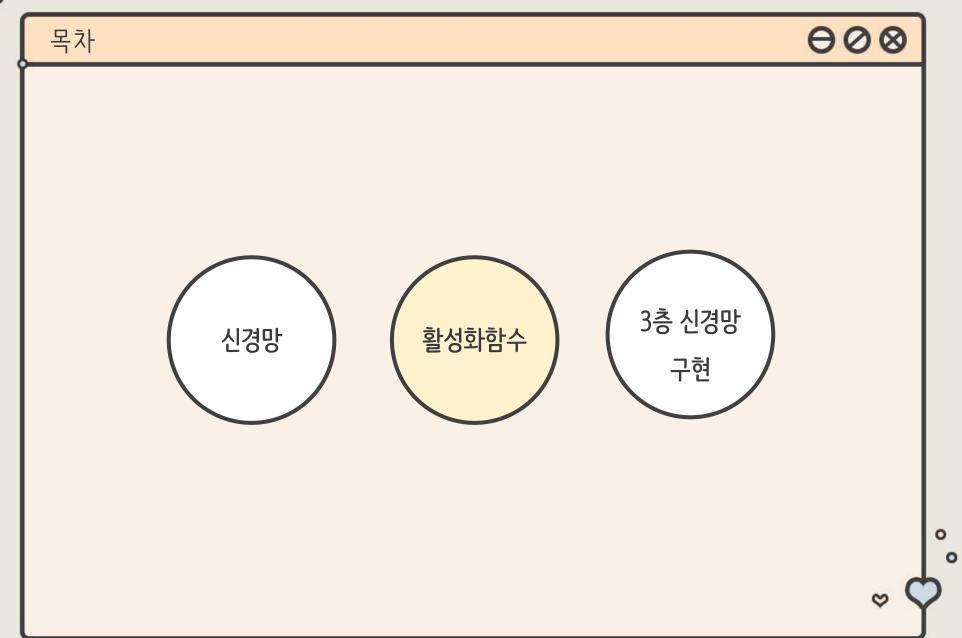


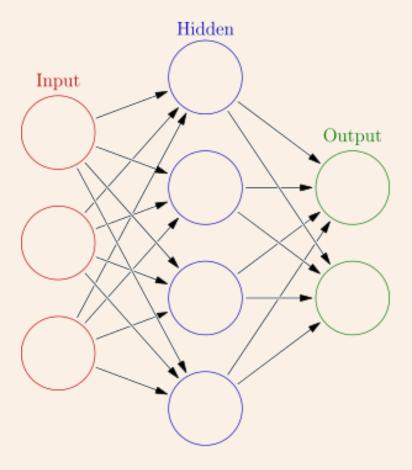
신경망



## 신경망



# 신경망? 다층 퍼셉트론?



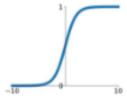


# 활성화 함수(Activation Function)

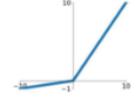
# **Activation Functions**

## **Sigmoid**

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

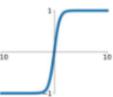


# Leaky ReLU max(0.1x, x)



## tanh

tanh(x)

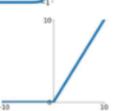


### Maxout

 $\max(w_1^T x + b_1, w_2^T x + b_2)$ 

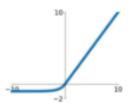
## ReLU

 $\max(0, x)$ 



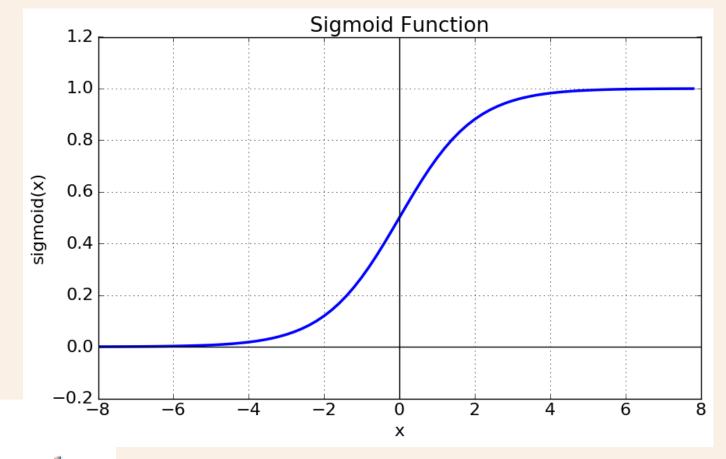
#### **ELU**

$$\begin{array}{ll}
x & x \ge 0 \\
\alpha(e^x - 1) & x < 0
\end{array}$$



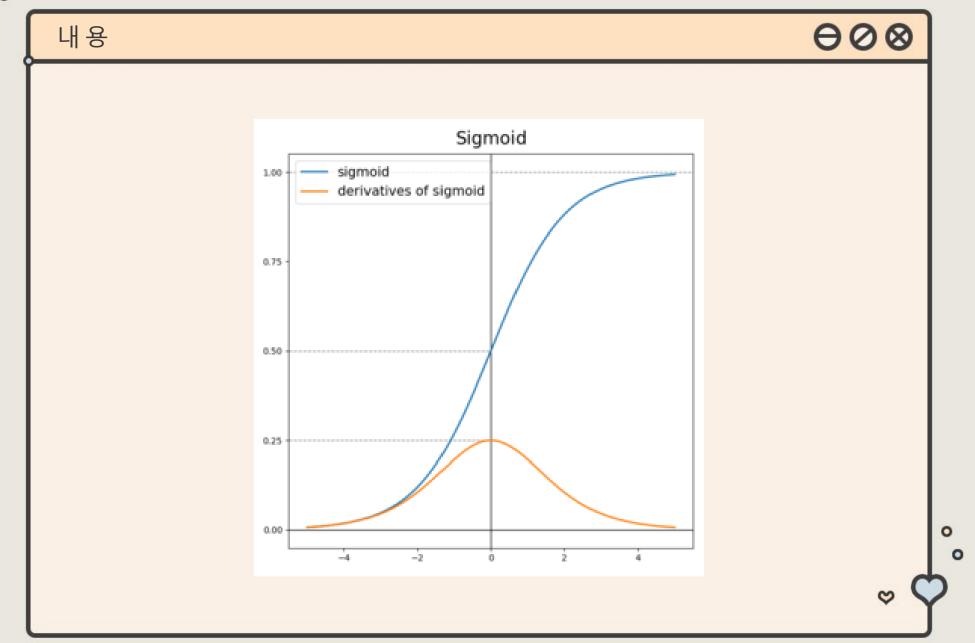
## 활성화 함수





Sigmoid

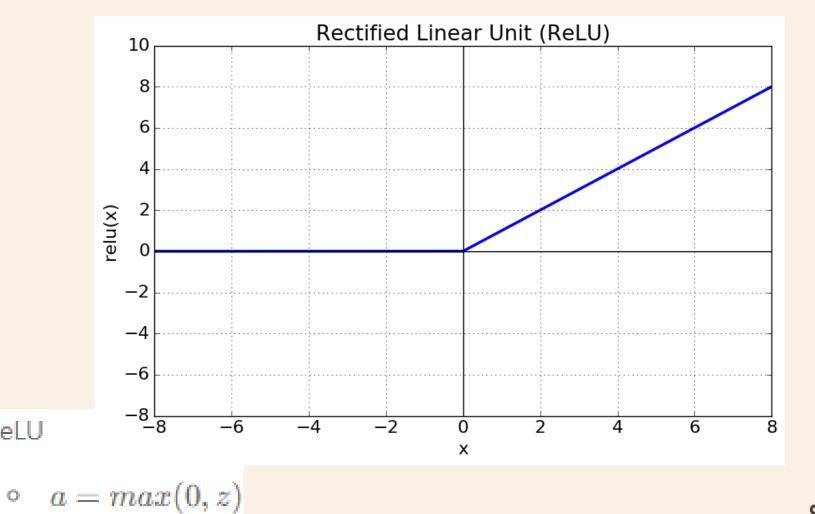
$$a = \frac{1}{1 + e^{-z}}$$



## 활성화 함수

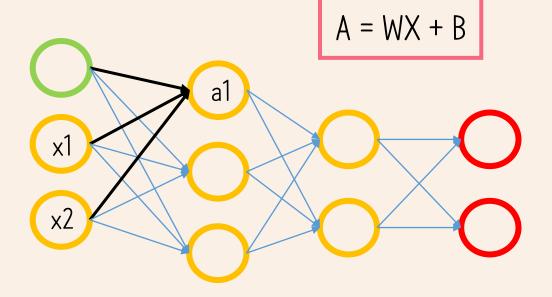
ReLU





## 3층 신경망 구현





```
X = np.array([1.0,0.5])

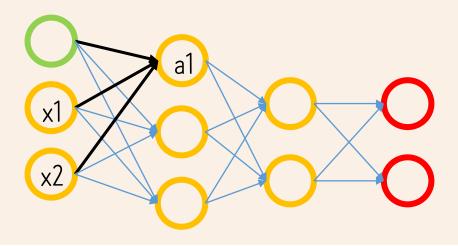
W1 = np.array([[0.1, 0.3, 0.5], [0.2, 0.4, 0.6]])

B1 = np.array([0.1,0.2,0.3])
```

## ° \_\_\_\_

3층 신경망 구현





```
def sigmoid(x):
    return 1/(1+np.exp(-x))
```

```
A1 = np.dot(X, W1) + B1

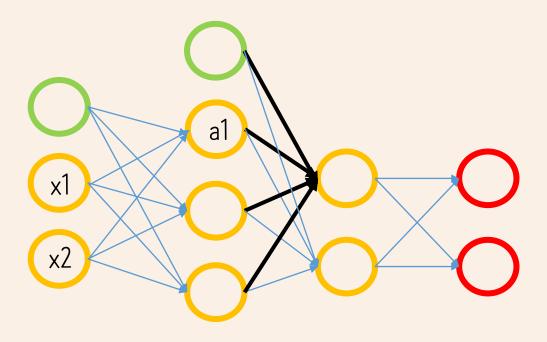
Z1 = sigmoid(A1)
```

print(A1)
print(Z1)

[0.3 0.7 1.1] [0.57444252 0.66818777 0.75026011]

## 3층 신경망 구현





```
W2 = np.array([[0.1,0.4],[0.2,0.5],[0.3,0.6]])
```

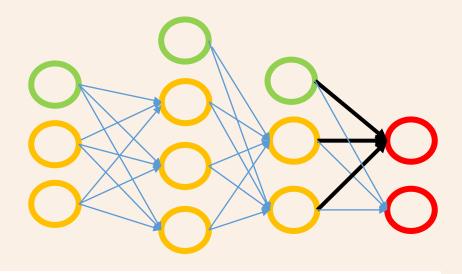
B2 = np.array([0.1,0.2])

A2 = np.dot(Z1, W2) + B2

Z2 = sigmoid(A2)

## 3층 신경망 구현





```
def identity_function(x):
    return x
```

```
W3 = np.array([[0.1,0.3],[0.2,0.4]])
B3 = np.array([0.1,0.2])
```

#### print(Y)

[0.31682708 0.69627909]

회귀 - 항등 함수 2클래스 분류 - 시그모이드 함수 다중 클래스 분류 - 소프트맥스 함수

$$f(\vec{x})_i = rac{e^{x_i}}{\sum_{k=1}^K e^{x_k}}$$
 for  $i$  = 1, ...,  $K$ 

```
def softmax(a):
    c = np.max(a)
    exp_a = np.exp(a-c)
    sum_exp_a = np.sum(exp_a)
    y = exp_a / sum_exp_a
    return y
```

```
a = np.array([0.3,2.9,4.0])
y = softmax(a)
print(y)
```

[0.01821127 0.24519181 0.73659691]

```
np.sum(y)
```

1.0

