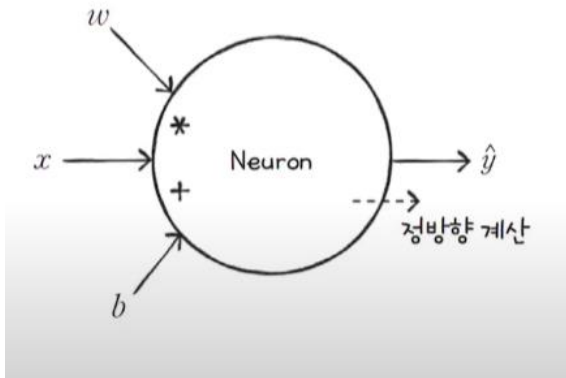


# 디지털 영상처리 연구실 연구보고서

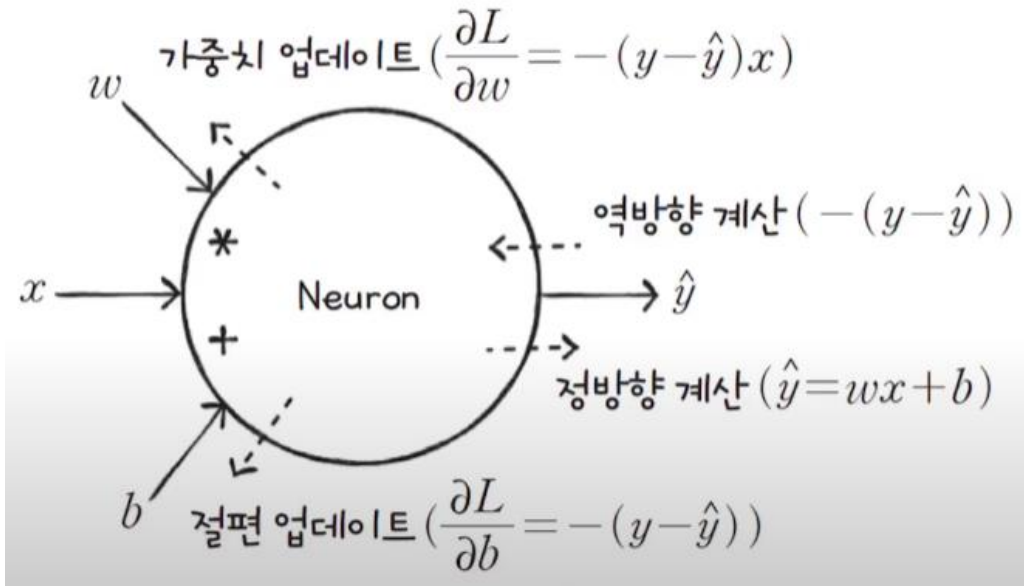
김우현

#뉴런으로 도식화



$$\hat{y} = w \times x + b$$

#역방향 계산



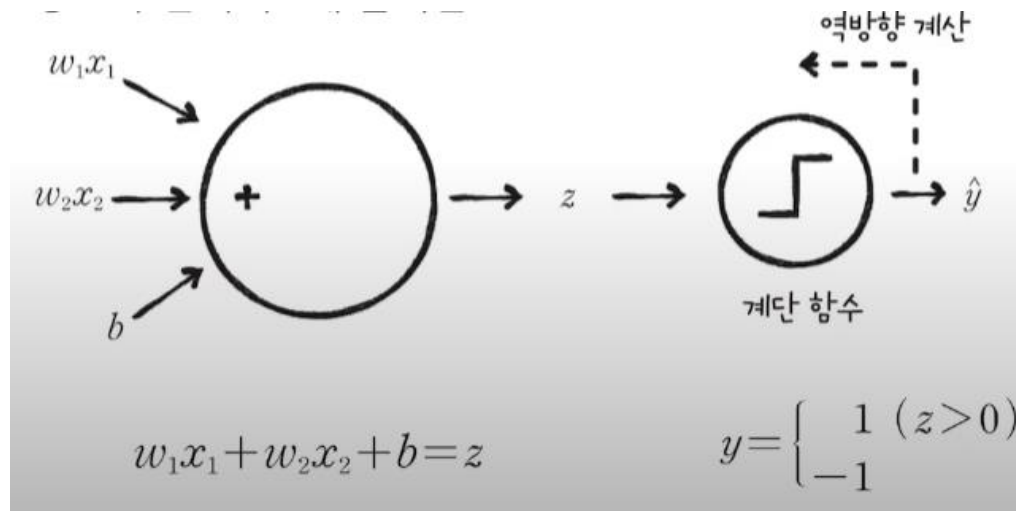
->  $w_{\text{new}} = w + w_{\text{rate}} * \text{err}$

$b_{\text{new}} = b + 1 * \text{err}$

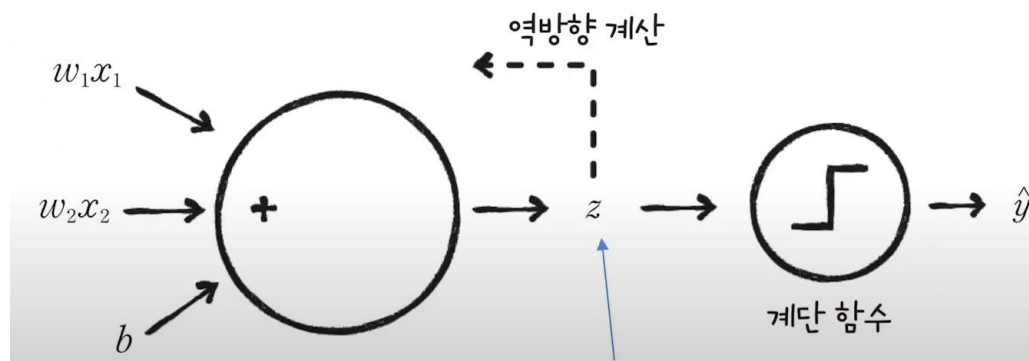
## ##로지스틱 회귀

### #퍼셉트론

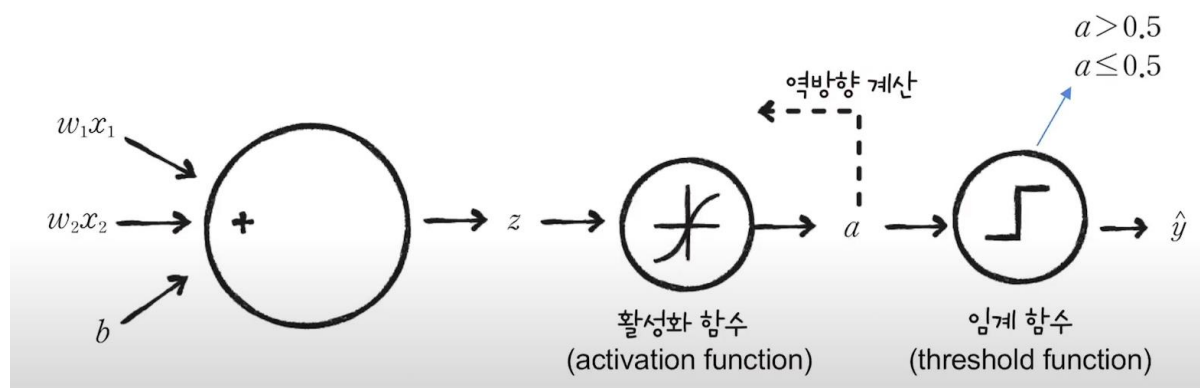
->이진분류를 위한 알고리즘



### #아달린

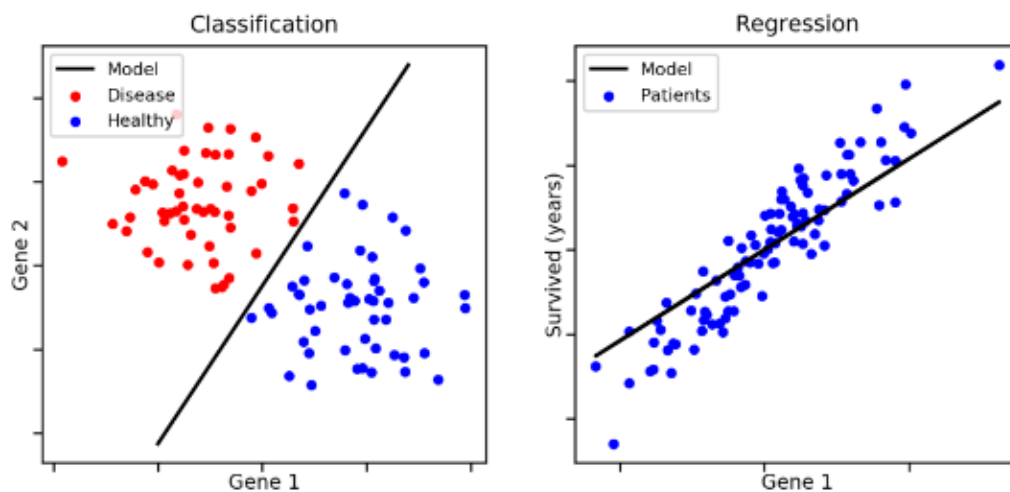


## #로지스틱 회귀



->활성화 함수는 비선형 함수사용(시그모이드함수)

### \*\*\*회귀와 분류\*\*\*

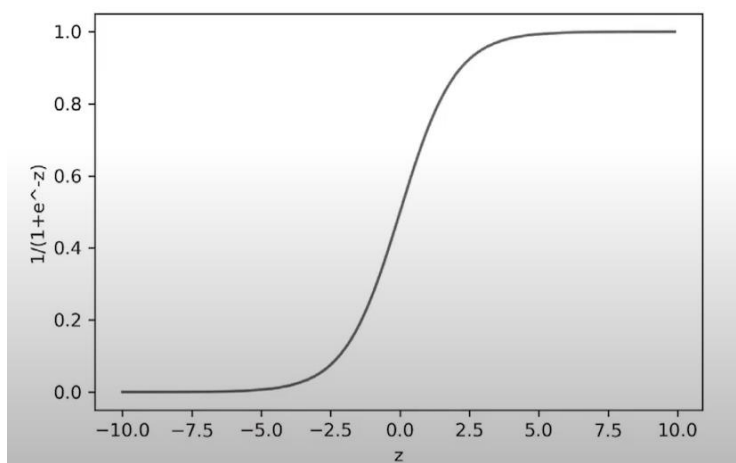
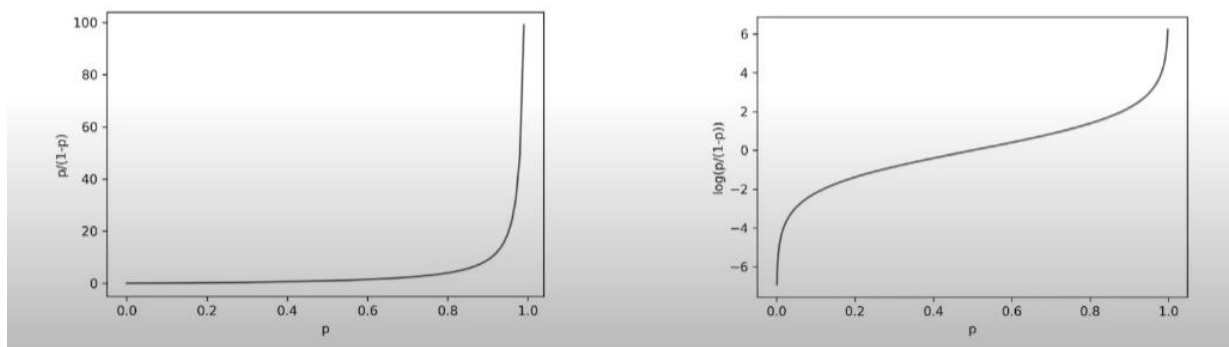


### #시그모이드 함수

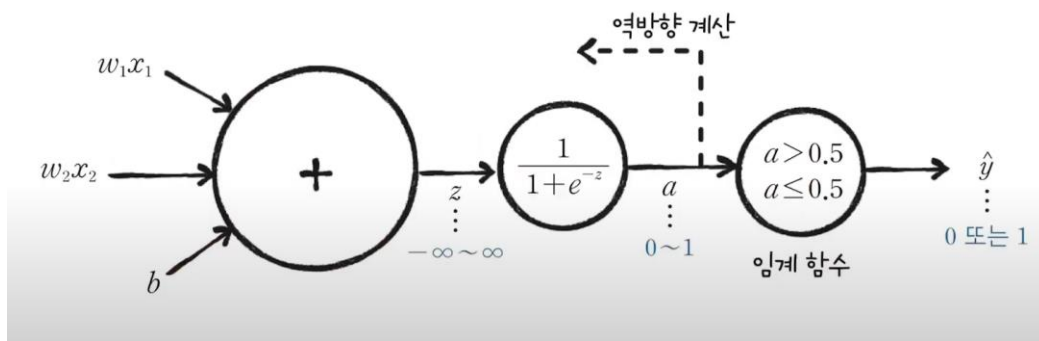
오즈 비(odds ratio) --> 로짓 함수(logit function) --> 시모이드 함수

$$OR(odds\ ratio) = \frac{p}{1-p} \quad (p = \text{성공 확률})$$

$$logit(p) = \log\left(\frac{p}{1-p}\right)$$



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



## #로지스틱 손실함수

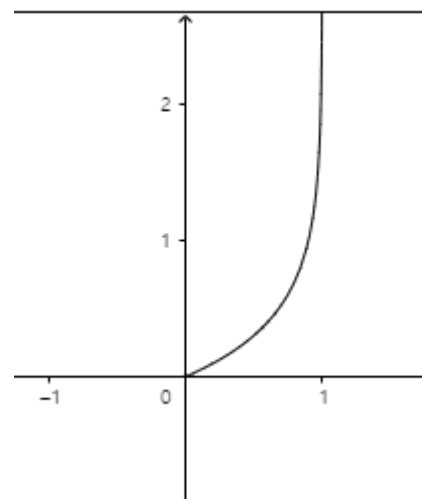
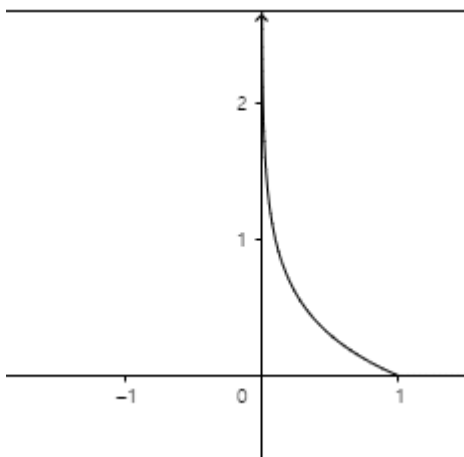
->경사 하강법을 이용하기 위하여 미분가능한 로지스틱 손실함수 사용

$$L = -(y \log(a) + (1 - y) \log(1 - a))$$

로지스틱 손실함수



	$L$
$y = 1$ (양성클래스)	$-\log(a)$
$y = 0$ (음성클래스)	$-\log(1 - a)$



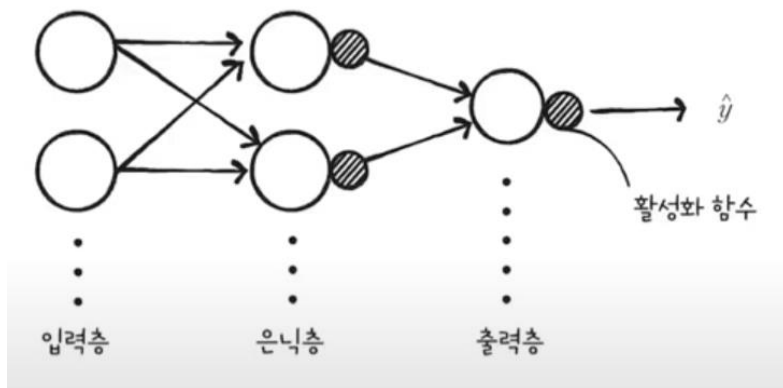
## #제곱오차의 미분과 로지스틱 손실함수의 미분

	제곱 오차의 미분	로지스틱 손실 함수의 미분
가중치에 대한 미분	$\frac{\partial SE}{\partial w} = -(y - \hat{y})x$	$\frac{\partial L}{\partial w} = -(y - a)x$
절편에 대한 미분	$\frac{\partial SE}{\partial b} = -(y - \hat{y})1$	$\frac{\partial L}{\partial b} = -(y - a)1$

## #가중치 업데이트

$$w_i = w_i - \frac{\partial L}{\partial w_i} = w_i + (y - a)x_i$$

## #인공 신경망(딥러닝,심층 신경망)



## #단일층 신경망

