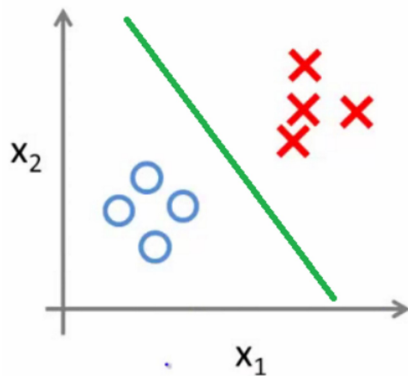


### 3. Logistic(sigmoid) Classification

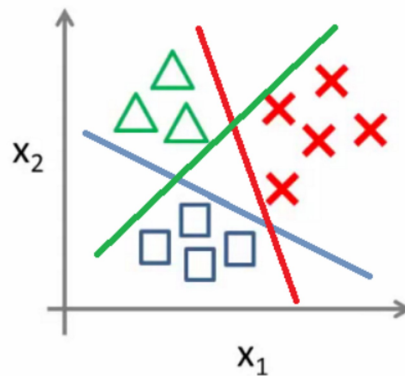
2019년 3월 29일 금요일    오후 12:47

#### 1. **Definition**

Binary classification:

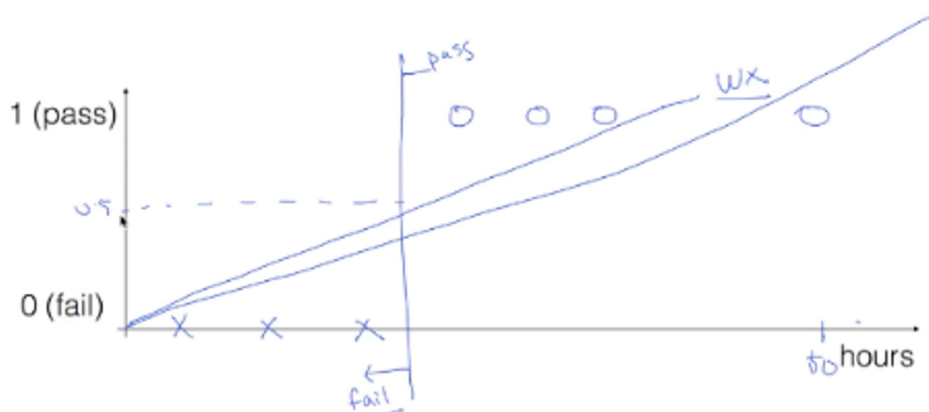


Multi-class classification:



- **Binary**한 분류 Classification (Binary) -> 0,1 encoding
- Logistic Regression 의 한계 :  $H(x) = Wx + b$  는 값이  $x$  값에 따라 1보다 무한정 커질 수 있다.

#### Linear Regression?



#### 2. **Hypothesis**



$$H_L(X) = WX$$



$$Z = H_L(X)$$



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



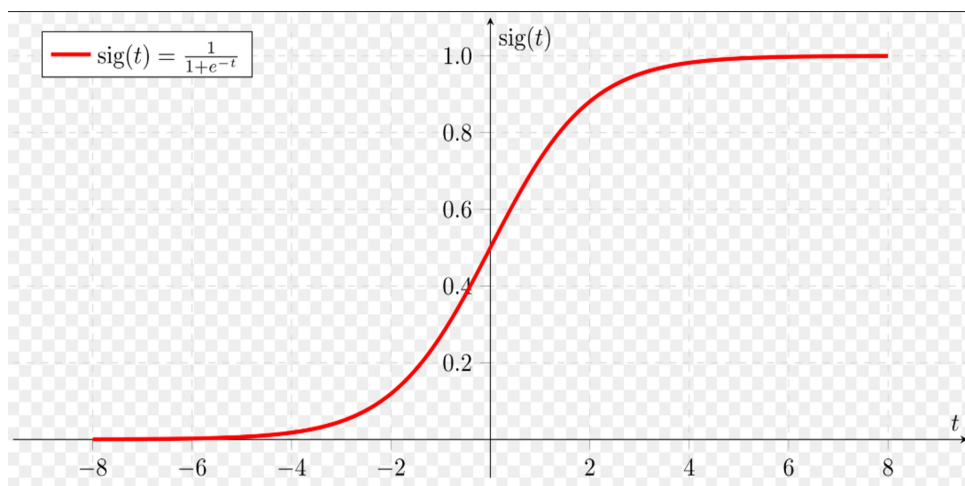
$$H_R(X) = g(H_L(X))$$



$$H(X) = \frac{1}{1 + e^{-W^T X}}$$

### - **sigmoid function :**

- logistic regression의 한계를 해결
- $g(z)$  : logistic, sigmoid function
- $z$ 가 무한히 커지면 1에 수렴하고, 무한히 작아지면 0에 수렴
- 어떤 값이라도 0과 1사이로 나오도록 조정



### 3. **Cost function**

- 일반적인 이차 코스트 함수를 쓸 수 없다. -> log를 취한다.

# Cost function

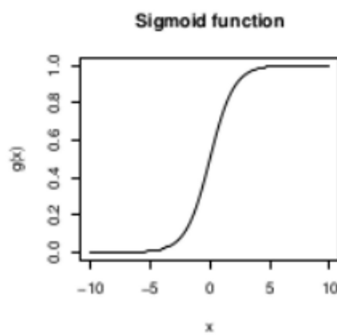
$$\text{cost}(W) = \frac{1}{m} \sum c(H(x), y)$$

$$J(\theta) = \frac{1}{m} \sum_{i=1}^m \text{Cost}(h_{\theta}(x^{(i)}), y^{(i)})$$

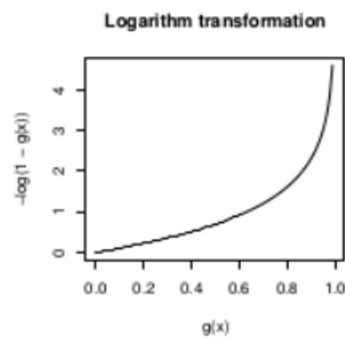
$$\text{Cost}(h_{\theta}(x), y) = -\log(h_{\theta}(x)) \quad \text{if } y = 1$$

$$\text{Cost}(h_{\theta}(x), y) = -\log(1 - h_{\theta}(x)) \quad \text{if } y = 0$$

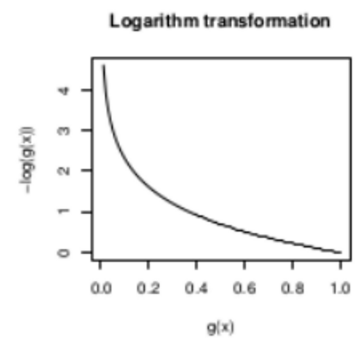
$$C(H(x), y) = y \log(H(x)) - (1 - y) \log(1 - H(x))$$



(a) Sigmoid function.



(b) Cost for  $y = 0$ .



(c) Cost for  $y = 1$ .

## 4. Gradient descent algorithm