

<복제물에 대한 경고>

본 저작물은 **저작권법 제25조 수업목적 저작물 이용 보상금제도**에 의거, **한국복제전송저작권협회**와 약정을 체결하고
적법하게 이용하고 있습니다. 약정범위를 초과하는 사용은 저작권법에 저촉될 수 있으므로

저작물의 재 복제 및 수업 목적 외의 사용을 금지합니다.

2020. 03. 30.

건국대학교(서울)한국복제전송저작권협회

<전송에 대한 경고>

본 사이트에서 수업 자료로 이용되는 저작물은 **저작권법 제25조 수업목적 저작물 이용 보상금제도**에 의거,

한국복제전송저작권협회와 약정을 체결하고 적법하게 이용하고 있습니다.

약정범위를 초과하는 사용은 저작권법에 저촉될 수 있으므로

수업자료의 대중 공개·공유 및 수업 목적 외의 사용을 금지합니다.

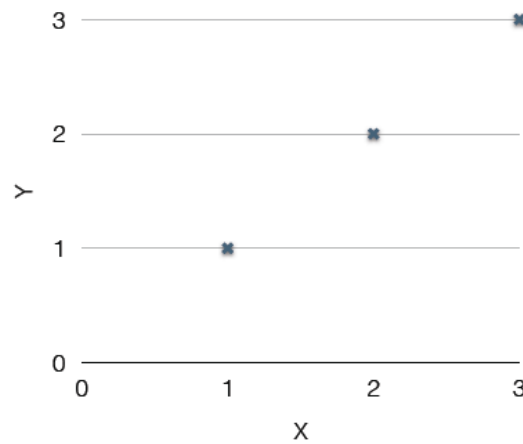
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Concept of Machine Learning

What is Learning?

x	Y
1	1
2	2
3	3



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Linear Hypothesis

$$H(x) = Wx + b$$



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Multi-variable

$$H(x_1, x_2) = w_1x_1 + w_2x_2 + b$$

$$H(x_1, x_2, x_3, \dots, x_n) = w_1x_1 + w_2x_2 + w_3x_3 + \dots + w_nx_n + b$$

Matrix representation

$$w_1x_1 + w_2x_2 + w_3x_3 + \dots + w_nx_n$$

$$\begin{bmatrix} w_1 & w_2 & w_3 \end{bmatrix} \times \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} w_1 \times x_1 + w_2 \times x_2 + w_3 \times x_3 \end{bmatrix}$$



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Matrix Representation

$$\begin{bmatrix} w_1 & w_2 & w_3 \end{bmatrix} \times \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} w_1 \times x_1 + w_2 \times x_2 + w_3 \times x_3 \end{bmatrix}$$

$$H(X) = WX + b$$

With b vector

$$\begin{bmatrix} b & w_1 & w_2 & w_3 \end{bmatrix} \times \begin{bmatrix} 1 \\ x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b \times 1 + w_1 \times x_1 + w_2 \times x_2 + w_3 \times x_3 \end{bmatrix}$$

$$H(X) = WX$$

Without b vector

$$H(X) = W^T X$$

Transpose representation

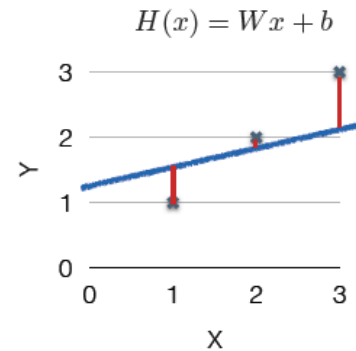


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Which hypothesis is better?

- How fit the line to our (training) data

$$H(x) - y$$

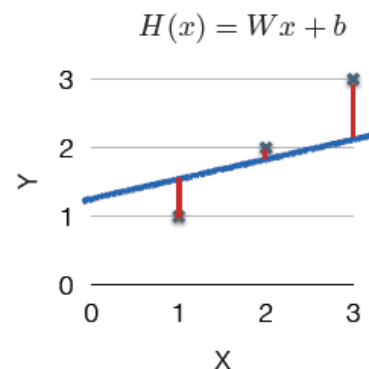


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Cost Function

$$\frac{(H(x^{(1)}) - y^{(1)})^2 + (H(x^{(2)}) - y^{(2)})^2 + (H(x^{(3)}) - y^{(3)})^2}{3}$$

$$cost = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2$$



Our goal? minimize $cost(W, b)$

Cost function을 최소로 하는 hypothesis가 무엇일까?



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Hypothesis and Cost

$$H(x) = Wx + b$$

$$\text{cost}(W, b) = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2$$



Simplifying without b vector

$$H(x) = Wx$$

$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$



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What cost(W) looks like?

$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$

x	y
1	1
2	2
3	3

- $W=1, \text{cost}(W)=0$

$$\frac{1}{3}((1 * 1 - 1)^2 + (1 * 2 - 2)^2 + (1 * 3 - 3)^2)$$

- $W=0, \text{cost}(W)=4.67$

$$\frac{1}{3}((0 * 1 - 1)^2 + (0 * 2 - 2)^2 + (0 * 3 - 3)^2)$$

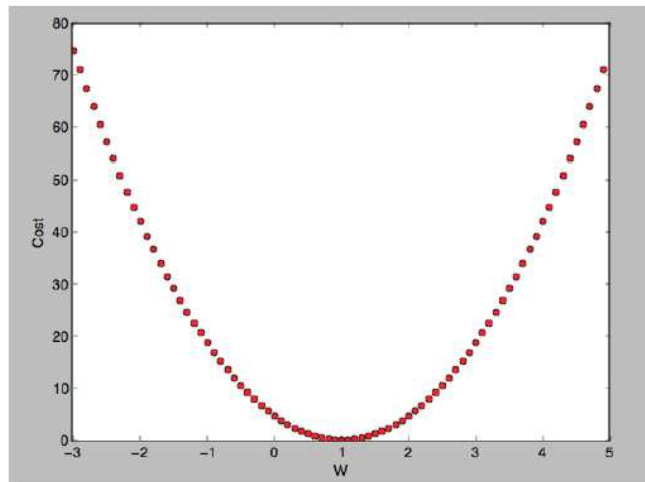
- $W=2, \text{cost}(W)=?$?



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What cost(W) looks like?

$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$

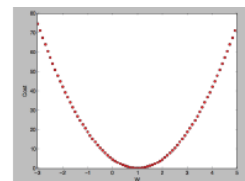


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How to Minimize Cost?

= How to find the lowest point?

- Start with initial guesses
 - Start at 0,0 (or any other value)
 - Keeping changing W and b a little bit to try and reduce cost(W, b)
- Each time you change the parameters, you select the gradient which reduces cost(W, b) the most possible
- Repeat
- Do so until you converge to a local minimum
- Has an interesting property
 - Where you start can determine which minimum you end up



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Formal Definition of Gradient Decent

$$cost(W) = \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2 \quad \longrightarrow \quad cost(W) = \frac{1}{2m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$

$$W := W - \alpha \frac{\partial}{\partial W} \frac{1}{2m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$

$$W := W - \alpha \frac{1}{2m} \sum_{i=1}^m 2(Wx^{(i)} - y^{(i)})x^{(i)} \quad \longleftarrow \quad W := W - \alpha \frac{\partial}{\partial W} cost(W)$$

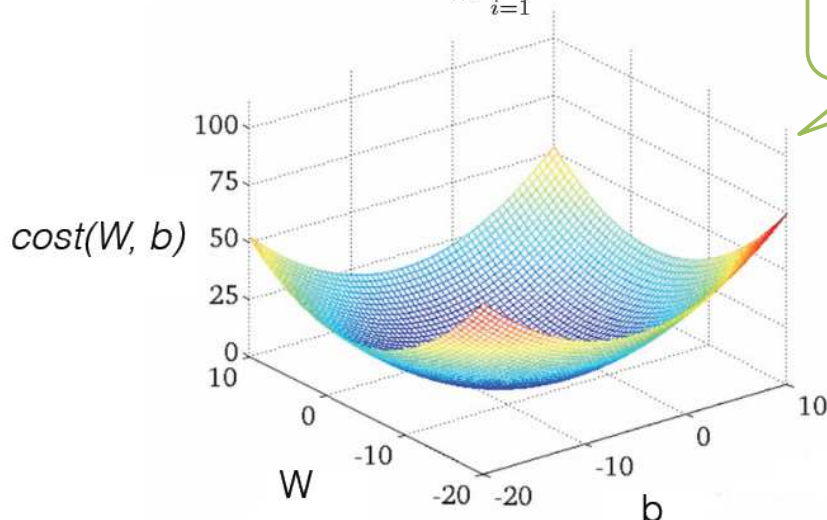
$$W := W - \alpha \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})x^{(i)}$$



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Convex Function

$$cost(W, b) = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2$$



Learning → Cost를 최소화하는 W와 b를 찾자!



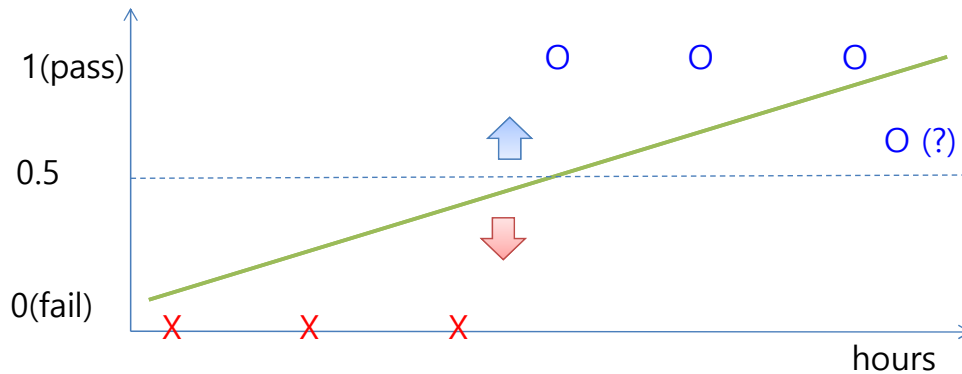
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Regression to Classification

Classification problems

- Spam Detection: Spam (1) or Ham (0)
- Facebook feed: show(1) or hide(0)
- Credit Card Fraudulent Transaction detection: legitimate(0) or fraud (1)

Pass/Fail based on study hours?

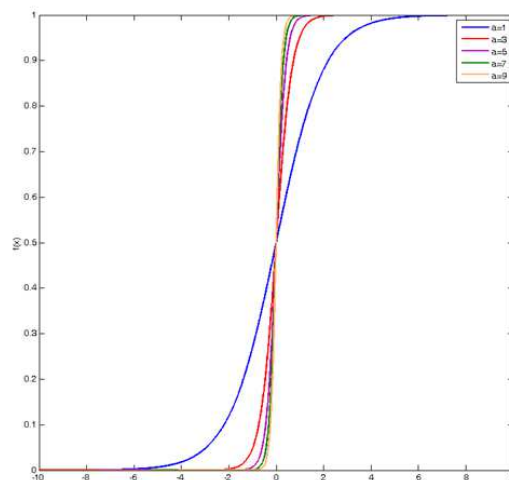


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Logistic Hypothesis

$$H(x) = Wx + b \Rightarrow g(z) = \frac{1}{1 + e^{-z}}$$

WHY?
0과 1 사이 값으로 변환



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Logistic Hypothesis & Cost Function

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

$$cost(W, b) = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2 \quad \Rightarrow \quad \text{Many local minimums}$$

New Cost Function

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

$$c(H(x), y) = \begin{cases} -\log(H(x)) & : y = 1 \\ -\log(1 - H(x)) & : y = 0 \end{cases} \quad H(x)=1 \text{ 일 때 } C\text{값은? } \boxed{?}$$



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Cost Function

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

$$c(H(x), y) = \begin{cases} -\log(H(x)) & : y = 1 \\ -\log(1 - H(x)) & : y = 0 \end{cases}$$



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

Minimize Cost → Gradient decent algorithm

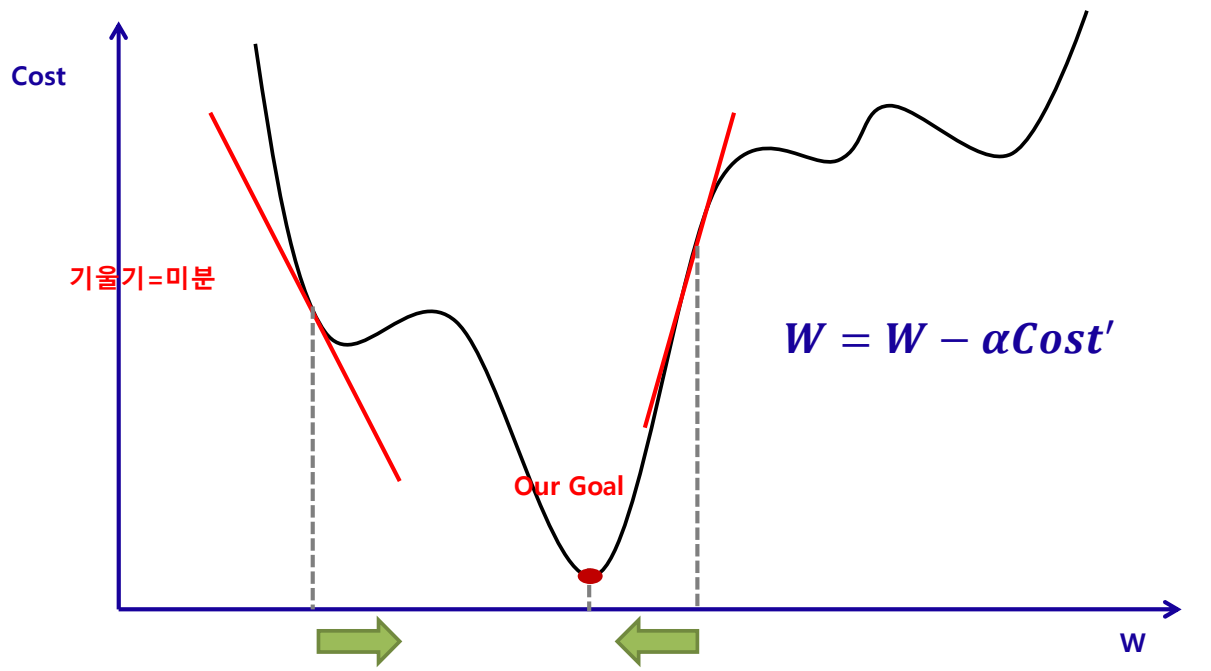
$$Cost(w) = -\frac{1}{m} \sum y \log(H(x)) + (1 - y) \log(1 - H(x))$$

$$W := W - \alpha \frac{\partial}{\partial W} cost(W)$$



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Goal of ML Models



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확인 문제

- 다음 학습 데이터와 비용 함수(cost function)가 주어지고, 초기 W값이 2이고 학습률이 0.1일 때, gradient decent 알고리즘에 의해 1회 학습 후 수정된 W 값을 구하시오.

[학습 데이터]

X (입력)	Y (출력)
1	1
2	3
3	5

[비용 함수]

$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx^{(i)} - y^{(i)})^2$$



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확인 문제

?



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Q&A

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