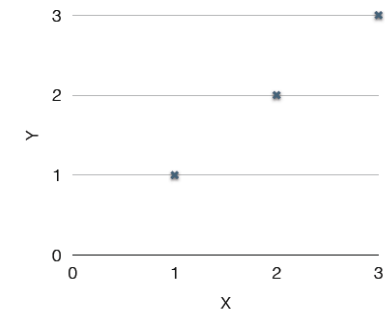


Concept of Machine Learning

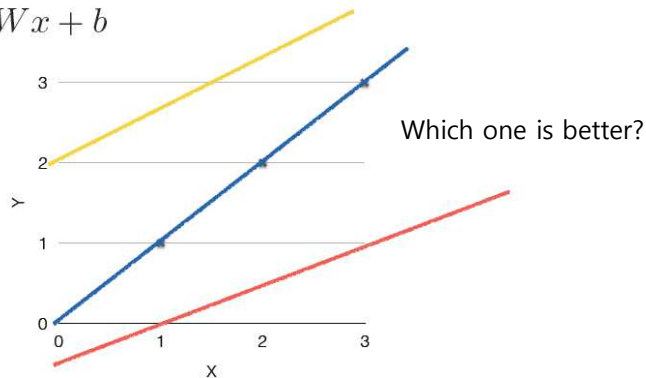
What is Learning?

x	Y
1	1
2	2
3	3



Linear Hypothesis

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



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

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

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

Matrix Representation

$$\begin{bmatrix} w1 & w2 & w3 \end{bmatrix} \times \begin{bmatrix} x1 \\ x2 \\ x3 \end{bmatrix} = \begin{bmatrix} w1 \times x1 + w2 \times x2 + w3 \times x3 \end{bmatrix}$$

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

With b vector

$$\begin{bmatrix} b & w1 & w2 & w3 \end{bmatrix} \times \begin{bmatrix} 1 \\ x1 \\ x2 \\ x3 \end{bmatrix} = \begin{bmatrix} b \times 1 + w1 \times x1 + w2 \times x2 + w3 \times x3 \end{bmatrix}$$

$$H(X) = WX$$

Without b vector

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

Transpose representation

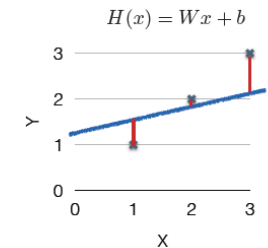


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Which hypothesis is better?

- How fit the line to our (training) data

$$H(x) - y$$

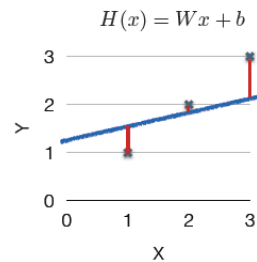


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Cost Function

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

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



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

Cost function을 최소로 하는 hypothesis가 무엇일까?



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Hypothesis and Cost

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

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



Simplifying without b vector

$$H(x) = Wx$$

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



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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

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

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

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

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

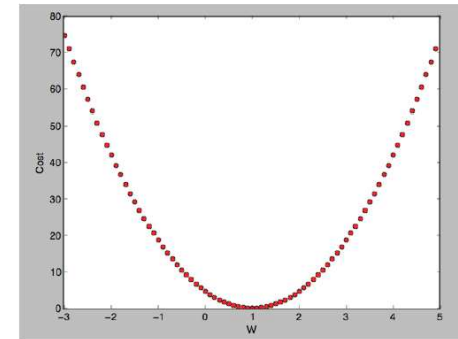
- $W=2$, $\text{cost}(W)=?$?



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What cost(W) looks like?

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

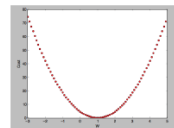


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How to Minimize Cost?

= 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 $\text{cost}(W, b)$
- Each time you change the parameters, you select the gradient which reduces $\text{cost}(W, b)$ 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

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

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

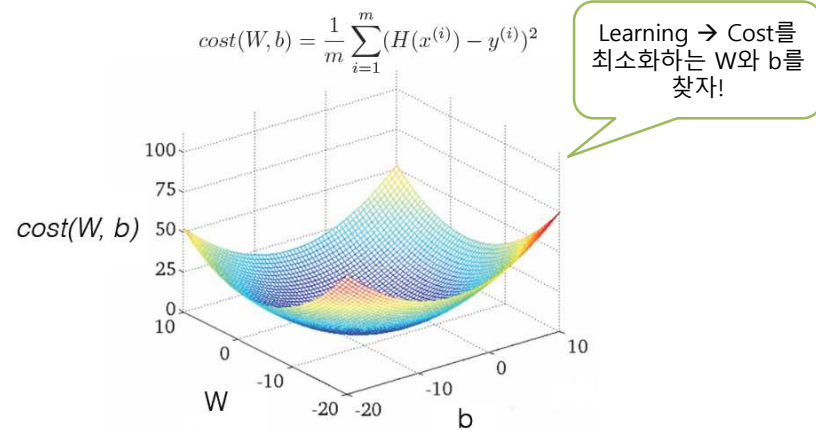
$$W := W - \alpha \frac{1}{2m} \sum_{i=1}^m 2(Wx^{(i)} - y^{(i)})x^{(i)} \quad \xleftarrow{\text{green arrow}} \quad W := W - \alpha \frac{\partial}{\partial W} \text{cost}(W)$$

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



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Convex Function

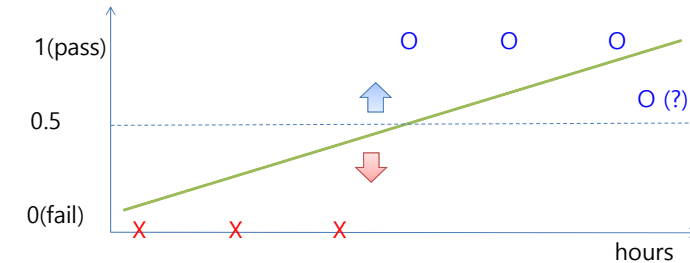


Regression to Classification

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)

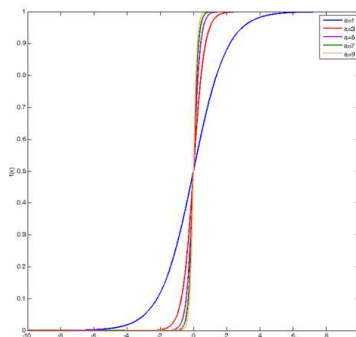
Pass/Fail based on study hours?



Logistic Hypothesis

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

WHY?
0과 1 사이 값으로 변환



Logistic Hypothesis & Cost Function

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

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

New Cost Function

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

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

H(x)=1일 때 C값은? ?

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

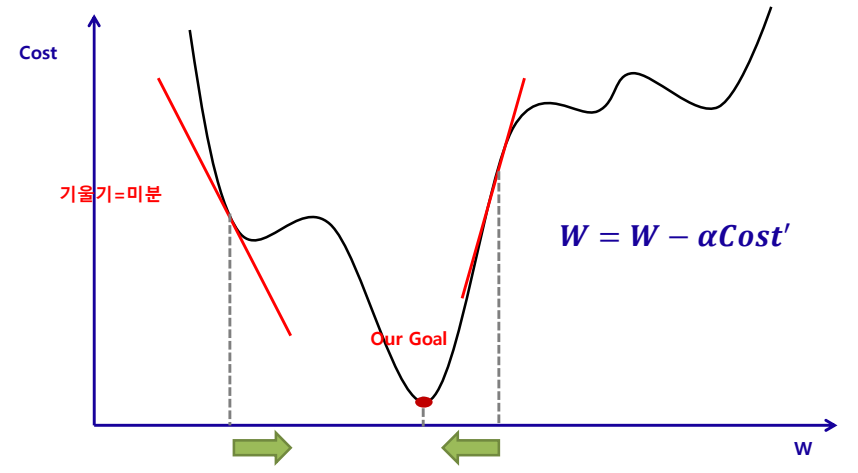
$$Cost(w) = -\frac{1}{m} \sum y \log(H(x)) + (1 - y) \log(1 - H(x))$$

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



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Goal of ML Models



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확인 문제

- 다음 학습 데이터와 비용 함수(cost function)가 주어지고, 초기 W 값이 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]

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



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확인 문제



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Q & A

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