

Lecture 2

Linear Regression

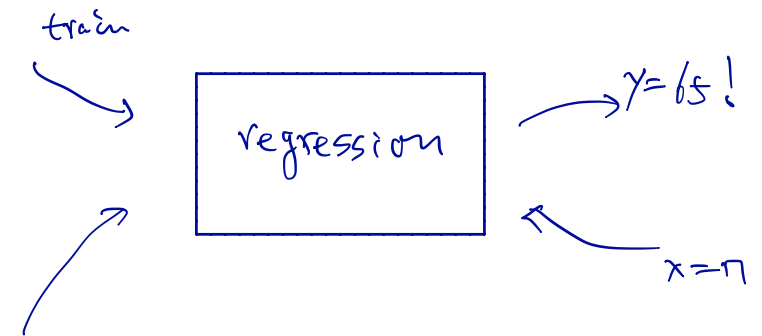
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Acknowledgement

- Andrew Ng's ML class
 - <https://class.coursera.org/ml-003/lecture>
 - <http://www.holehouse.org/mlclass/> (note)
- Convolutional Neural Networks for Visual Recognition.
 - <http://cs231n.github.io/>
- Tensorflow
 - <https://www.tensorflow.org>
 - <https://github.com/aymericdamien/TensorFlow-Examples>

Predicting exam score: regression

x (hours)	y (score)
10	90
9	80
3	50
2	30

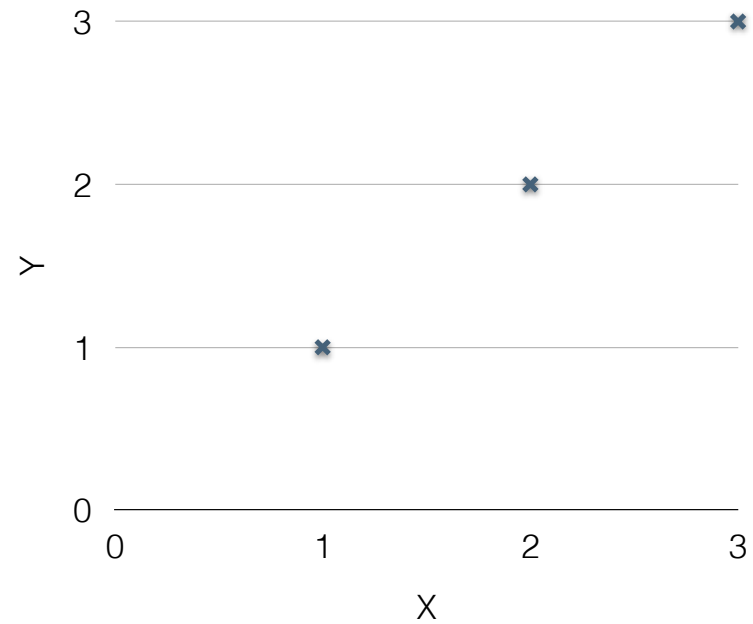


Regression (data)

x	y
1	1
2	2
3	3

Regression (presentation)

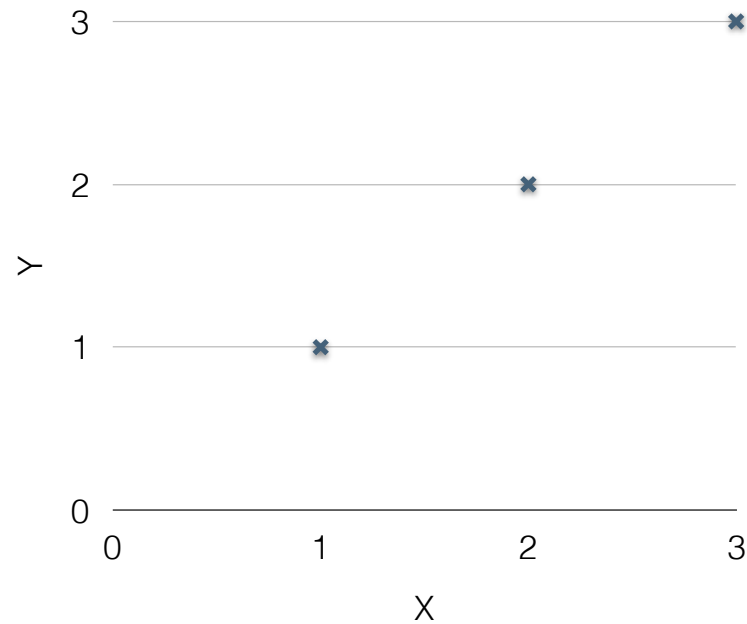
x	Y
1	1
2	2
3	3



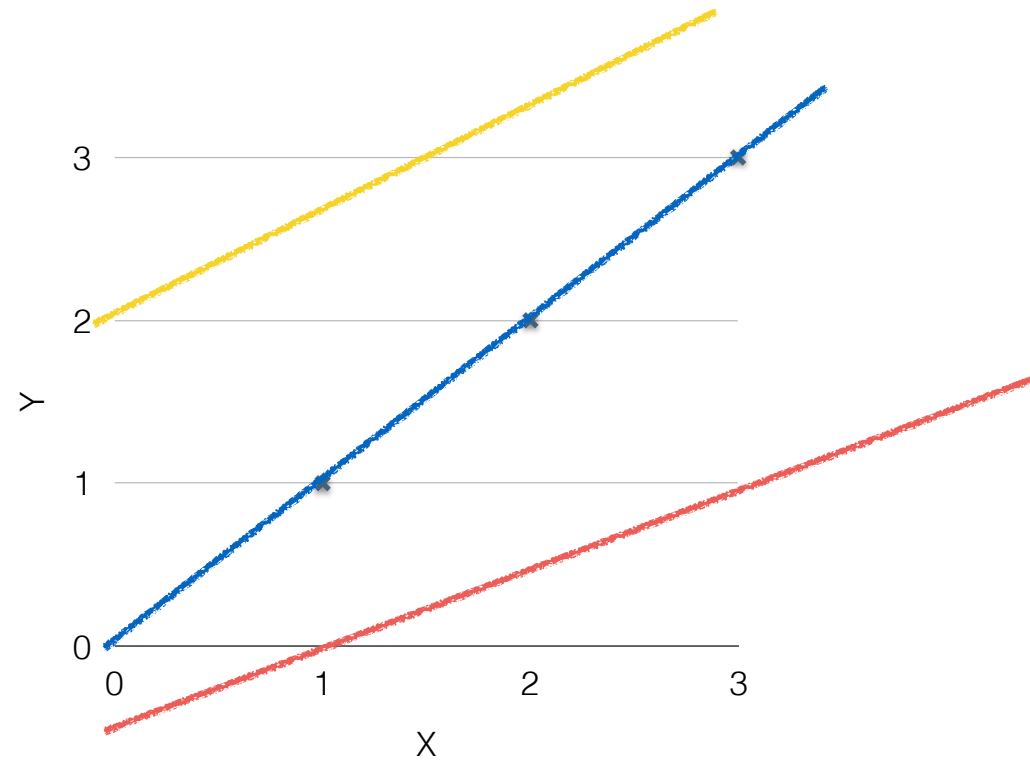
(Linear) Hypothesis

Linear regression
(선형 회귀)

모델을 가설로 사용하여
적절한 선을 찾기

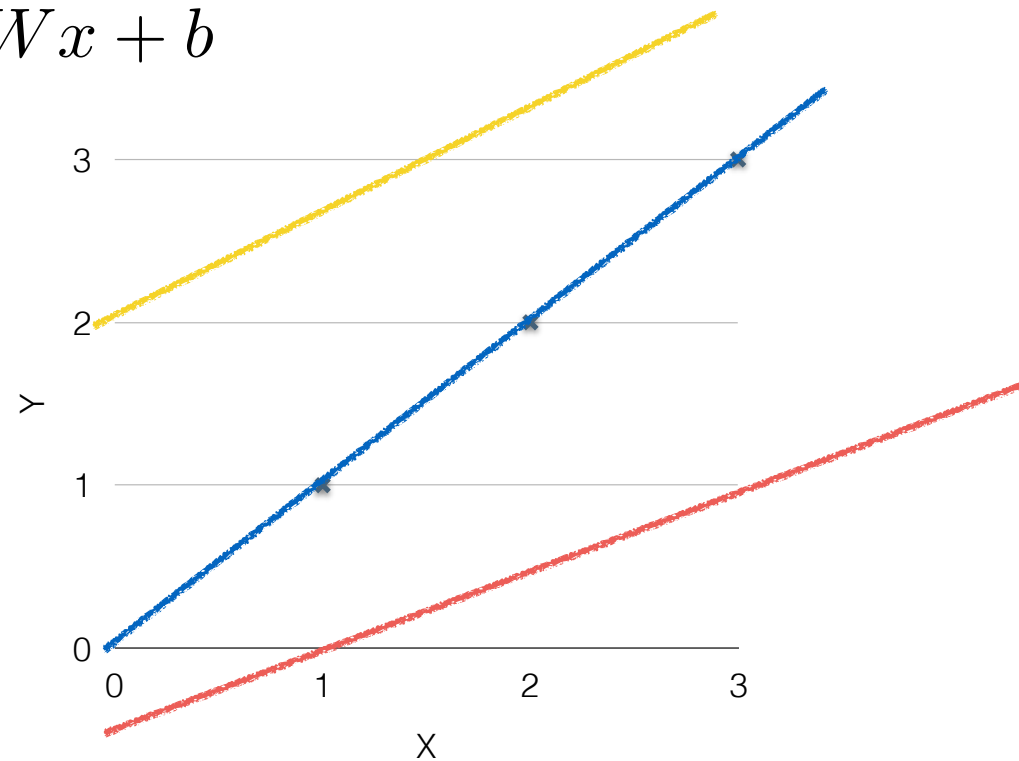


(Linear) Hypothesis

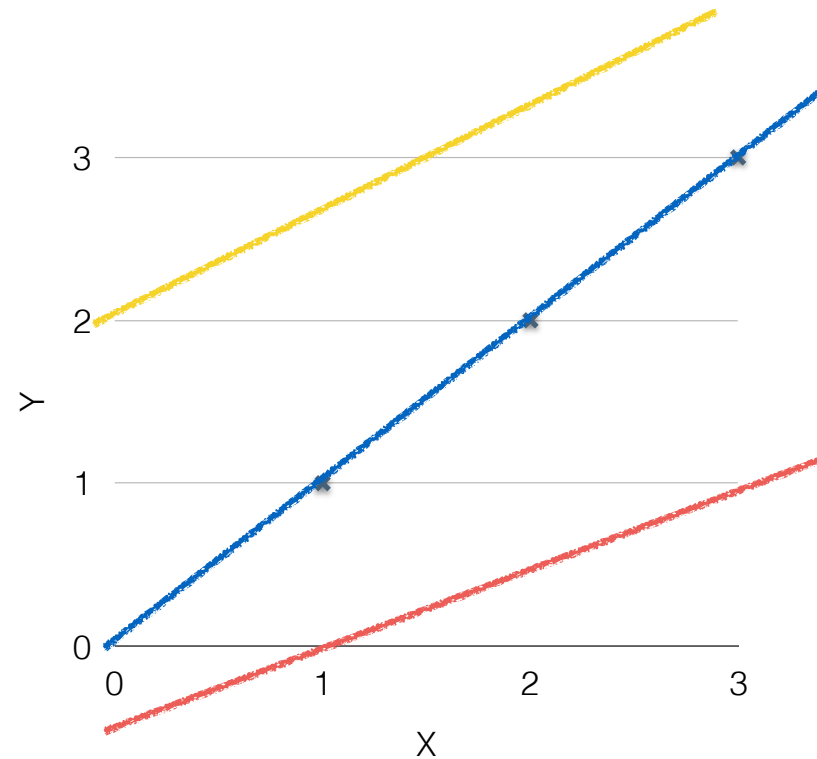


(Linear) Hypothesis

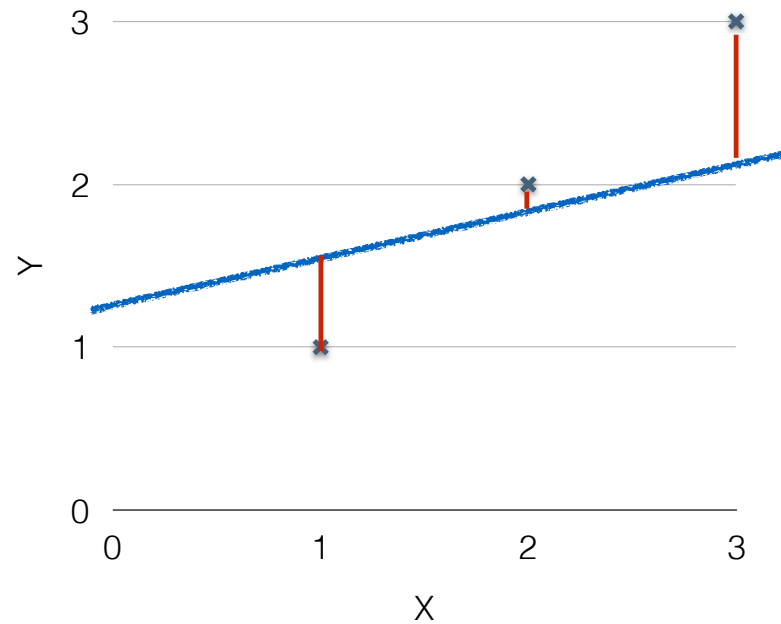
$$H(x) = Wx + b$$



Which hypothesis is better?



Which hypothesis is better?



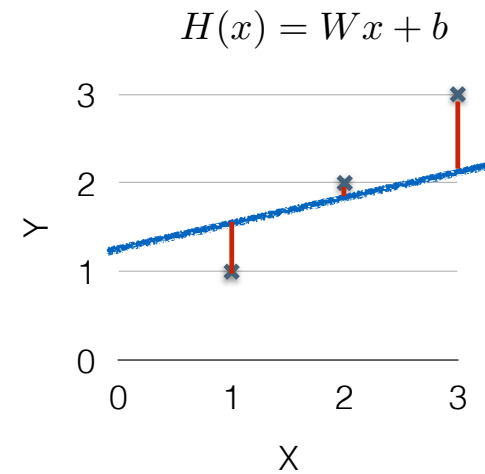
Cost function

- How fit the line to our (training) data

$$H(x) - y$$

↘ 대신에

$$(H(x) - y)^2$$



Cost function

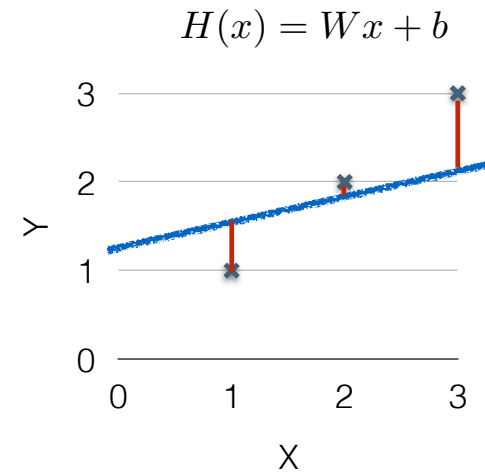
- How fit the line to our (training) data

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

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

Handwritten notes for the formula above:

- Red arrow pointing to the summation symbol \sum : 데이터 개수 (Number of data)
- Red arrow pointing to the term $(H(x^{(i)}) - y^{(i)})^2$: 가중치! (Weight!)
- Red arrow pointing to the denominator m : 차이의 제곱의 평균 (Average of the square of the difference)



Cost function

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

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

가중치 W와 편향 b에 의존!

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

Goal: Minimize cost

$\underset{W, b}{\text{minimize cost}}(W, b)$

선형 회귀의
목적 함수

