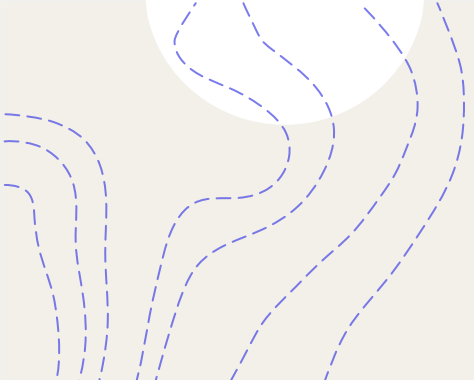




# Linear regression



train

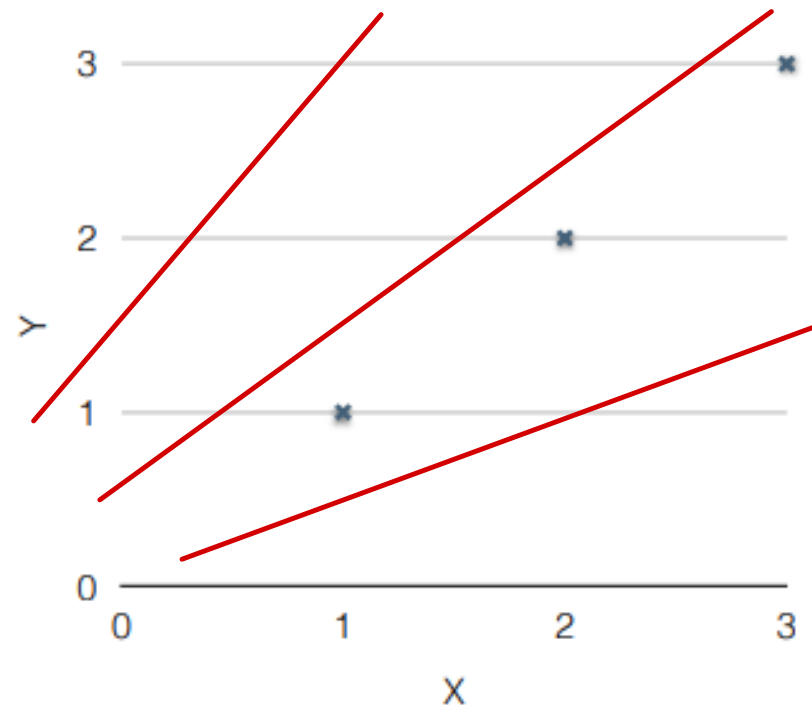
regression

$x=7$   $y=?$

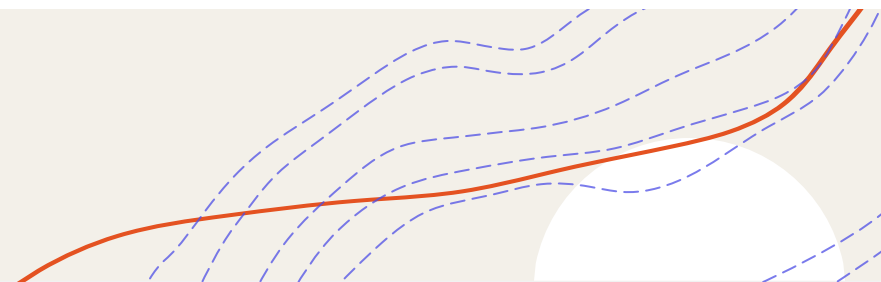
x (hours)	y (score)
10	90
9	80
3	50
2	30

x	Y
1	1
2	2
3	3

(Linear) Hypothesis



가설을 세워 linear하게 되는것이 맞을 것이라 판단

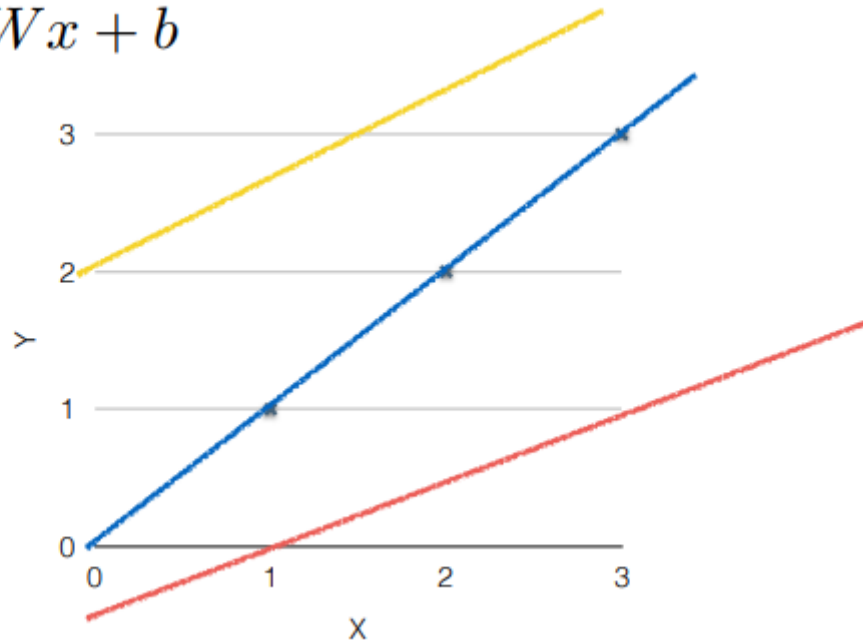


# COST FUNCTION

$(H(x) - y)^2$  → 거리가 음수, 양수 판단

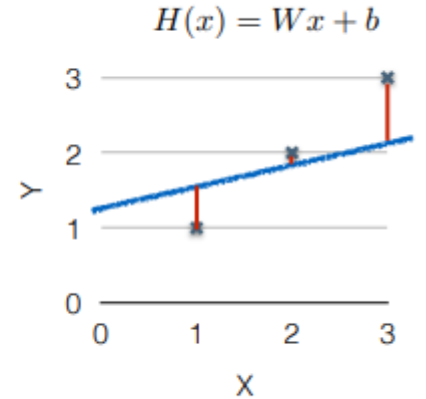
어떤 그래프가 가장 알맞을지?

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



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

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



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

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

→ 가장 작은 cost값을 구하는 것이 linear 학습

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

$$H(x) = Wx$$

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

# Gradient descent algorithm

(경사)

(내려감)

↳ 경사를 따라 내려가는 알고리즘.

Start: (0,0) (or any other value)

W를 바꾸며 Cost를 줄여나감

↳ 최저점 도달.

x	Y
1	1
2	2
3	3

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

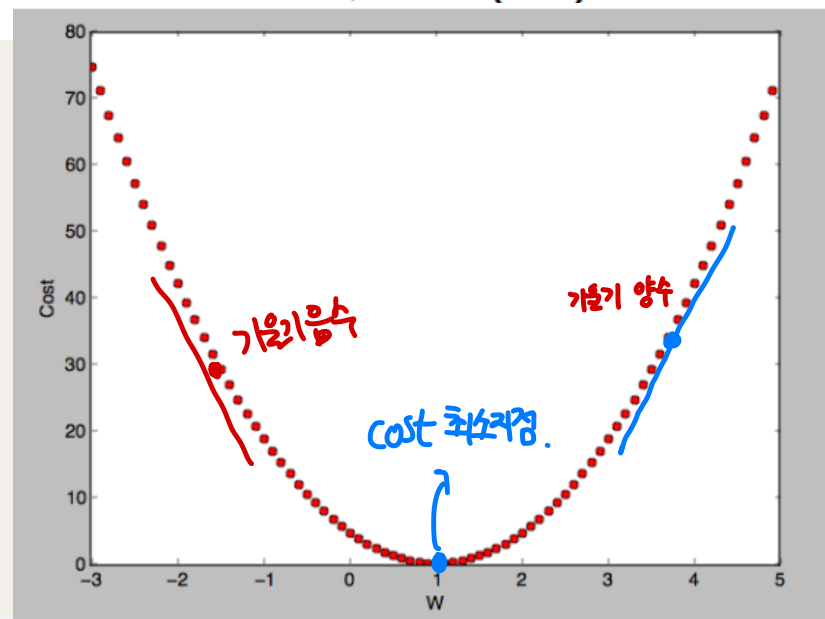
- $W=1, \text{cost}(W)=0$

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

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

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

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



# Formal definition

↪ 경사로 계산: 미분 → 기울기를 계산

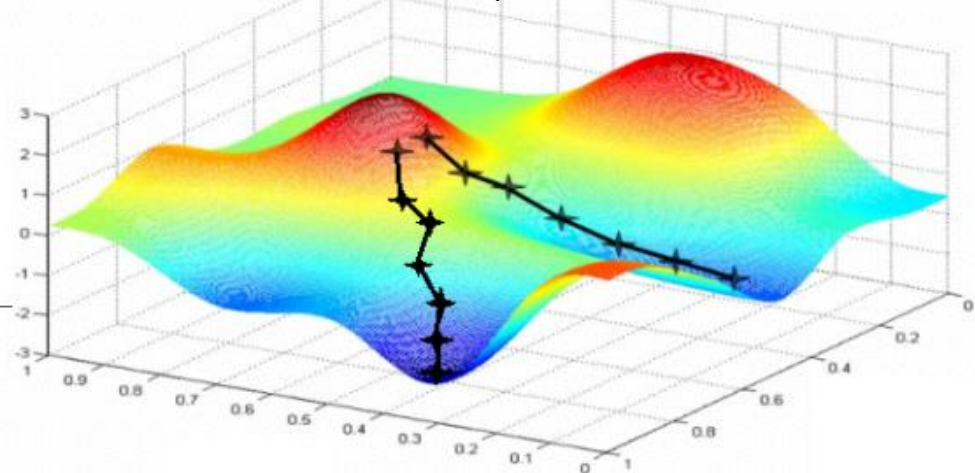
$$W := W - \alpha \frac{\partial}{\partial W} \frac{1}{2m} \sum_{i=1}^m (W x^{(i)} - y^{(i)})^2$$

↪ 미분을 편하게 하기 위해

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

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

↪ 경사로를 따라 내려가면  
도함수가 다름.



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

