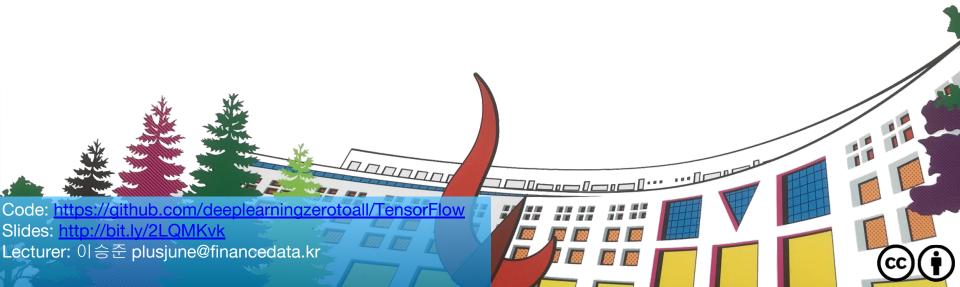
ML/DL for Everyone Season2



03 - How to minimize cost



Hypothesis and Cost

Hypothesis H(x) = Wx + b

Cost $cost(W,b) = rac{1}{m} \sum_{i=1}^m \left(H(x_i) - y_i
ight)^2$

Simplified hypothesis

Hypothesis H(x)=Wx

Cost $cost(W) = rac{1}{m} \sum_{i=1}^m (Wx_i - y_i)^2$

$$cost(W) = rac{1}{m} \sum_{i=1}^m {(Wx_i - y_i)^2}$$

• W = 0, cost(W) = ?

Х	У
1	1
2	2
3	3

$$cost(W) = rac{1}{m} \sum_{i=1}^m \left(Wx_i - y_i
ight)^2$$

X	У
1	1
2	2
3	3

• W = 0, cost(W) = 4.67

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

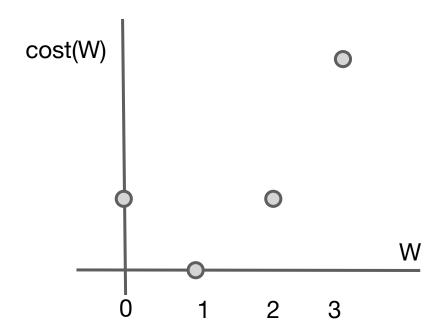
$$cost(W) = rac{1}{m} \sum_{i=1}^m (Wx_i - y_i)^2$$

Х	у
1	1
2	2
3	3

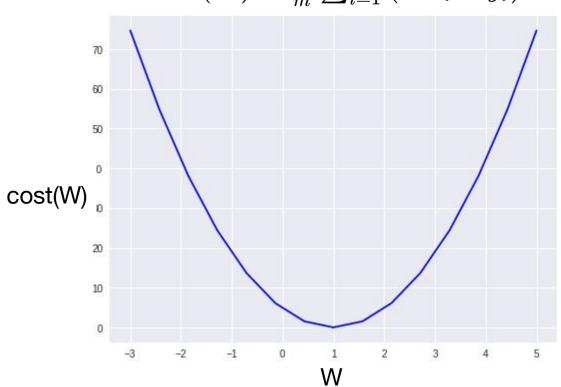
- W = 0, cost(W) = 4.67 $\frac{1}{3}((0*1-1)^2 + (0*2-2)^2 + (0*3-3)^2))$
- W = 1, cost(W) = 0 $\frac{1}{3}((1*1-1)^2 + (1*2-2)^2 + (1*3-3)^2))$
- W = 2, cost(W) = 4.67 $\frac{1}{3}((2*1-1)^2 + (2*2-2)^2 + (2*3-3)^2))$
- W = 3, cost(W) = 18.67 $\frac{1}{3}((3*1-1)^2 + (3*2-2)^2 + (3*3-3)^2))$

- W = 0, cost(W) = 4.67
- W = 1, cost(W) = 0
- W = 2, cost(W) = 4.67
- W = 3, cost(W) = 18.67

- W = 0, cost(W) = 4.67
- W = 1, cost(W) = 0
- W = 2, cost(W) = 4.67
- W = 3, cost(W) = 18.67

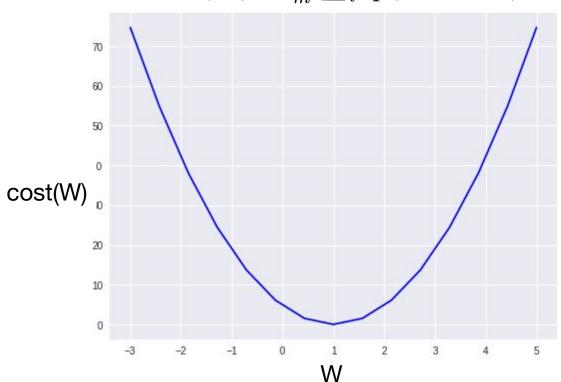


$$cost(W) = rac{1}{m} \sum_{i=1}^m {(Wx_i - y_i)^2}$$



How to minimize cost?

$$cost(W) = rac{1}{m} \sum_{i=1}^m \left(Wx_i - y_i
ight)^2$$

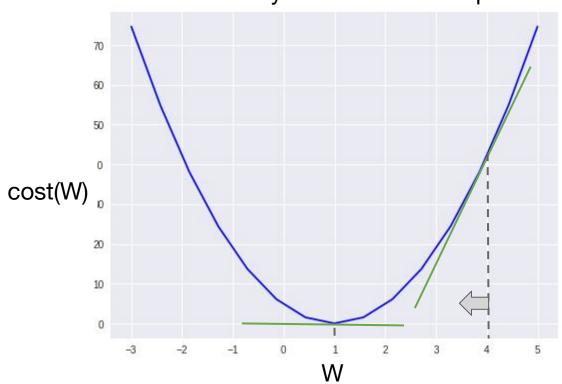


Gradient descent algorithm

- Minimize cost function
- Gradient descent is used many minimization problems
- For a given cost function, cost (W, b), it will find W, b to minimize cost
- It can be applied to more general function: cost (w1, w2, ...)

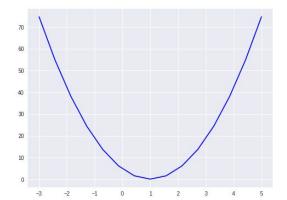
How it works?

How would you find the lowest point?

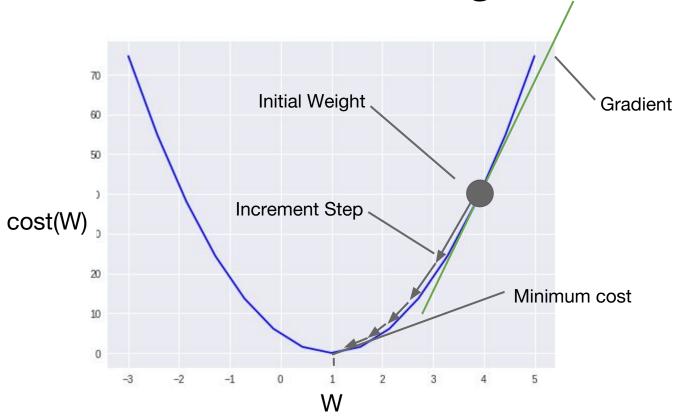


How it works?

- 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

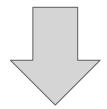


Gradient descent algorithm



Formal definition

$$cost(W,b) = rac{1}{m} \sum_{i=1}^m \left(H(x_i) - y_i
ight)^2$$



$$cost(W,b) = rac{1}{2m} \sum_{i=1}^m \left(H(x_i) - y_i
ight)^2$$

Formal definition

$$W := W - lpha rac{\partial}{\partial W} rac{1}{2m} \sum_{i=1}^m \left(W(x_i) - y_i
ight)^2$$

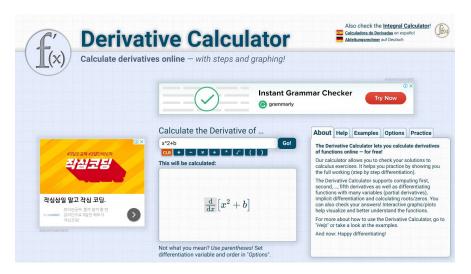
$$W := W - lpha rac{1}{2m} \sum_{i=1}^m 2(W(x_i) - y_i) x_i$$

$$W := W - lpha rac{1}{m} \sum_{i=1}^m (W(x_i) - y_i) x_i$$

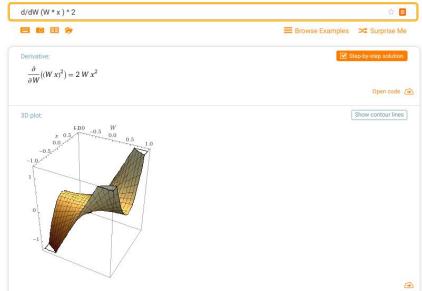
Formal definition

$$cost(W,b) = rac{1}{2m} \sum_{i=1}^m \left(H(x_i) - y_i
ight)^2$$

$$W:=W-lpharac{\partial}{\partial W}cost(W)$$



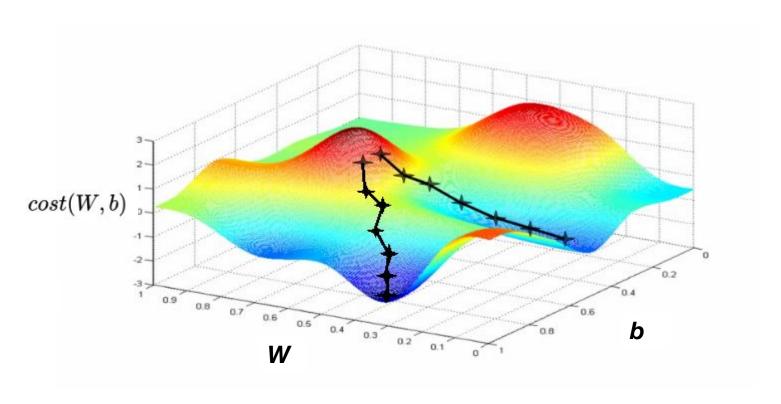
Wolfram Alpha computational intelligence.



Gradient descent algorithm

$$W := W - lpha rac{1}{m} \sum_{i=1}^m \left(W(x_i) - y_i
ight) x_i$$

Convex function



Convex function

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

-10

-10

-20

-20

W

cost(W, b)

What's Next?

• Multi-Variable Linear regression