

Linear Regression Cost 함수 최소화

$$H(x) = wx + b.$$

$$\frac{(H(x_1) - y(1))^2 + (H(x_2) - y(2))^2 + (H(x_3) - y(3))^2}{3}$$

\Rightarrow 일차화 $Cost = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2$, $H(x) = Wx + b$.

(Note: Red arrows point from the handwritten text above to $x^{(i)}$ and $y^{(i)}$ in the formula. The text above $x^{(i)}$ is "가설 값" (Hypothesis value) and the text above $y^{(i)}$ is "정답 labeled 값" (Correct labeled value).)

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

Goal: minimize $Cost(W, b)$

" W 와 b 를 최소화"

$$H(x) = wx$$

$$\text{Cost}(w) = \frac{1}{m} \sum_{z=1}^m (wx(z) - y(z))^2$$

(Min)

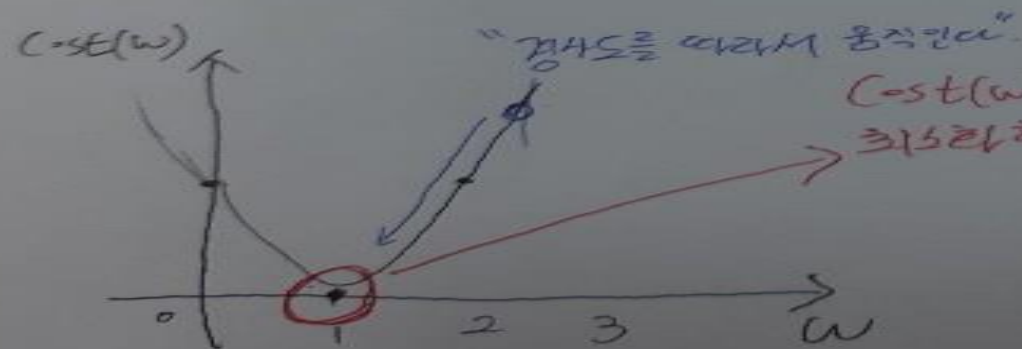
x	y
1	1
2	2
3	3

i) $w=1$, $\text{Cost}(w)=0$

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

ii) $w=0$, $\text{Cost}(w)=?$

$$\frac{(0 \times 1 - 1)^2 + (0 \times 2 - 2)^2 + (0 \times 3 - 3)^2}{3} = \frac{14}{3} = 4.67$$



\Rightarrow Gradient descent algorithm.

- 아무 점에서 시작.
- w, b 를 조금 바꿈.

Gradient descent algorithm

$$\text{Cost}(w) = \frac{1}{2m} \sum_{i=1}^m (w x(i) - y(i))^2$$

$$w := w - \overset{0.1}{\downarrow} \frac{\partial}{\partial w} \text{Cost}(w)$$

미분.

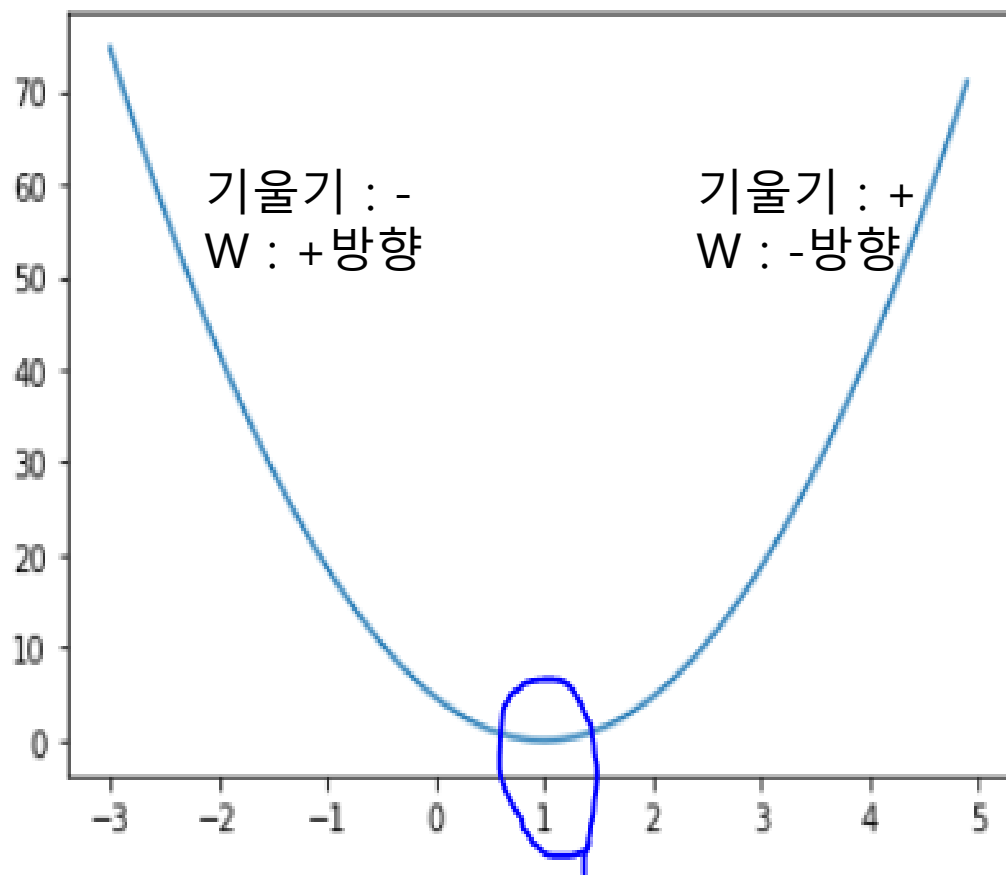
$$w := w - 2 \frac{1}{2m} \sum_{i=1}^m 2(w x(i) - y(i)) x(i)$$

$$\therefore w := w - \frac{1}{m} \sum_{i=1}^m (w x(i) - y(i)) x(i) //$$

Convex function

- Cost function이 convex function이 되어야 한다.
 - 어느 점에서 시작하더라도 항상 도착 지점이 원하는 지점이다.
 - 항상 답을 찾음을 보장함.

Tensorflow로 linear regression cost 최소화 구현



미분한 cost function

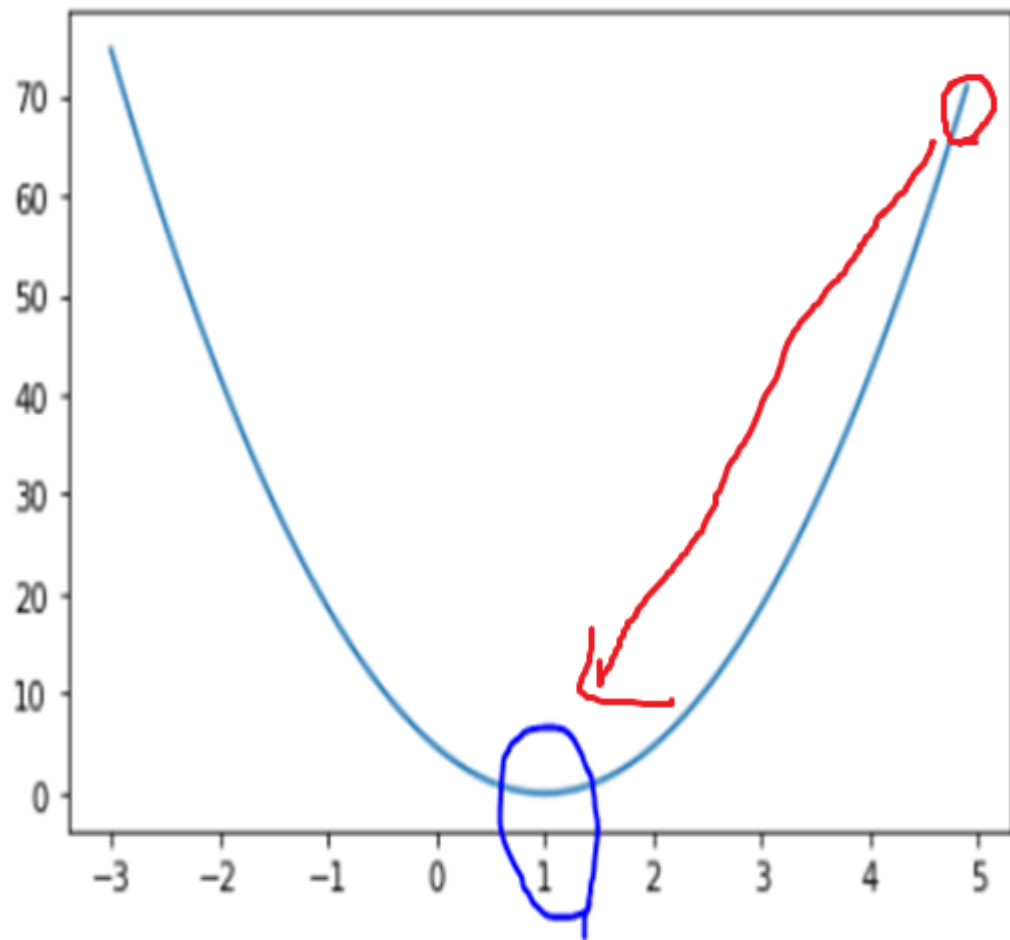
$$\sum_{i=1}^m (Wx(i) - y(i)) x(i)$$

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

```
# cost(loss) function
cost = tf.reduce_sum(tf.square(hypothesis - Y))

# gradient descent
learning_rate = 0.1
gradient = tf.reduce_mean((W*X - Y) * X)
descent = W - learning_rate * gradient
update = W.assign(descent)
```

Tensorflow로 linear regression cost 최소화 구현2



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

같은 표현

Part1 - Build graph using TF operations

X = [1,2,3]

Y = [1,2,3]

W = tf.Variable(5.0)

$H(x) = Wx + b$

hypothesis = W * X

cost(loss) function

cost = tf.reduce_mean(tf.square(hypothesis - Y))

gradient descent

optimizer = tf.train.GradientDescentOptimizer(learning_rate = 0.1)

train = optimizer.minimize(cost)

sess = tf.Session()

sess.run(tf.global_variables_initializer())

for step in range(100):

print(step, sess.run(W))

sess.run(train)