

ML/DL for Everyone Season2

with  TensorFlow

03 - How to minimize cost LAB

Code: <https://github.com/deeplearningzerotoall/TensorFlow>

Slides: <http://bit.ly/2LQMKvk>

Lecturer: 이승준 plusjune@financedata.kr



Simplified hypothesis

Hypothesis $H(x) = Wx$

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

Cost function in pure Python

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

```
import numpy as np
```

```
X = np.array([1, 2, 3])
```

```
Y = np.array([1, 2, 3])
```

```
def cost_func(W, X, Y):
```

```
    c = 0
```

```
    for i in range(len(X)):
```

```
        c += (W * X[i] - Y[i]) ** 2
```

```
    return c / len(X)
```

```
for feed_W in np.linspace(-3, 5, num=15):
```

```
    curr_cost = cost_func(feed_W, X, Y)
```

```
    print("{:6.3f} | {:.10.5f}".format(feed_W, curr_cost))
```

W	cost
-3.000	74.66667
-2.429	54.85714
-1.857	38.09524
-1.286	24.38095
-0.714	13.71429
-0.143	6.09524
0.429	1.52381
1.000	0.00000
1.571	1.52381
2.143	6.09524
2.714	13.71429
3.286	24.38095
3.857	38.09524
4.429	54.85714
5.000	74.66667

Cost function in pure Python

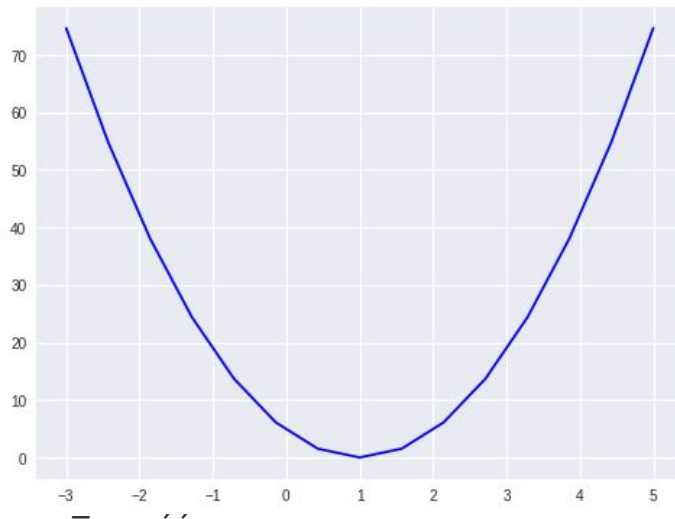
$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx_i - y_i)^2$$

```
import numpy as np

X = np.array([1, 2, 3])
Y = np.array([1, 2, 3])

def cost_func(W, X, Y):
    c = 0
    for i in range(len(X)):
        c += (W * X[i] - Y[i]) ** 2
    return c / len(X)

for feed_W in np.linspace(-3, 5, num=15):
    curr_cost = cost_func(feed_W, X, Y)
    print("{:6.3f} | {:10.5f}".format(feed_W,
```



Cost function in TensorFlow

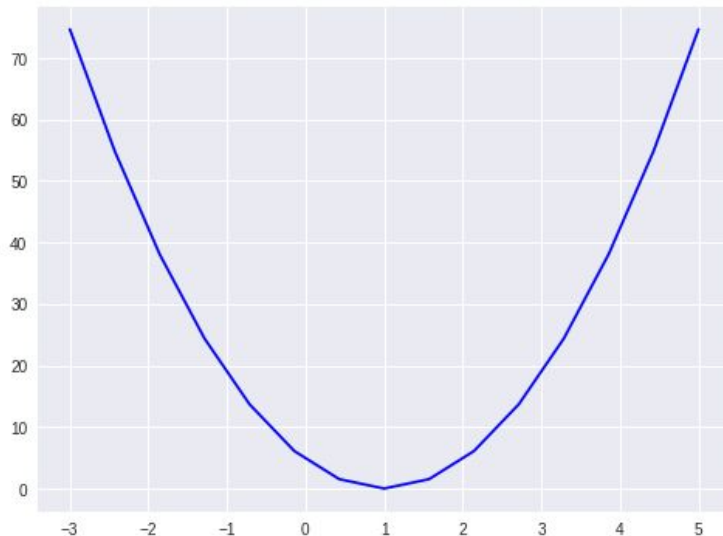
$$\text{cost}(W) = \frac{1}{m} \sum_{i=1}^m (Wx_i - y_i)^2$$

```
X = np.array([1, 2, 3])
Y = np.array([1, 2, 3])

def cost_func(W, X, Y):
    hypothesis = X * W
    return tf.reduce_mean(tf.square(hypothesis - Y))

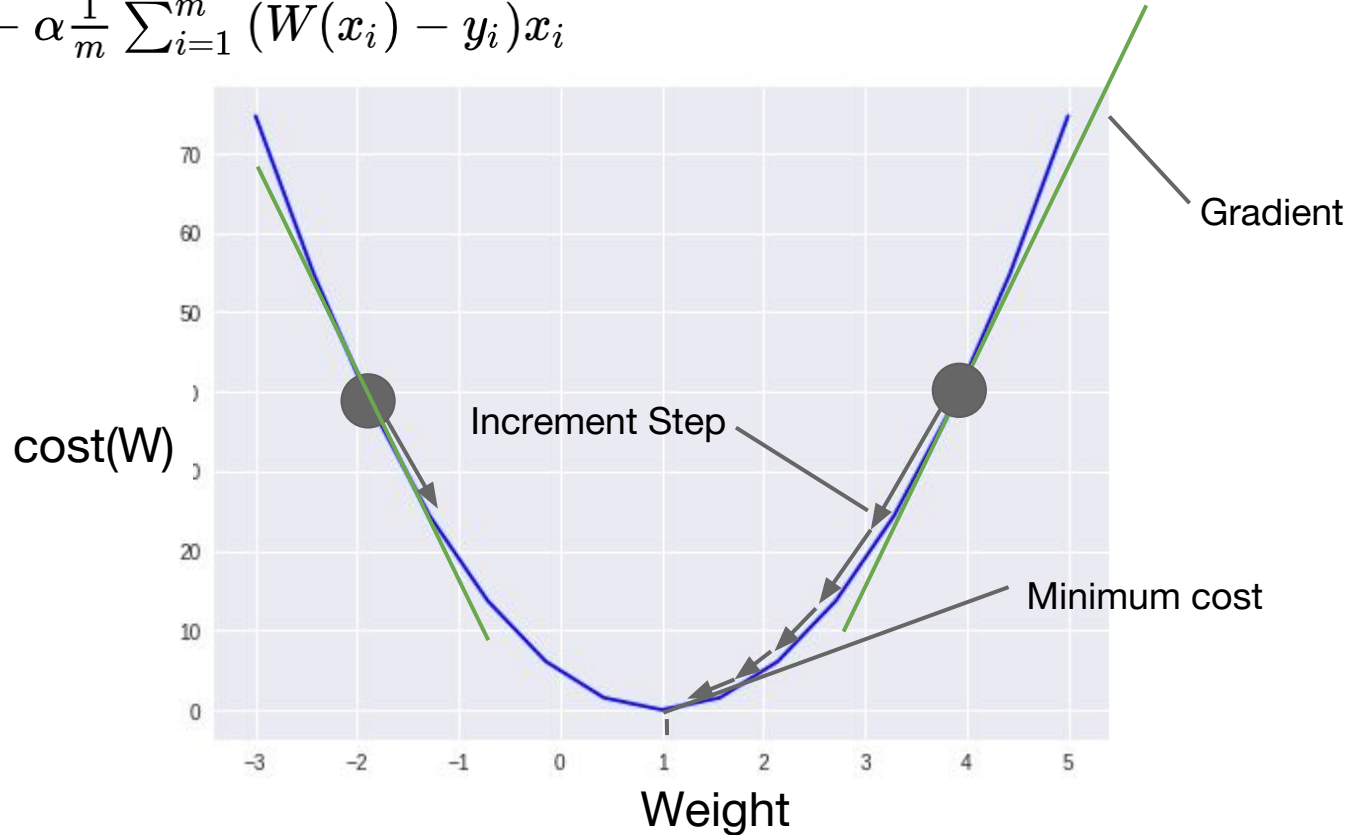
W_values = np.linspace(-3, 5, num=15)
cost_values = []

for feed_W in W_values:
    curr_cost = cost_func(feed_W, X, Y)
    cost_values.append(curr_cost)
    print("{:6.3f} | {:10.5f}".format(feed_W, curr_cost))
```




Gradient descent

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



Gradient descent

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


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

```
alpha = 0.01
gradient = tf.reduce_mean(tf.multiply(tf.multiply(W, X) - Y, X))
descent = W - tf.multiply(alpha, gradient)
W.assign(descent)
```

Gradient descent

```
tf.set_random_seed(0) # for reproducibility

x_data = [1., 2., 3., 4.]
y_data = [1., 3., 5., 7.]

W = tf.Variable(tf.random_normal([1], -100., 100.))

for step in range(300):
    hypothesis = W * X
    cost = tf.reduce_mean(tf.square(hypothesis - Y))

    alpha = 0.01
    gradient = tf.reduce_mean(tf.multiply(tf.multiply(W, X) - Y, X))
    descent = W - tf.multiply(alpha, gradient)
    W.assign(descent)

    if step % 10 == 0:
        print('{:5} | {:10.4f} | {:10.6f}'.format(
            step, cost.numpy(), W.numpy()[0]))
```

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

Gradient descent

```
tf.set_random_seed(0) # for reproducibility
```

```
x_data = [1., 2., 3., 4.]
```

```
y_data = [1., 3., 5., 7.]
```

```
W = tf.Variable(tf.random_normal([1], -100., 100.))
```

```
for step in range(300):
```

```
    hypothesis = W * X
```

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

```
    alpha = 0.01
```

```
    gradient = tf.reduce_mean(tf.multiply(tf.multiply(W, X) - Y, X)
```

```
    descent = W - tf.multiply(alpha, gradient)
```

```
    W.assign(descent)
```

```
if step % 10 == 0:
```

```
    print('{:5} | {:.10.4f} | {:.10.6f}'.format(  
        step, cost.numpy(), W.numpy()[0]))
```

step	cost	W
0	11716.3086	48.767971
10	4504.9126	30.619968
20	1732.1364	19.366755
30	666.0052	12.388859
40	256.0785	8.062004
50	98.4620	5.379007
60	37.8586	3.715335
70	14.5566	2.683725
80	5.5970	2.044044
...
240	0.0000	1.000499
250	0.0000	1.000309
260	0.0000	1.000192
270	0.0000	1.000119
280	0.0000	1.000074
290	0.0000	1.000046

Output when $W=5$

```
tf.set_random_seed(0) # for reproducibility
```

```
x_data = [1., 2., 3., 4.]
```

```
y_data = [1., 3., 5., 7.]
```

```
W = tf.Variable([5.0])
```

```
for step in range(300):
```

```
    hypothesis = W * X
```

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

```
    alpha = 0.01
```

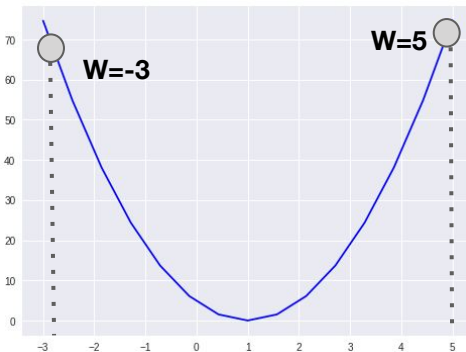
```
    gradient = tf.reduce_mean(tf.multiply(tf.multiply(W, X) - Y, X)
```

```
    descent = W - tf.multiply(alpha, gradient)
```

```
    W.assign(descent)
```

```
    if step % 10 == 0:
```

```
        print('{:5} | {:.10.4f} | {:.10.6f}'.format(
            step, cost.numpy(), W.numpy()[0]))
```



step	cost	W
0	74.6667	4.813334
10	28.7093	3.364572
20	11.0387	2.466224
30	4.2444	1.909177
40	1.6320	1.563762
50	0.6275	1.349578
60	0.2413	1.216766
70	0.0928	1.134412
80	0.0357	1.083346
.....		
270	0.0000	1.000009
280	0.0000	1.000006
290	0.0000	1.000004

What's Next?

- Multi-Variable Linear regression