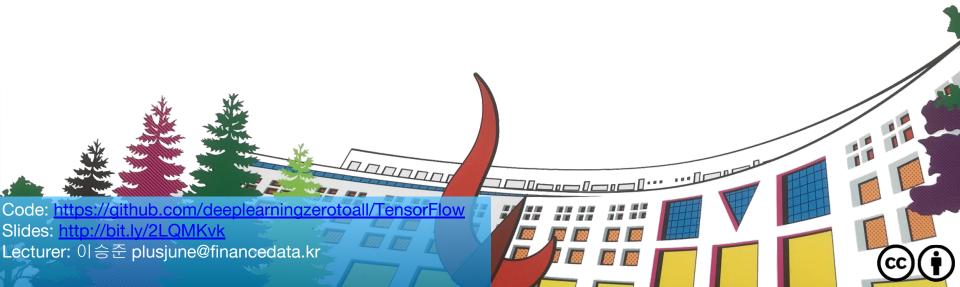
ML/DL for Everyone Season2



02 - Simple Linear Regression LAB



Hypothesis and Cost

Hypothesis

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

Cost

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

Build hypothesis and cost

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

```
x_data = [1, 2, 3, 4, 5]
y_data = [1, 2, 3, 4, 5]

W = tf.Variable(2.9)
b = tf.Variable(0.5)

# hypothesis = W * x + b
hypothesis = W * x_data + b
```

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

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

Build hypothesis and cost

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

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

tf.reduce_mean()

```
v = [1., 2., 3., 4.]
tf.reduce_mean(v) # 2.5
```

tf.square()

```
tf.square(3) # 9
```

Gradient descent

 $minimize \ cost(W,b)$

```
# learning rate initialize
learning rate = 0.01
# Gradient_descent
GradientTape()은 보통 with 구문과 같이 쓰이는데, with 구문 안의 변수들의 변화(기록)를 tape에 기록
with tf.GradientTape() as tape:
     hypothesis = W * x_data + b 여기서는 W와 b에
     cost = tf.reduce mean(tf.square(hypothesis - y data))
이후에 tape의 gradient 메서드를 호출해서. 경사도(미분값)을 구한다
W_grad, b_grad = tape.gradient(cost, [W, b])
gradient(cost, [W, b])은 함수 cost의 변수 W, b에 대한 개별 미분값(기<u>울기)을.</u> 구해서 tuple로 반환
W.assign sub(learning rate * W grad)
                                                                               A.assign sub(B)
b.assign sub(learning rate * b grad)
                                                                               A = A - B
                                                                               A -= B
```

Parameter(W,b) Update

```
W = tf.Variable(2.9)
b = tf.Variable(0.5)
for i in range(100):
   # Gradient descent
   with tf.GradientTape() as tape:
        hypothesis = W * x data + b
        cost = tf.reduce mean(tf.square(hypothesis - y data))
   W grad, b grad = tape.gradient(cost, [W, b])
   W.assign sub(learning rate * W grad)
   b.assign sub(learning rate * b grad)
   if i % 10 == 0:
      print("{:5}|{:10.4}|{:10.4}|{:10.6f}\".format(i, W.numpy(), b.numpy(), cost))
```

```
Full Code (less than 20 lines)
import tensorflow as tf
tf.enable eager execution()
# Data
x data = [1, 2, 3, 4, 5]
y_{data} = [1, 2, 3, 4, 5]
# W. b initialize
W = tf.Variable(2.9)
b = tf.Variable(0.5)
learning rate = 0.01
for i in range(100+1): # W, b update
   # Gradient descent
   with tf.GradientTape() as tape:
       hypothesis = W * x data + b
       cost = tf.reduce_mean(tf.square(hypothesis - y_data))
   W grad, b grad = tape.gradient(cost, [W, b])
   W.assign sub(learning rate * W grad)
   b.assign sub(learning rate * b grad)
   if i % 10 == 0:
```

print("{:5}|{:10.4f}|{:10.4f}|{:10.6f}".format(i, W.numpy(), b.numpy(), cost))

```
cost
      2.4520| 0.3760| 45.660004
10| 1.1036| 0.0034| 0.206336
201
     1.0128|
               -0.0209| 0.001026
301
     1.00651
               -0.0218| 0.000093
401
      1.00591
               -0.0212| 0.000083
50 I
     1.0057|
               -0.02051 0.000077
601
     1.0055|
               -0.0198| 0.000072
701
     1.0053|
               -0.0192 \mid 0.000067
801
     1.0051|
               -0.0185|
                        0.000063
```

-0.0179|

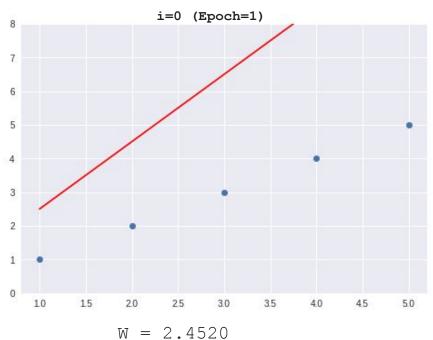
0.000059

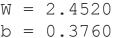
1.0050|

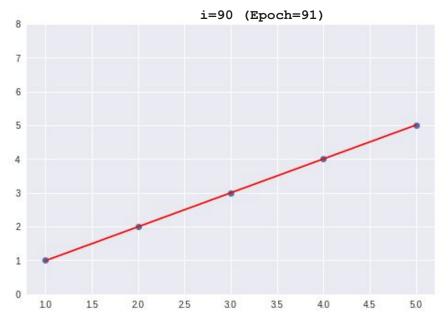
901

Training

W = tf.Variable(2.9) b = tf.Variable(0.5)







$$W = 1.0050$$

b = -0.0179

Predict

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

```
# Data
x_data = [1, 2, 3, 4, 5]
y_data = [1, 2, 3, 4, 5]
```

```
print(W * 5 + b)
print(W * 2.5 + b)
```

```
tf.Tensor(5.0066934, shape=(), dtype=float32)
tf.Tensor(2.4946523, shape=(), dtype=float32)
```

What's Next?

How to minimize cost