

Minimizing Cost

Lecture 03

Simplified hypothesis

$$H(x) = Wx$$

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

```
import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]
```

matplotlib

<http://matplotlib.org/users/installing.html>

```
W = tf.placeholder(tf.float32)
```

*# Our hypothesis for linear model X * w*

```
hypothesis = X * W
```

$$H(x) = Wx$$

cost/loss function

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

Launch the graph in a session.

```
sess = tf.Session()
```

Initializes global variables in the graph.

```
sess.run(tf.global_variables_initializer())
```

Variables for plotting cost function

```
W_val = []
```

```
cost_val = []
```

```
for i in range(-30, 50):
```

```
    feed_W = i * 0.1
```

```
    curr_cost, curr_W = sess.run([cost, W], feed_dict={W: feed_W})
```

```
    W_val.append(curr_W)
```

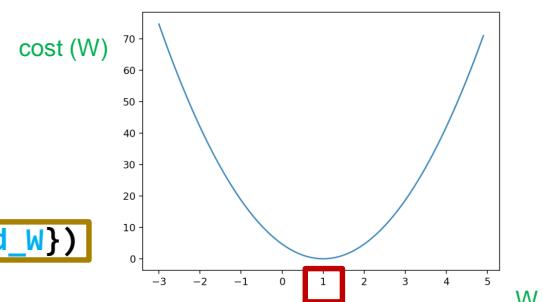
```
    cost_val.append(curr_cost)
```

Show the cost function

```
plt.plot(W_val, cost_val)
```

```
plt.show()
```

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



```

import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]

W = tf.placeholder(tf.float32)
# Our hypothesis for linear model X * W
hypothesis = X * W

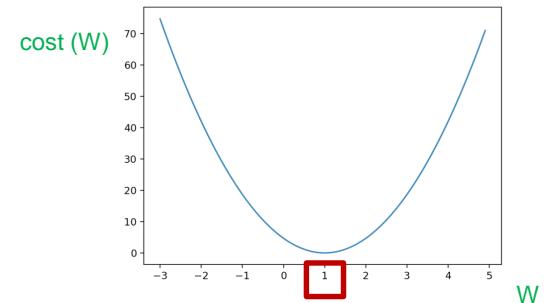
# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))
# Launch the graph in a session.
sess = tf.Session()
# Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())
# Variables for plotting cost function
W_val = []
cost_val = []
for i in range(-30, 50):
    feed_W = i * 0.1
    curr_cost, curr_W = sess.run([cost, W], feed_dict={W: feed_W})
    W_val.append(curr_W)
    cost_val.append(curr_cost)

# Show the cost function
plt.plot(W_val, cost_val)
plt.show()

```

X	1	2	3
Y	1	2	3
W		-3	
hypothesis = X * W	-3	-6	-9
hypothesis - Y	-4	-8	-12
tf.square(hypothesis - Y)	16	64	144
tf.reduce_mean(tf.square(hypothesis - Y))	74.667		

→ sum : 224
 average :
 $(\text{sum}/3)$ 74.667



```

import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]

W = tf.placeholder(tf.float32)
# Our hypothesis for linear regression
hypothesis = X * W

# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))

# Launch the graph in a session
sess = tf.Session()
# Initializes global variables in the graph
sess.run(tf.global_variables_initializer())

# Variables for plotting
W_val = []
cost_val = []
for i in range(-30, 50):
    feed_W = i * 0.1
    print("feed_W : %17.16f" %feed_W, end=" ")

    curr_cost, curr_W = sess.run([cost, W], feed_dict={W: feed_W})
    print("\t curr_cost : %8.6f" %curr_cost, "\t curr_W :", curr_W)

    W_val.append(curr_W)
    cost_val.append(curr_cost)

# Show the cost function
plt.plot(W_val, cost_val)
plt.show()

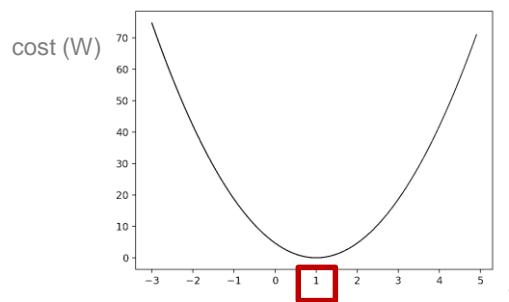
```

feed_W : -3.00000000000000000000
 feed_W : -2.90000000000000000004
 feed_W : -2.80000000000000000003
 feed_W : -2.70000000000000000002
 feed_W : -2.60000000000000000001
 .
 .
 .
 feed_W : 0.80000000000000000000
 feed_W : 0.90000000000000000000
 feed_W : 1.00000000000000000000
 feed_W : 1.10000000000000000001
 feed_W : 1.20000000000000000002
 feed_W : 1.30000000000000000000
 .
 .
 .
 feed_W : 4.50000000000000000000
 feed_W : 4.60000000000000000005
 feed_W : 4.70000000000000000002
 feed_W : 4.80000000000000000007
 feed_W : 4.90000000000000000004

curr_cost : 74.666664
 curr_cost : 70.980003
 curr_cost : 67.386665
 curr_cost : 63.886669
 curr_cost : 60.479992
 curr_W : -3.0
 curr_W : -2.9
 curr_W : -2.8
 curr_W : -2.7
 curr_W : -2.6
 curr_W : 0.8
 curr_W : 0.9
 curr_W : 1.0
 curr_W : 1.1
 curr_W : 1.2
 curr_W : 1.3
 curr_W : 4.5
 curr_W : 4.6
 curr_W : 4.7
 curr_W : 4.8
 curr_W : 4.9

print("feed_W : %17.16f" %feed_W, end=" ")

print("\t curr_cost : %8.6f" %curr_cost, "\t curr_W :", curr_W)



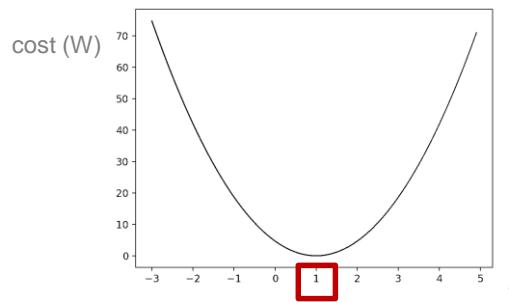
```

import tensorflow as tf
import math
X = [1, 2, 3, 4, 5]
Y = [1, 2, 3, 4, 5]
W_val 출력
<class 'list'>
W = tf.placeholder(tf.float32)
# Our hypothesis
hypothesis = W * X + Y
# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))
# Launch the graph in a session.
sess = tf.Session()
# Initialize the variables.
sess.run(tf.global_variables_initializer())
# Variables
W_val = []
cost_val = []
for i in range(100):
    feed_W = {W: i}
    curr_W = sess.run(W, feed_dict=feed_W)
    print(curr_W)
    print(cost_val.append(sess.run(cost, feed_dict=feed_W)))
    W_val.append(curr_W)
    cost_val.append(sess.run(cost, feed_dict=feed_W))

print("\nW_val 출력 ")
print(type(W_val), end="")
for i in range(len(W_val)):
    if (i%5==0):
        print("\n", end="")
    print("\t[",i,"]:", W_val[i], end="")

# Show the cost function
plt.plot(W_val, cost_val)
plt.show()

```



```

print ("WnW_val 출력 ")
print(type (W_val) , end="" )
for i in range(len(W_val)):
    if (i%5==0):
        print("Wn", end="")
    print ("Wt[",i,"]:" , W_val[i], end="")

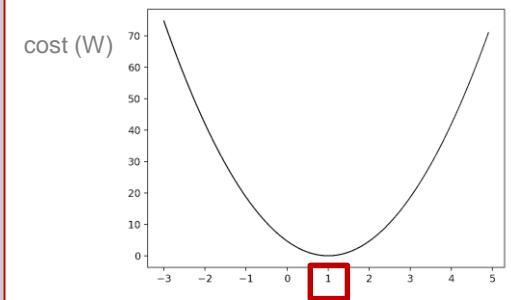
# Show the cost function
plt.plot(W_val, cost_val)
plt.show()

```

```
import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]
```

cost_val 출력
<class 'list'>

```
[ 0 ]: 74.666664      [ 1 ]: 70.980003      [ 2 ]: 67.386665      [ 3 ]: 63.886669      [ 4 ]: 60.479992
[ 5 ]: 57.166668      [ 6 ]: 53.946674      [ 7 ]: 50.819996      [ 8 ]: 47.786671      [ 9 ]: 44.846661
[ 10 ]: 42.000000     [ 11 ]: 39.246666     [ 12 ]: 36.586662     [ 13 ]: 34.020004     [ 14 ]: 31.546667
[ 15 ]: 29.166666     [ 16 ]: 26.879999     [ 17 ]: 24.686666     [ 18 ]: 22.586670     [ 19 ]: 20.580000
[ 20 ]: 18.666666     [ 21 ]: 16.846666     [ 22 ]: 15.120000     [ 23 ]: 13.486667     [ 24 ]: 11.946668
[ 25 ]: 10.500000     [ 26 ]: 9.146666     [ 27 ]: 7.886667     [ 28 ]: 6.720000     [ 29 ]: 5.646666
[ 30 ]: 4.666667     [ 31 ]: 3.780000     [ 32 ]: 2.986667     [ 33 ]: 2.286666     [ 34 ]: 1.680000
[ 35 ]: 1.166667     [ 36 ]: 0.746667     [ 37 ]: 0.420000     [ 38 ]: 0.186667     [ 39 ]: 0.046667
[ 40 ]: 0.000000     [ 41 ]: 0.046667     [ 42 ]: 0.186667     [ 43 ]: 0.420000     [ 44 ]: 0.746666
[ 45 ]: 1.166667     [ 46 ]: 1.680000     [ 47 ]: 2.286667     [ 48 ]: 2.986666     [ 49 ]: 3.779999
[ 50 ]: 4.666667     [ 51 ]: 5.646666     [ 52 ]: 6.720001     [ 53 ]: 7.886665     [ 54 ]: 9.146668
[ 55 ]: 10.500000    [ 56 ]: 11.946666    [ 57 ]: 13.486669    [ 58 ]: 15.119998    [ 59 ]: 16.846670
[ 60 ]: 18.666666    [ 61 ]: 20.579996    [ 62 ]: 22.586670    [ 63 ]: 24.686666    [ 64 ]: 26.880005
[ 65 ]: 29.166666    [ 66 ]: 31.546661    [ 67 ]: 34.020004    [ 68 ]: 36.586662    [ 69 ]: 39.246670
[ 70 ]: 42.000000    [ 71 ]: 44.846661    [ 72 ]: 47.786663    [ 73 ]: 50.820007    [ 74 ]: 53.946674
[ 75 ]: 57.166668    [ 76 ]: 60.479992    [ 77 ]: 63.886658    [ 78 ]: 67.386665    [ 79 ]: 70.980003
```



```
print ("\nW_val 출력 ")
print(type (W_val) , end="")
for i in range(len(W_val)):
    if (i%5==0):
        print("\n", end="")
    print ("\\t[" ,i,"]:" , W_val[i], end=" ")

print ("\ncost_val 출력 ")
print(type (cost_val) , end="")
for i in range(len(cost_val)):
    if (i%5==0):
        print("\n", end="")
    print ("\\t[" ,i,"]: %8.6f" %cost_val[i], end=" ")

# Show the cost function
plt.plot(W_val, cost_val)
plt.show()
```

```
print ("Wncost_val 출력 ")
print(type (cost_val) , end="")
for i in range(len(cost_val)):
    if (i%5==0):
        print("Wn", end="")
    print ("Wt[" ,i,"]: %8.6f" %cost_val[i], end="")
```

```
import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]

W = tf.placeholder(tf.float32)
# Our hypothesis for linear model X * W
hypothesis = X * W

# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))
# Launch the graph in a session.
sess = tf.Session()
# Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())
# Variables for plotting cost function
W_val = []
cost_val = []

for i in range(-30, 50):
    feed_W = i * 0.1
    print("i:%3d" %i, " -> ", "curr_cost:%10.6f" %curr_cost, "\t curr_W:%5.1f" %curr_W)

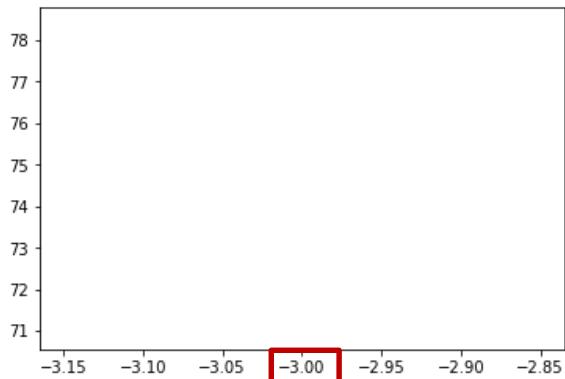
    curr_cost, curr_W = sess.run([cost, W], feed_dict={W: feed_W})

    W_val.append(curr_W)
    cost_val.append(curr_cost)

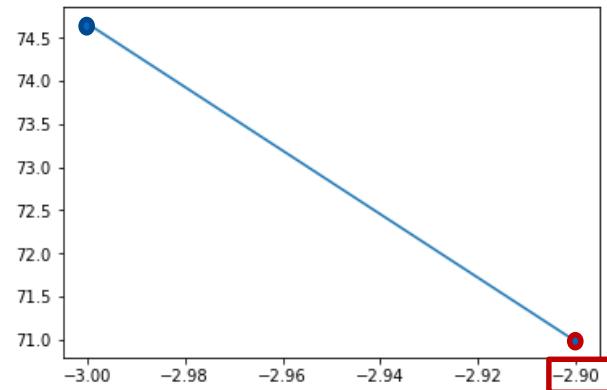
print("i:%3d" %i, " -> ", "curr_cost:%10.6f" %curr_cost, "\t curr_W:%5.1f" %curr_W)

plt.plot(W_val, cost_val)
plt.show()
```

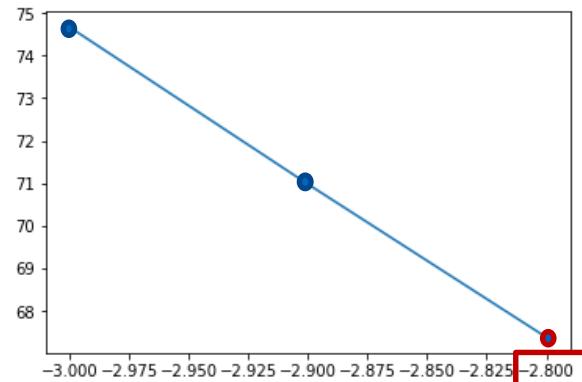
i:-30 -> curr_cost: 74.666664 curr_W: -3.0



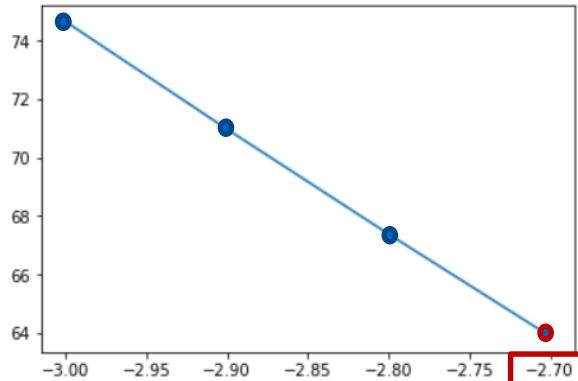
i:-29 -> curr_cost: 70.980003 curr_W: -2.9



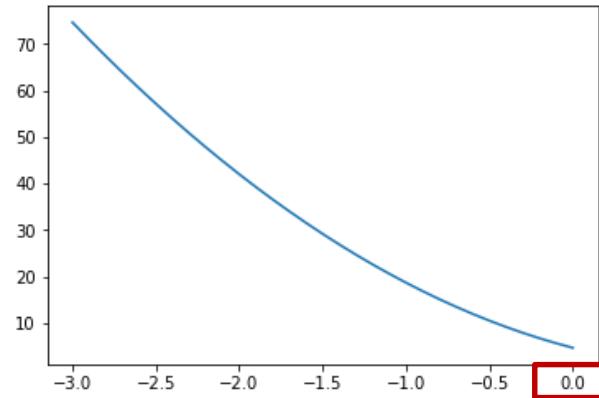
i:-28 -> curr_cost: 67.386665 curr_W: -2.8



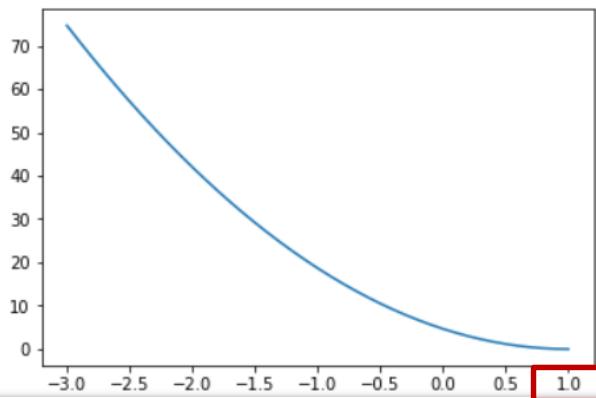
i:-27 -> curr_cost: 63.886669 curr_W: -2.7



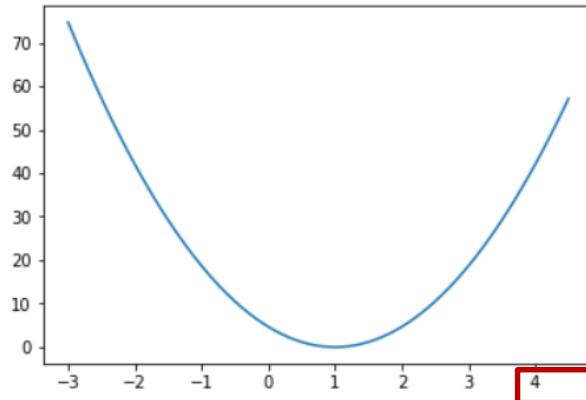
i: 0 -> curr_cost: 4.666667 curr_W: 0.0



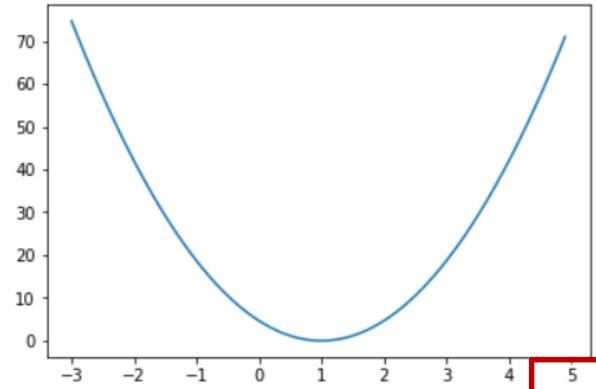
i: 10 -> curr_cost: 0.000000 curr_W: 1.0



i: 45 -> curr_cost: 57.166668 curr_W: 4.5



i: 49 -> curr_cost: 70.980003 curr_W: 4.9



```

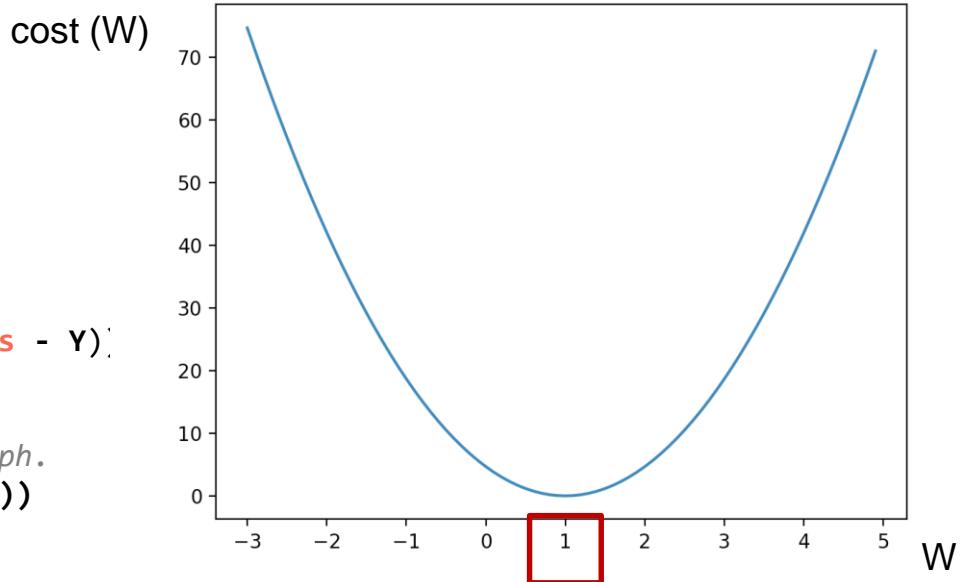
import tensorflow as tf
import matplotlib.pyplot as plt
X = [1, 2, 3]
Y = [1, 2, 3]

W = tf.placeholder(tf.float32)
# Our hypothesis for linear model X * w
hypothesis = X * W

# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))
# Launch the graph in a session.
sess = tf.Session()
# Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())
# Variables for plotting cost function
W_val = []
cost_val = []
for i in range(-30, 50):
    feed_W = i * 0.1
    curr_cost, curr_W = sess.run([cost, W], feed_dict={W: feed_W})
    W_val.append(curr_W)
    cost_val.append(curr_cost)

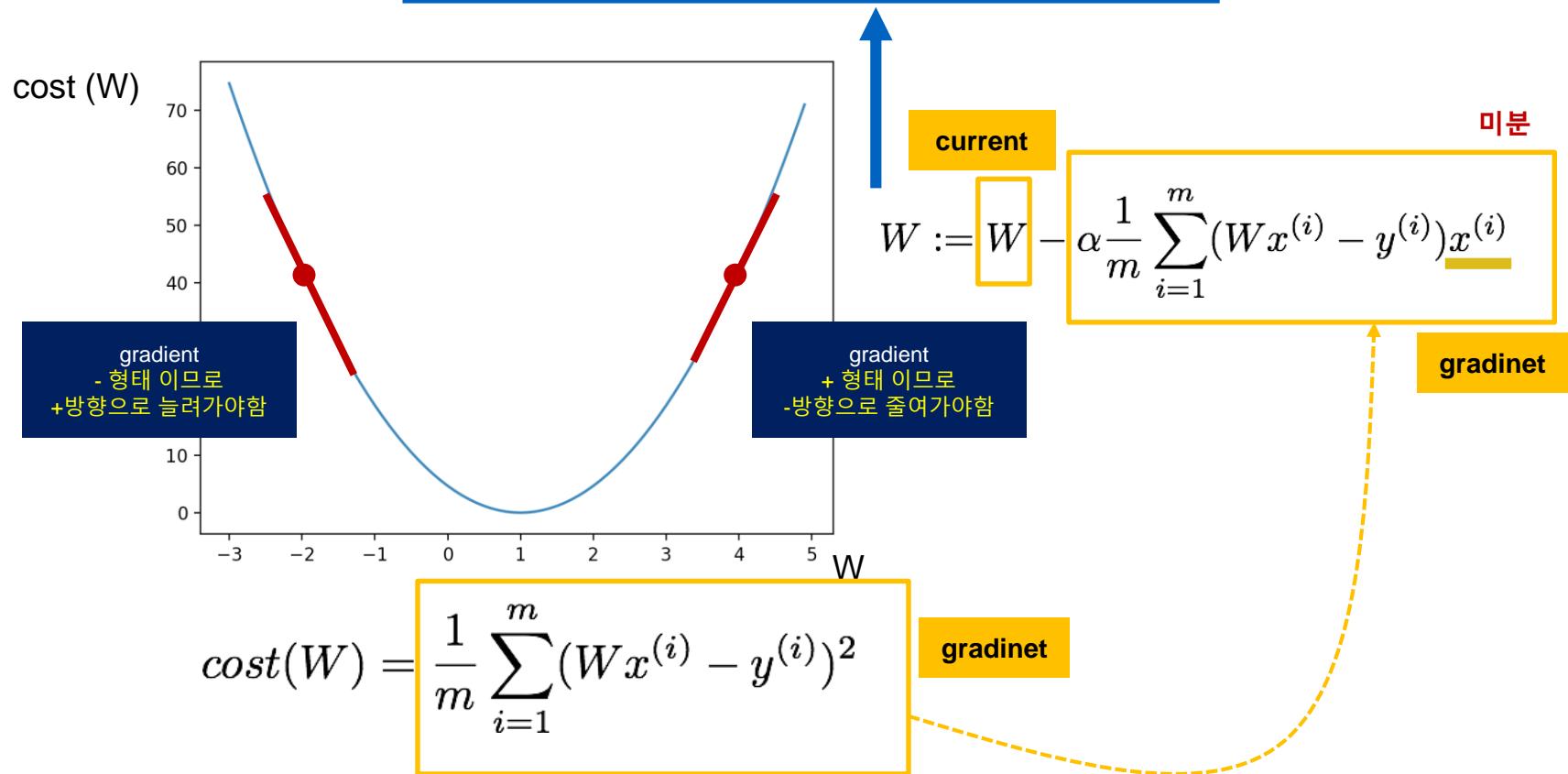
# Show the cost function
plt.plot(W_val, cost_val)
plt.show()

```



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

Gradient descent



```
import tensorflow as tf
x_data = [1, 2, 3]
y_data = [1, 2, 3]

# Create a tensor of shape [1] consisting of random normal values
W = tf.Variable(tf.random_normal([1]), name='weight')
X = tf.placeholder(tf.float32)
Y = tf.placeholder(tf.float32)
```

Our hypothesis for Linear model X * W

hypothesis = $x * w$

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

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

```
# Minimize: Gradient Descent using derivative: w -= learning_rate * derivative
```

```
learning_rate = 0.1
```

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

```
descent = W - learning rate * gradient
```

```
update = W.assign(descent)
```

```
# Launch the graph in a session.
```

```
sess = tf.Session()
```

Initializes global variables in the graph.

```
sess.run(tf.global_variables_initializer())
```

```
gra_val = []
cost_val = []
```

```
print("step \t gradient \t \t cost \t\t w ")
```

```
for step in range(41):
```

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

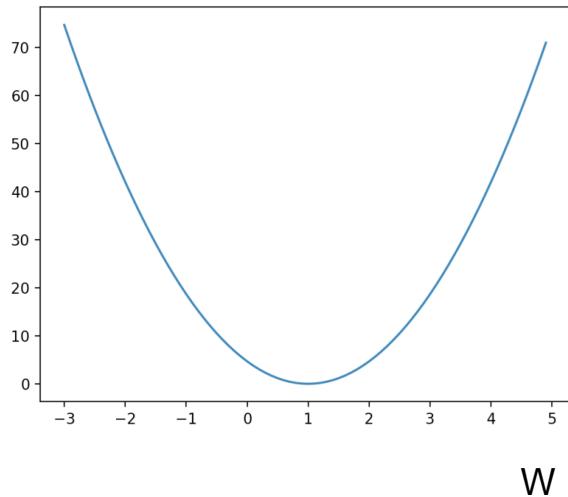
Print

Next slide

step	gradient	cost	w
0	-1.8486409	0.7323157	[0.78872675]
1	-0.9859419	0.20830317	[0.88732094]
2	-0.5258357	0.059250683	[0.9399045]
3	-0.28044555	0.01685351	[0.9679491]
4	-0.14957096	0.004793887	[0.98290616]
5	-0.07977118	0.0013635947	[0.9908833]
6	-0.042544704	0.00038786823	[0.99513775]
7	-0.022690436	0.000110326255	[0.9974068]
8	-0.012101631	3.1382027e-05	[0.99861693]
9	-0.006454309	8.926737e-06	[0.9992624]
10	-0.0034421086	2.538881e-06	[0.9996066]
11	-0.0018358231	7.2219564e-07	[0.9997902]
12	-0.0009791056	2.0542454e-07	[0.9998881]
13	-0.00052215654	5.8424458e-08	[0.99994034]
14	-0.00027849278	1.661962e-08	[0.9999682]
15	-0.00014841557	4.720112e-09	[0.999983]
16	-7.933378e-05	1.3486824e-09	[0.99999094]
17	-4.227956e-05	3.8304884e-10	[0.9999952]
18	-2.259016e-05	1.09353714e-10	[0.99999744]
19	-1.1901061e-05	3.0350833e-11	[0.9999986]
20	-6.3379607e-06	8.608225e-12	[0.9999993]
21	-3.33786e-06	2.3874236e-12	[0.99999964]
22	-1.5497208e-06	5.163277e-13	[0.9999998]
23	-7.748604e-07	1.2908193e-13	[0.9999999]
24	-6.755193e-07	9.947598e-14	[0.99999994]
25	-3.3775964e-07	2.4868996e-14	[1.]
26	0.0	0.0	[1.]
27	0.0	0.0	[1.]
28	0.0	0.0	[1.]
29	0.0	0.0	[1.]
30	0.0	0.0	[1.]
31	0.0	0.0	[1.]
32	0.0	0.0	[1.]
33	0.0	0.0	[1.]
34	0.0	0.0	[1.]
35	0.0	0.0	[1.]
36	0.0	0.0	[1.]
37	0.0	0.0	[1.]
38	0.0	0.0	[1.]
39	0.0	0.0	[1.]
40	0.0	0.0	[1.]

Gradient descent

cost (W)



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

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

```

W = tf.Variable(tf.random_normal([1]), name='weight')
learning_rate = 0.1
gradient = tf.reduce_mean((W * X - Y) * X)
descent = W - learning_rate * gradient
update = W.assign(descent)

```

텐서플로에서는 `=`로 할당하지 않고 `assign`으로 할당할 수 있다.
즉, `w`의 값을 `assign` 메소드를 사용하여 새롭게 할당한다.
이러게 할당하는 부분을 `update`라는 연산자로 사용하겠다는 뜻

step	cost	W
0	0.65983266	[0.6239778]
1	0.18768568	[0.7994548]
2	0.053386167	[0.89304256]
3	0.015185393	[0.94295603]
4	0.0043194103	[0.96957654]
5	0.0012286264	[0.9837742]
6	0.00034947737	[0.99134624]
7	9.940494e-05	[0.9953847]
8	2.8275383e-05	[0.9975385]
9	8.042801e-06	[0.9986872]
10	2.2877316e-06	[0.9992998]
11	6.5078694e-07	[0.9996266]
12	1.8508689e-07	[0.99980086]
13	5.262274e-08	[0.9998938]
14	1.4987608e-08	[0.9999433]
15	4.258099e-09	[0.9999698]
16	1.204801e-09	[0.9999839]
17	3.43789e-10	[0.9999914]
18	9.884715e-11	[0.9999954]
19	2.8162361e-11	[0.99999756]
20	7.716494e-12	[0.9999987]

step	cost	W
21	2.3874236e-12	[0.9999993]
22	5.163277e-13	[0.99999964]
23	1.2908193e-13	[0.9999998]
24	9.947598e-14	[0.9999999]
25	2.4868996e-14	[0.99999994]
26	0.0	[1.]
27	0.0	[1.]
28	0.0	[1.]
29	0.0	[1.]
30	0.0	[1.]
31	0.0	[1.]
32	0.0	[1.]
33	0.0	[1.]
34	0.0	[1.]
35	0.0	[1.]
36	0.0	[1.]
37	0.0	[1.]
38	0.0	[1.]
39	0.0	[1.]
40	0.0	[1.]

```
import tensorflow as tf
import matplotlib.pyplot as plt
```

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

```
# Create a tensor of shape [1] consisting of random normal values
W = tf.Variable(tf.random_normal([1]), name='weight')
X = tf.placeholder(tf.float32)
Y = tf.placeholder(tf.float32)
```

```
# Our hypothesis for Linear model X * W
hypothesis = X * W
```

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

```
# Minimize: Gradient Descent using derivative: W -= learning_rate * derivative
learning_rate = 0.1
gradient = tf.reduce_mean((W * X - Y) * X)
descent = W - learning_rate * gradient
update = W.assign(descent)
```

```
# Launch the graph in a session.
```

```
sess = tf.Session()
```

```
# Initializes global variables in the graph.
```

```
sess.run(tf.global_variables_initializer())
```

ex03_3.ipynb

Print

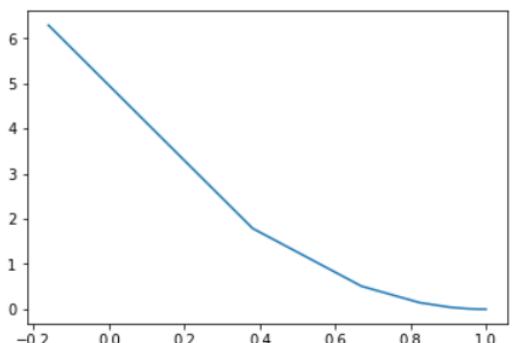
Next slide

step	cost	W
------	------	---

0	6.2836437	[-0.16038573]
1	1.7873474	[0.3811276]
2	0.508401	[0.66993475]
3	0.1446118	[0.8239652]
4	0.04113406	[0.90611476]
5	0.011700347	[0.94992787]
6	0.0033280961	[0.97329485]
7	0.0009466668	[0.98575723]
8	0.0002692744	[0.99240386]
9	7.6593424e-05	[0.99594873]
10	2.1786791e-05	[0.99783933]
11	6.196909e-06	[0.99884766]
12	1.7625867e-06	[0.9993854]
13	5.013033e-07	[0.99967223]
14	1.4264435e-07	[0.9998252]
15	4.0554635e-08	[0.9999068]
16	1.1543709e-08	[0.9999503]
17	3.286285e-09	[0.9999735]
18	9.329296e-10	[0.9999859]
19	2.6142763e-10	[0.9999925]
20	7.3949735e-11	[0.999996]
21	2.1486812e-11	[0.99999785]
22	5.8513194e-12	[0.99999887]
23	1.8047785e-12	[0.9999994]
24	4.5119464e-13	[0.9999997]
25	1.2908193e-13	[0.9999998]

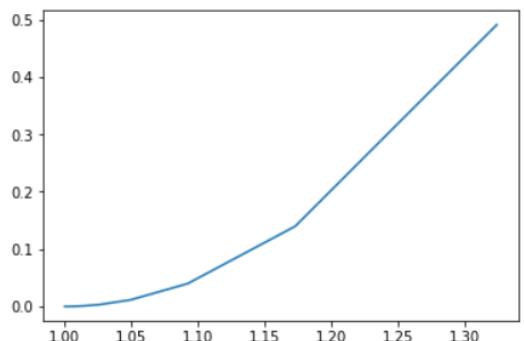
step	cost	W
------	------	---

26	9.947598e-14	[0.9999999]
27	2.4868996e-14	[0.99999994]
28	0.0	[1.]
29	0.0	[1.]
30	0.0	[1.]
31	0.0	[1.]
32	0.0	[1.]
33	0.0	[1.]
34	0.0	[1.]
35	0.0	[1.]
36	0.0	[1.]
37	0.0	[1.]
38	0.0	[1.]
39	0.0	[1.]
40	0.0	[1.]
41	0.0	[1.]
42	0.0	[1.]
43	0.0	[1.]
44	0.0	[1.]
45	0.0	[1.]
46	0.0	[1.]
47	0.0	[1.]
48	0.0	[1.]
49	0.0	[1.]
50	0.0	[1.]



step	cost	W
0	0.49110034	[1.3244007]
1	0.13969077	[1.1730137]
2	0.039734248	[1.092274]
3	0.011302193	[1.0492128]
4	0.0032148378	[1.0262468]
5	0.00091444066	[1.0139983]
6	0.0002601041	[1.0074657]
7	7.3986324e-05	[1.0039817]
8	2.1045045e-05	[1.0021236]
9	5.9866684e-06	[1.0011326]
10	1.7025194e-06	[1.000604]
11	4.841698e-07	[1.0003221]
12	1.3774762e-07	[1.0001718]
13	3.9115548e-08	[1.0000916]
14	1.1147942e-08	[1.0000489]
15	3.1744254e-09	[1.0000261]
16	9.111479e-10	[1.000014]
17	2.5492378e-10	[1.0000074]
18	7.316222e-11	[1.0000039]
19	2.1486812e-11	[1.0000021]
20	6.631732e-12	[1.0000012]
21	1.8047785e-12	[1.0000006]
22	5.163277e-13	[1.0000004]
23	2.6526928e-13	[1.0000002]
24	9.947598e-14	[1.0000001]
25	0.0	[1.]

step	cost	W
26	0.0	[1.]
27	0.0	[1.]
28	0.0	[1.]
29	0.0	[1.]
30	0.0	[1.]
31	0.0	[1.]
32	0.0	[1.]
33	0.0	[1.]
34	0.0	[1.]
35	0.0	[1.]
36	0.0	[1.]
37	0.0	[1.]
38	0.0	[1.]
39	0.0	[1.]
40	0.0	[1.]
41	0.0	[1.]
42	0.0	[1.]
43	0.0	[1.]
44	0.0	[1.]
45	0.0	[1.]
46	0.0	[1.]
47	0.0	[1.]
48	0.0	[1.]
49	0.0	[1.]
50	0.0	[1.]



```

import tensorflow as tf
x_data = [1, 2, 3]
y_data = [1, 2, 3]
# Create a tensor of shape [1] consisting of random normal values
W = tf.Variable(tf.random_normal([1]), name='weight')
X = tf.placeholder(tf.float32)
Y = tf.placeholder(tf.float32)

# Our hypothesis for linear model X * w
hypothesis = X * W

# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))

# Minimize: Gradient Descent using derivative: w -= learning_rate * derivative
learning_rate = 0.1
gradient = tf.reduce_mean((W * X - Y) * X)
descent = W - learning_rate * gradient
update = W.assign(descent)

# Launch the graph in a session.
sess = tf.Session()
# Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())
for step in range(41):
    sess.run(update, feed_dict={X: x_data, Y: y_data})
    print(step, "\t", sess.run(cost, feed_dict={X: x_data, Y: y_data}), "\t", sess.run(W))

```

Minimize: Gradient Descent Magic

```

optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.1)
t = optimizer.minimize(cost)

```

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

```

import tensorflow as tf
import matplotlib.pyplot as plt
# tf Graph Input
X = [1, 2, 3]
Y = [1, 2, 3]

# Set wrong model weights
W = tf.Variable(5.0)
# Linear model
hypothesis = X * W
# cost/Loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))

```

```

# Minimize: Gradient Descent Magic
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.1)
t = optimizer.minimize(cost)

```

```

# Launch the graph in a session.
sess = tf.Session()
# Initializes global variables in the graph.
sess.run(tf.global_variables_initializer())

W_val = []
cost_val = []

print("step \t cost \t \t W")
for step in range(21):
    curr_W = sess.run(W)
    curr_cost = sess.run(cost)
    print(step, "\t", curr_cost, "\t", curr_W)
    W_val.append(curr_W)
    cost_val.append(curr_cost)
    sess.run(t)

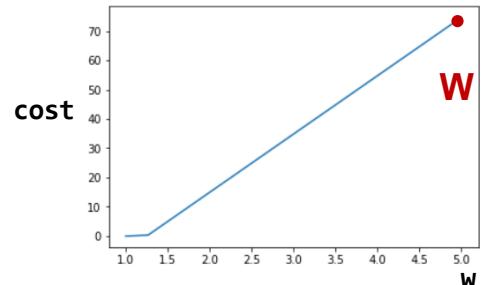
```

```

plt.plot(W_val, cost_val)
plt.show()

```

Output when W=5



step	cost	W
0	74.666664	5.0
1	0.3318512	1.2666664
2	0.0014748968	1.0177778
3	6.555027e-06	1.0011852
4	2.91322e-08	1.000079
5	1.2839034e-10	1.0000052
6	5.163277e-13	1.0000004
7	0.0	1.0
8	0.0	1.0
9	0.0	1.0
10	0.0	1.0
11	0.0	1.0
12	0.0	1.0
13	0.0	1.0
14	0.0	1.0
15	0.0	1.0
16	0.0	1.0
17	0.0	1.0
18	0.0	1.0
19	0.0	1.0
20	0.0	1.0

```
import tensorflow as tf
```

```
import matplotlib.pyplot as plt
```

```
# tf Graph Input
```

```
X = [1, 2, 3]
```

```
Y = [1, 2, 3]
```

```
# Set wrong model weights
```

```
W = tf.Variable(-3.0)
```

```
# Linear model
```

```
hypothesis = X * W
```

```
# cost/loss function
```

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

```
# Minimize: Gradient Descent Magic
```

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

```
t = optimizer.minimize(cost)
```

```
# Launch the graph in a session.
```

```
sess = tf.Session()
```

```
# Initializes global variables in the graph.
```

```
sess.run(tf.global_variables_initializer())
```

```
W_val = []
```

```
cost_val = []
```

```
print("step \t cost \t\t W")
```

```
for step in range(21):
```

```
    curr_W = sess.run(W)
```

```
    curr_cost = sess.run(cost)
```

```
    print(step, "\t", curr_cost, "\t", curr_W)
```

```
    W_val.append(curr_W)
```

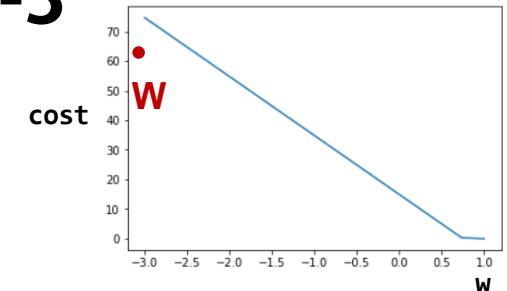
```
    cost_val.append(curr_cost)
```

```
    sess.run(t)
```

```
plt.plot(W_val, cost_val)
```

```
plt.show()
```

Output when $W = -3$



step	cost	W
0	74.666664	-3.0
1	0.3318512	0.7333336
2	0.0014748932	0.9822226
3	6.555027e-06	0.9988148
4	2.91322e-08	0.99992096
5	1.3195844e-10	0.9999947
6	5.163277e-13	0.99999964
7	2.4868996e-14	0.99999994
8	0.0	1.0
9	0.0	1.0
10	0.0	1.0
11	0.0	1.0
12	0.0	1.0
13	0.0	1.0
14	0.0	1.0
15	0.0	1.0
16	0.0	1.0
17	0.0	1.0
18	0.0	1.0
19	0.0	1.0
20	0.0	1.0

```

import tensorflow as tf
X = [1, 2, 3]
Y = [1, 2, 3]
# Set wrong model weights
W = tf.Variable(5.)
# Linear model
hypothesis = X * W

```

```

# Manual gradient
manual_cost = tf.reduce_mean((W * X - Y) * X) * 2
# cost/loss function
cost = tf.reduce_mean(tf.square(hypothesis - Y))

```

```

# Minimize: Gradient Descent Magic
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.1)
t = optimizer.minimize(cost)
# Get gradients
gvs = optimizer.compute_gradients(cost, W)
# Optional: modify gradient if necessary
# Apply gradients
apply_gradients = optimizer.apply_gradients(gvs)

```

```

sess = tf.Session()
sess.run(tf.global_variables_initializer())
print("step \t manul_cost \tW \t computer_gvs (cost, W)")
print("*"*80)
for step in range(100):
    print(step, "\t", sess.run(manual_cost), "\t %f" %sess.run(W), "\t" , sess.run(gvs)) # print(step, sess.run([gradient, W, gvs]))
    sess.run(apply_gradients) # 반드시 필요한 부분, 생략하면 W가 처음값으로 고정된다

```

Optional: compute_gradient and apply_gradient

Optional: compute_gradient and apply_gradient

step	manual_cost	W	computer_gvs (cost, W)
0	37.333332	5.000000	[37.333336, 5.0]
1	33.84889	4.626667	[33.84889, 4.626665]
2	30.689657	4.288177	[30.689657, 4.2881775]
3	27.825287	3.981281	[27.825287, 3.9812808]
4	25.228262	3.703028	[25.228264, 3.703028]
5	22.873621	3.450745	[22.873623, 3.4507453]
.			
.			
.			
95	0.0033854644	1.000363	[0.0033854644, 1.0003628]
96	0.0030694802	1.000329	[0.0030694804, 1.0003289]
97	0.0027837753	1.000298	[0.0027837753, 1.0002983]
98	0.0025234222	1.000270	[0.0025234222, 1.0002704]
99	0.0022875469	1.000245	[0.0022875469, 1.0002451]

Hypothesis visualization

```

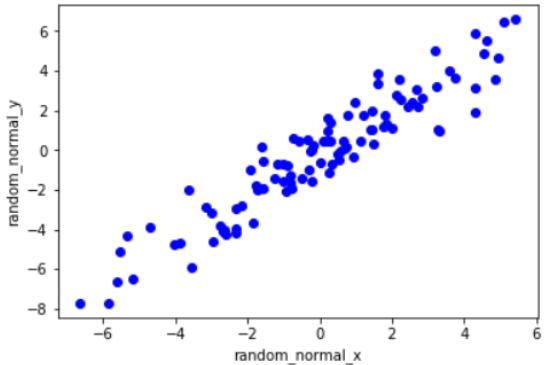
import numpy as np
import matplotlib.pyplot as plt
import tensorflow as tf
import math

# 100개의 데이터 생성
num_datas = 100
data = []

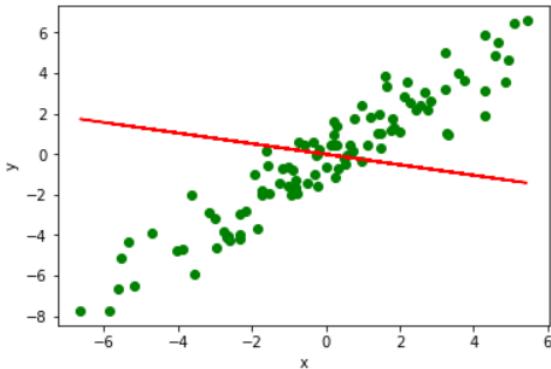
# random.normal 이용하여 x, y 데이터 100개 생성
for i in range(num_datas):
    x = np.random.normal()
    y = x*1 + np.random.normal()
    data.append([x,y])

print(i, "*30")
print("data : ", data)
x_data = [v[0] for v in data] # x_data //는 data의 1열 요소가 들어감, 즉 x가 들어감
print("x_data : ", x_data)
y_data = [v[1] for v in data] # y_data //는 data의 2열 요소가 들어감, 즉 y가 들어감
print("y_data : ", y_data)

```



1 번째 hypothesis



Hypothesis visualization

```
# graph
plt.plot(x_data,y_data,'bo')
plt.xlabel('random_normal_x')
plt.ylabel('random_normal_y')
plt.show()
```

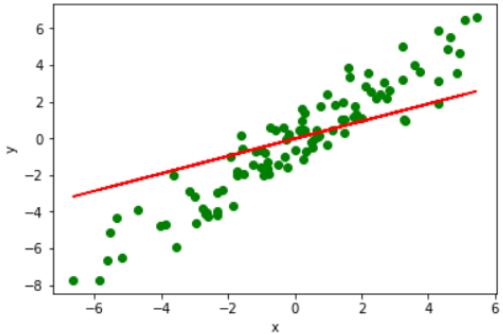
```
# linear regression
W = tf.Variable(tf.random_uniform([1], -1.0, 1.0))
b = tf.Variable(tf.zeros([1])) # 초기값은 0
hypothesis = W * x_data + b
```

```
# cost function
# learning rate : 너무 작으면 느리게 학습하고 너무 크면 발산해 버림
cost = tf.reduce_mean(tf.square(hypothesis - y_data))
optimizer = tf.train.GradientDescentOptimizer(0.01)
train = optimizer.minimize(cost)
```

```
# Tensorflow session
sess = tf.Session()
init = tf.initialize_all_variables()
sess.run(init)
```

```
for step in range(10): # learning 10步
    sess.run(train)
    print(step+1,"번째 hypothesis")
    plt.plot(x_data, y_data, 'go')
    plt.plot(x_data, sess.run(hypothesis), 'r') # hypothesis 출력
    plt.xlabel('x')
    plt.ylabel('y')
    plt.show()
```

6 번째 hypothesis



10 번째 hypothesis

