## 0. how tf works

#### In [ ]:

```
import tensorflow as tf
import numpy as np

data = np.array([[1,2,3],[4,5,6]])

# computational graph
x1 = tf.placeholder(tf.float32, [1,3])
x2 = tf.placeholder(tf.float32, [3,1])
product = tf.matmul(x1, x2)

with tf.Session() as sess:
    print(sess.run(product, feed_dict={x1:data[0].reshape(1,3), x2:data[1].reshape(3,1)}))
# note: reshape is necessary because (3,) can not work
```

# 1, how to use tf to do optimization

#### In [2]:

```
data = np.array([2, 2]).reshape(1,-1)

# computational graph
x = tf.placeholder(tf.float32, [1, None]) # None means dont know
w = tf.Variable(tf.random_normal([1, 2]))
cost = tf.matmul(x-w, tf.transpose((x-w), [1,0]))

# more complex computational graph -> a training step
optimizer = tf.train.AdamOptimizer(0.1)
train_step = optimizer.minimize(cost)

with tf.Session() as sess:
    # since we use Variable we need to initialize it
    sess.run(tf.initializers.global_variables())

for _ in range(100):
    sess.run(train_step, feed_dict={x: data})
    print(sess.run(cost, feed_dict={x: data}))
print('\nw suppose to be [2,2], we got: ', sess.run(w))
```

## bonus

```
In [3]:
```

```
# constant

c1 = tf.constant(1)
 c2 = tf.constant(2)
 c3 = tf.constant(3)

add = tf.add(c1, c2)
 ma = tf.multiply(add, c3)

# m1 = tf.constant([[3, 3]])
 # m2 = tf.constant([[2], [3]])
```

### In [6]:

```
p1 = tf.placeholder(tf.float32)  # tensor
p2 = tf.placeholder(tf.float32)
state = tf.multiply(p1, p2)  # elementwise

a=1
b=2
# pic = tf.placeholder(tf.float32, [None, n_dim])
# wI = tf.Variable(tf.random_normal([n_dim, nb_n_1]))
update = tf.assign(state, tf.add(state, 1))  # only variable can be assigned. Constant can not be as
```

### In [7]:

```
# fetch and feed
with tf.Session() as sess:
    # fetch
    print(sess.run([add, ma]))

# feed
    print(sess.run(state, feed_dict={p1: [a], p2: [b]}))
```

[3, 9] [2.]

# build a NN

## theoritical one

In [ ]:

```
# assume we already have data, and we ignore the feed for the train
# so the following programming can not work without some concret cases
x train =
y_train =
# assume the x_train is (1000, 5) data, y_train is (1000, 1)
x = tf.placeholder(tf.float32, [None, 5])
y = tf.placeholder(tf.float32, [None, 1])
n1 = 4
w1 = tf.Variable(tf.random_normal([5, n1])) # it is transpose of the form in textbook of NN
b1 = tf. Variable(tf.zeros([1, n1])) # b1 is (1,4) because the output is 1 dimensional
z1 = tf.matmul(x, w1) + b1 # here broadcast will happen
a1 = tf. nn. relu(z1)
n2 = 1
w2 = tf. Variable(tf. random normal([n1, 1]))
b2 = tf. Variable(tf. zeros([1, n2]))
z2 = tf. matmul(a1, w1) + b2
a2 = z2
loss = tf. reduce mean(tf. square(a2-y))
optimizer = tf. train. GradientDescentOptimizer (0.2)
train = optimizer.minimize(loss)
init = tf.initializers.global variables()
with tf. Session() as sess:
    sess.run(init)
    for i in range (200):
        sess.run(train, feed_dict={x: x_train, y: y_train})
    print(sess.run(loss, feed_dict={x: x_train, y: y_train}))
```

### real one

#### In [13]:

```
from tensorflow.examples.tutorials.mnist import input data
mnist = input_data.read_data_sets("MNIST_data/", one_hot=True)
# new we going to build the NN; lets start with placeholder for data and parameter
                = tf. placeholder (tf. float32, [None, 784])
pic
the label
                = tf.placeholder(tf.float32, [None, 10])
learning rate = tf. placeholder (tf. float32)
# laver 1
nb \ n \ 1 = 500
w1 = tf. Variable(tf. truncated_normal([784, nb_n_1], stddev = 0.1))
b1 = tf. Variable(tf. zeros([nb n 1]) + 0.1)
11 = tf. nn. tanh(tf. matmul(pic, w1) + b1)
nb n = nb n 1
# layer 2
nb \ n \ 2 = 300
w2 = tf. Variable(tf. truncated normal([nb n, nb n 2], stddev = 0.1))
b2 = tf. Variable(tf. zeros([nb_n_2]) + 0.1)
12 = tf. nn. tanh(tf. matmul(11, w2) + b2)
nb_n = nb_n_2
# layer 3
nb \ n \ 3 = 10
w3 = tf. Variable(tf. truncated normal([nb n, 10], stddev = 0.1))
b3 = tf. Variable(tf. zeros([nb_n_3]) + 0.1)
13 = tf.nn.softmax(tf.matmul(12, w3) + b3) # softmax layer takes no activation function
loss = tf. losses. softmax cross entropy (the label, 13)
# loss = tf. reduce mean(tf. square(the label-13))
optimizer = tf. train. AdamOptimizer (learning_rate)
train = optimizer. minimize (loss)
# www also want to see some performance during the training
correct predict = tf. equal(tf. argmax(13, 1), tf. argmax(the label, 1))
                = tf.reduce mean(tf.cast(correct predict, tf.float32))
accuracy
# tf.cast make True to be 1
n \text{ epoch} = 5
batch size = 5000
n batch = mnist.train.num examples // batch size
# we need n batch to write the loop
# in the loop we will use:
# batch xs, batch ys = mnist. train. next batch (batch size)
with tf. Session() as sess:
    sess.run(tf.initializers.global variables())
    for epoch in range (n epoch + 1):
        for batch in range (n batch):
            batch_xs, batch_ys = mnist.train.next_batch(batch_size)
            sess.run(train, feed dict={
                pic:
                                batch xs,
                 the label:
                                batch ys,
                 learning rate: 0.001 * (0.98**epoch)}
            if epoch \% 2 == 0:
                acc = sess.run(accuracy, feed dict={pic: mnist.test.images, the label: mnist.test.la
```

```
print('{}%\tIteration {} : accuracy : {}'.format(float(epoch)*100/n_epoch, epoch, accuracy : float(epoch)*100/n_epoch, epoch, epoch, accuracy : float(epoch)*100/n_epoch, epoch, epoc
```

In [ ]: