TensorFlow tutorials

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EC546 BSPLAB

Deep Learning Framework

- C++
 - Caffe
- Python
 - TensorFlow, Theano, Keras, MXNet, pyTorch
- Lua
 - Torch7
- Matlab
 - MatConvNet

Tensor Flow

Tensor: n-dimensional arrays

– Vector: 1-D tensor

Matrix: 2-D tensor

Deep learning processes are flows of tensors

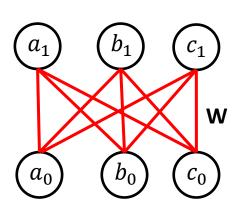
TensorFlow

- Graph: A TensorFlow computation, represented as a dataflow graph
- Op.: data operation
- Session: A class for running TensorFlow operations
- Tensor: data, n-dim arrays
- Variable: A variable maintains state in the graph across calls to run().
- Feed: lets you inject data into any Tensor in a computation graph.
- Fetch: lets you get data out of any Tensor in a computation graph.

TensorFlow

- 1. Building Graph: create operations
- Define Variable
- 3. Initialize Variable
- 4. Open session, and run Graph

A simple ReLU network



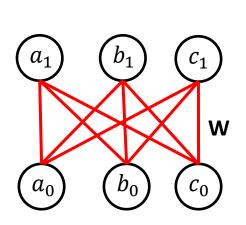
$$a_1 = \text{ReLU}(a_0 w_{a,a} + b_0 w_{b,a} + c_0 w_{c,a})$$

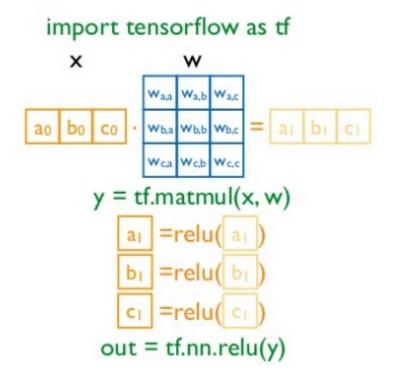
$$b_1 = \text{ReLU}(a_0 w_{a,b} + b_0 w_{b,b} + c_0 w_{c,b})$$

$$c_1 = \text{ReLU}(a_0 w_{a,c} + b_0 w_{b,c} + c_0 w_{c,c})$$
where ReLU(x) = max(x,0)

Matrix operation

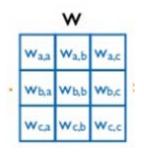
Create operation





Define Variable

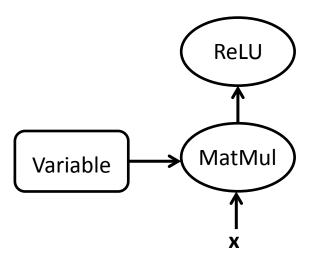
Variable(<initial-value>,name=<optional-name>)



```
import tensorflow as tf
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
```

Define Graph

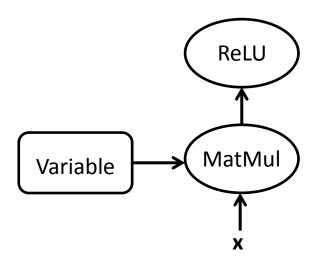
Code defines a data flow graph



```
import tensorflow as tf
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
```

Session

Session: manage resource for graph execution

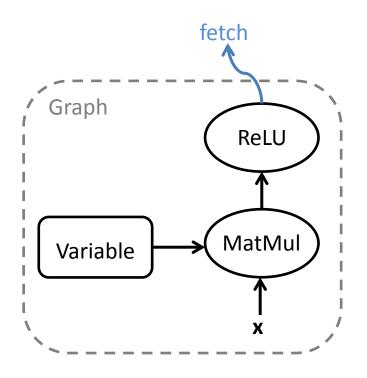


```
import tensorflow as tf
sess = tf.Session()
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
result = sess.run(out)
```

Sess = tf.InteractiveSession()
For interactive mode

Fetch

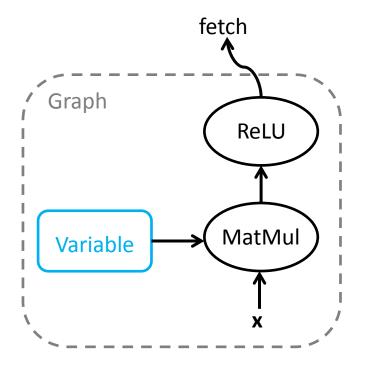
Retrieve content from a node



```
import tensorflow as tf
sess = tf.Session()
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
print sess.run(out)
```

Initialize Variable

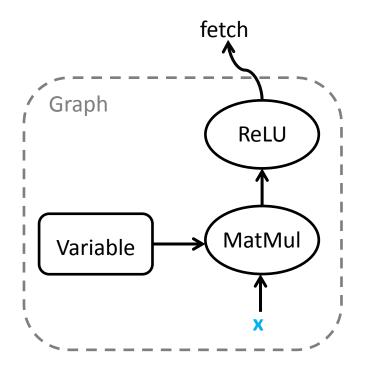
- Variable is an empty node
 - Fill in the content of a variable node



```
import tensorflow as tf
sess = tf.Session()
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
sess.run(tf.global_variables_initializer())
print sess.run(out)
```

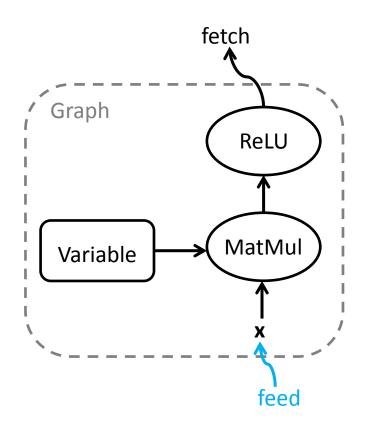
Placeholder

 placeholder(<data type>,shape=<optionalshape>,name=<optional-name>)



```
import tensorflow as tf
sess = tf.Session()
x = tf.placeholder("float",[1,3])
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
sess.run(tf.global_variables_initializer())
print sess.run(out)
```

Feed



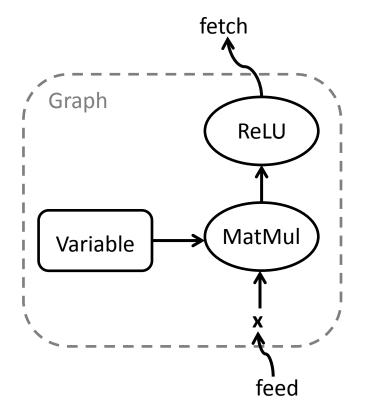
```
import numpy as np
import tensorflow as tf
sess = tf.Session()
x = tf.placeholder("float",[1,3])
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
sess.run(tf.global_variables_initializer())
print sess.run(out, feed_dict={x:np.array([[1,2,3]])})
```

Session management

Need to release resources after use

```
sess.close()
```

Common usage



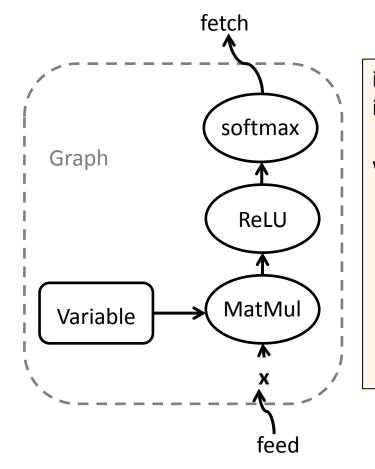
```
import numpy as np
import tensorflow as tf

with tf.Session() as sess:
  x = tf.placeholder("float",[1,3])
  w = tf.Variable(tf.random_normal([3,3]), name='w')
  y = tf.matmul(x,w)
  out = tf.nn.relu(y)
  sess.run(tf.global_variables_initializer())
  print sess.run(out, feed_dict={x:np.array([[1,2,3]])})
```

Prediction

Softmax

$$\sigma(\mathbf{z})_j = rac{e^{z_j}}{\sum_{k=1}^K e^{z_k}}$$
 for j = 1, ..., K .



```
import numpy as np
import tensorflow as tf

with tf.Session() as sess:
    x = tf.placeholder("float",[1,3])
    w = tf.Variable(tf.random_normal([3,3]), name='w')
    y = tf.matmul(x,w)
    out = tf.nn.relu(y)
    softmax = tf.nn.softmax(out)
    sess.run(tf.global_variables_initializer())
    print sess.run(softmax, feed_dict={x:np.array([[1,2,3]])})
```

Loss function

- Define loss function

- Cross entropy
$$\mathcal{L}(X,Y) = -\frac{1}{n} \sum_{i=1}^{n} y^{(i)} \ln a(x^{(i)}) + (1-y^{(i)}) \ln (1-a(x^{(i)}))$$

```
import numpy as np
import tensorflow as tf
with tf.Session() as sess:
x = tf.placeholder("float",[1,3])
w = tf.Variable(tf.random normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
labels = tf.placeholder("float",[1,3])
answer = np.array([[0,1,0]])
cross entropy = tf.nn.softmax cross_entropy_with_logits(logits=out, labels=labels)
sess.run(tf.global variables initializer())
 sess.run(softmax, feed_dict={x:np.array([[1,2,3]]), labels:answer})
```

Optimization

- Gradient descent
 - GradientDescentOptimizer(learning rate)

```
import numpy as np
import tensorflow as tf
with tf.Session() as sess:
x = tf.placeholder("float",[1,3])
w = tf.Variable(tf.random normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
labels = tf.placeholder("float",[1,3])
answer = np.array([[0,1,0]])
cross_entropy = tf.nn.softmax_cross_entropy_with_logits(out, labels, name='xen')
sess.run(tf.global variables initializer())
 train op = tf.train.GradientDescentOptimizer(0.1).minimize(cross entropy)
 sess.run(train op , feed dict={x:np.array([[1,2,3]]), labels:answer})
```

Iterative update

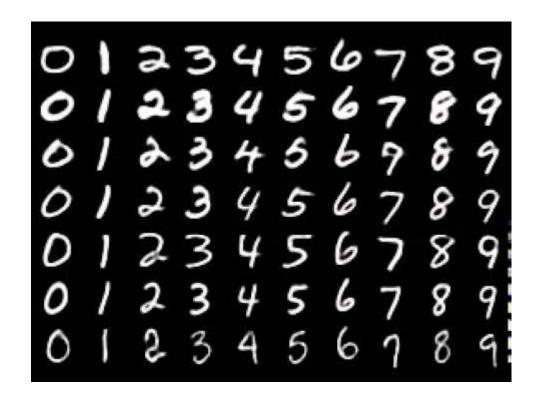
```
import numpy as np
import tensorflow as tf
x = tf.placeholder('float',[1,3])
w = tf.Variable(tf.random_normal([3,3]), name='w')
y = tf.matmul(x,w)
out = tf.nn.relu(y)
labels = tf.placeholder('float',[1,3])
answer = np.array([[0,1,0]])
cross entropy = tf.nn.softmax cross entropy with logits(logits=out, labels=labels)
train_op = tf.train.GradientDescentOptimizer(0.1).minimize(cross_entropy)
with tf.Session() as sess:
sess.run(tf.global variables initializer())
for step in range(10):
 sess.run(train_op , feed_dict={x:np.array([[1,2,3]]), labels:answer})
```

convolution

 Conv2d(input, filter, strides, padding, use_cudnn_on_gpu=None, name=None)

MNIST

• 60000 28×28 grey images for training, 10000 for testing

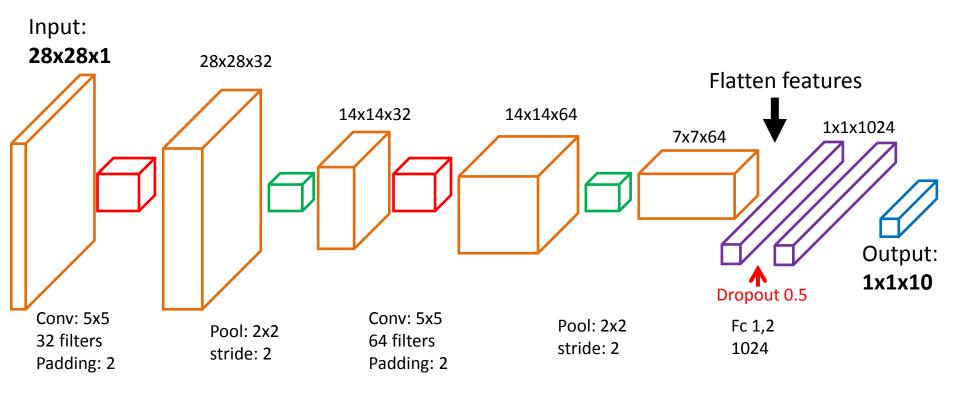


First CNN

https://www.tensorflow.org/versions/master/tu torials/mnist/pros/#deep-mnist-for-experts

- Build a Multilayer Convolutional Network
- 2 convolutional layers
- 2 fully connected layers

First CNN



First CNN

- How many parameters in this model?
- 1^{st} conv: 5x5x1x32 + 32 (bias) = 832
- 2^{nd} conv: 5x5x32x64 + 64 = 51,200
- 3^{rd} fc: 7x7x64x1024 + 1024 = 3,212,288
- 4^{th} fc: 1x1x1024x10 + 10 = 10,250

• Total: 3,274,570

Git the example code

- Execute command
- git clone https://github.com/JiaRenChang/DLcourse_NCTU.git

Run this example

```
cd DLcourse_NCTU
python CNN mnist.py
```

CNN code

Import functions and data

```
from __future__ import print_function
import tensorflow as tf

#use others' data loader
from tensorflow.examples.tutorials.mnist import input_data
mnist = input_data.read_data_sets('MNIST_data', one_hot=True)
```

CNN code

Define functions

```
12 # define functions
13 def weight variable(shape):
       initial = tf.truncated normal(shape, stddev=0.1)
14
       return tf.Variable(initial)
15
16
17 def bias variable(shape):
       initial = tf.constant(0.1, shape=shape)
18
       return tf.Variable(initial)
19
20
21 def conv2d(x, W):
       return tf.nn.conv2d(x, W, strides=[1, 1, 1, 1], padding='SAME')
22
23
24 def max pool 2x2(x):
       return tf.nn.max_pool(x, ksize=[1,2,2,1], strides=[1,2,2,1], padding='SAME'|)
25
```

CNN code

Define placeholder

```
27 #define placeholder
28 x = tf.placeholder(tf.float32, [None, 784])
29 y_ = tf.placeholder(tf.float32, [None, 10])
30 keep_prob = tf.placeholder(tf.float32)#Dropout rate
31 x_image = tf.reshape(x, [-1, 28, 28, 1]) # resize back to 28 x 28
```

ConvNet

```
34 ## conv1 layer ##
35 W conv1 = weight variable([5,5, 1,32]) # patch 5x5, in size 1, out size 32
36 b conv1 = bias variable([32])
37 h conv1 = tf.nn.relu(conv2d(x image, W conv1) + b conv1) # output size 28x28x32
38 h pool1 = max pool 2x2(h conv1)
                                                                            # output size 14x14x32
39
40 ## conv2 layer ##
41 W conv2 = weight variable([5,5, 32, 64]) # patch 5x5, in size 32, out size 64
42 b conv2 = bias variable([64])
43 h_conv2 = tf.nn.relu(conv2d(h_pool1, W_conv2) + b_conv2) # output size 14x14x64
44 h pool2 = max pool 2x2(h conv2) # output size 7x7x64
45
46 ## fc1 layer ##
47 W fc1 = weight variable([7*7*64, 1024])
48 b fc1 = bias variable([1024])
49 # [n samples, 7, 7, 64] ->> [n samples, 7*7*64]
50 h pool2 flat = tf.reshape(h pool2, [-1, 7*7*64])
51 h fc1 = tf.nn.relu(tf.matmul(h pool2 flat, W fc1) + b fc1)
52
53 ## dropout ##
54 h fc1 drop = tf.nn.dropout(h fc1, keep prob)
55
56 ## fc2 layer ##
57 W_fc2 = weight_variable([1024, 10])
58 b fc2 = bias variable([10])
59 y fc = tf.matmul(h fc1 drop, W fc2) + b fc2
```

Learning rate

```
62 ## the learning rate
63 # learning rate 0.1 for first 80 epoch (469 iteration for 1 epoch)
64 # decay learning rate to 0.01 at 81th epoch
65 # decay learning rate to 0.01 at 121th epoch
66 global_step = tf.Variable(0, trainable=False)
67 boundaries = [37520, 56280]
68 values = [0.1, 0.01, 0.001]
69 learning_rate = tf.train.piecewise_constant(global_step, boundaries, values)
```

Loss function

```
# the loss function

74 cross_entropy = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(labels=y_, logits=y_fc))

75

76 # optimizer SGD

77 train_step = tf.train.GradientDescentOptimizer(learning_rate).minimize(cross_entropy)

78

79 # prediction

80 correct_prediction = tf.equal(tf.argmax(y_fc,1), tf.argmax(y_,1))

81 accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
```

Start session

```
83 # start session
84 sess = tf.InteractiveSession()
85 # initize variable
86 sess.run(tf.global variables initializer())
87
88 for i in range(76916): # 164 epoch for all (496*164)
89 batch = mnist.train.next batch(128)
90 ## test every 100 step
91 if i%100 == 0:
       print("step %d, Test accuracy %g"%(i, (accuracy.eval(feed dict={
92
           x:mnist.test.images,y : mnist.test.labels, keep prob: 1.0})))
93
94 ## trianing
    train step.run(feed dict={x: batch[0], y : batch[1], keep prob: 0.5})
95
96
97 ##final testing
98 print("Final test accuracy %g"%accuracy.eval(feed dict={
       x: mnist.test.images, y : mnist.test.labels, keep prob: 1.0}))
100
```

It's should be ~99.28% test accuracy