

# TensorFlow Tutorial

---

SUNIL KUMAR SAHU

# Deep Learning Libraries

---

- ☐ Caffe : Python, Matlab, C++
- ☐ CNTK : Python, C++
- ☐ Theano : Python
- ☐ Torch : Lua
- ☐ Keras : Python and Theano
- ☐ Lasagne : Python and Theano
- ☐ TensorFlow : Python, C++
- ☐ etc.

# Deep Learning Libraries

---

☐ Caffe

☐ CNTK

Configuration File

☐ Theano

☐ Torch

☐ Keras

☐ Lasagne

☐ TensorFlow

☐ etc.

Programmatic generation

# TensorFlow

---

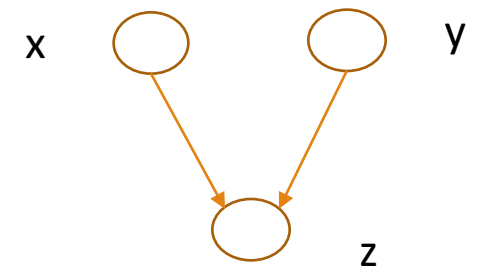
- ❑ Developed by *Google* as part of *Google Brain* project
- ❑ Large number of inbuilt functions available and keep getting update
- ❑ Pretty good documentations available and getting update
- ❑ Large number of users : You can get lots of code in github
- ❑ Better support for distributed system

# Summing two matrices

```
[>>> import numpy as np
[>>> x = np.zeros((3,2))
[>>> y = np.ones((3,2))
[>>> z = x + y
[>>> z
array([[ 1.,  1.],
       [ 1.,  1.],
       [ 1.,  1.]])
... █
```

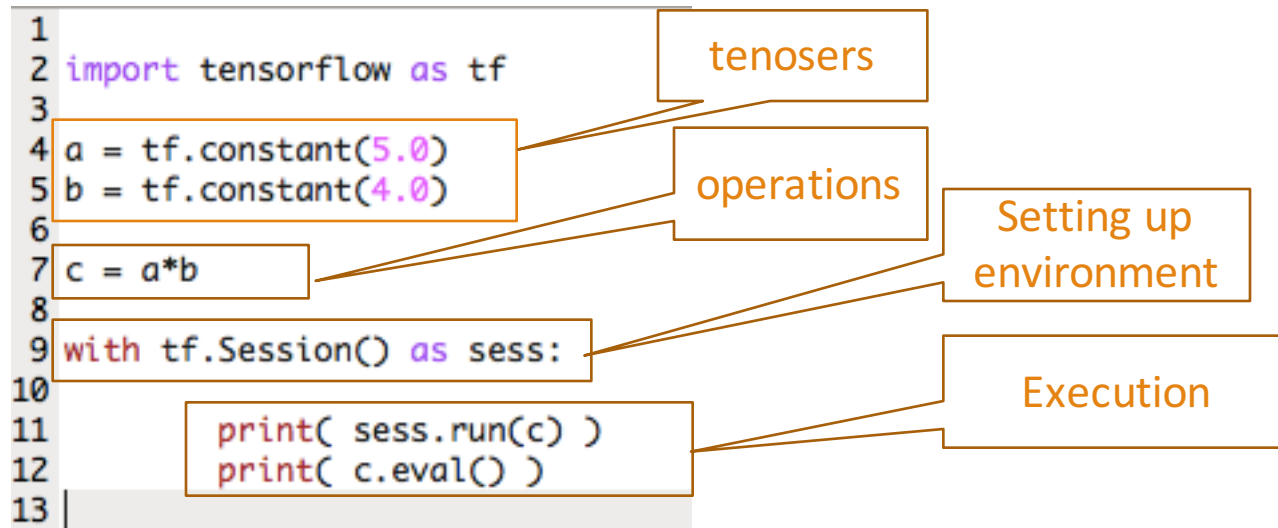
```
>>> import tensorflow as tf
>>> x = tf.zeros((3,2))
>>> y = tf.ones((3,2))
>>> z = x + y
>>> tf.InteractiveSession()
<tensorflow.python.client.session.InteractiveSession object at 0x11352aa90>
>>> z.eval()
array([[ 1.,  1.],
       [ 1.,  1.],
       [ 1.,  1.]], dtype=float32)
>>> █
```

Computation Graph



# Flow of Writing Codes

1. Create tensors
2. Write operations for tensors
3. Setting up environment for those operations to execute
4. Execute operations in given environment



# Tensors in TensorFlow

1. **Constant tensor :**
2. **Variable tensor :** initialize with `init_op`, restoring with save file, `assign()` works

```
1 import tensorflow as tf
2 import numpy as np
3
4 W1 = tf.convert_to_tensor(np.random.rand(4,3) )
5 W2 = tf.Variable(tf.ones((3,5), dtype='float64', name = 'weights') )
6
7 A = tf.matmul(W1, W2)
8
9 with tf.Session() as sess:
10     print (sess.run(W1) )
11
12     sess.run( tf.initialize_all_variables() )
13
14     print (sess.run(W2) )
15
16     print (sess.run(A))
17
18
```

Constant tensor

Variable tensor

Initialize all variable tensor

# Place Holder

---

- ❑ Kind of a Variable or dummy node in computation graph
- ❑ We can write sequence of operations with place holder
- ❑ Real assignment happens when we call run()
- ❑ Syntax :

```
X = tf.placeholder(tf.float32, [3, None], name='X' )  
X = tf.placeholder(tf.float32, [None, 3], name='X' )  
X = tf.placeholder(tf.float32, [None, None], name='X' )
```

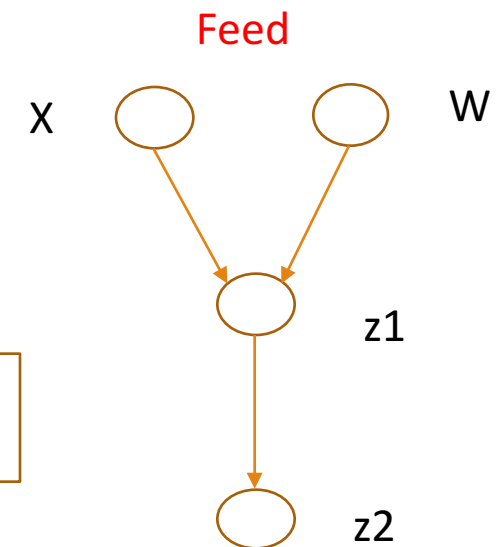


# Fetching and Feeding

```
1 import tensorflow as tf
2 import numpy as np
3
4 X = tf.placeholder(tf.float32, [None, 3], name='X' )
5
6 W = tf.Variable( tf.random_uniform( [3,2], -1.0, +1.0) , name = 'W' )
7
8 z1 = tf.matmul(X, W)
9 z2 = tf.nn.sigmoid(z1)
10
11 with tf.Session() as sess:
12     sess.run( tf.initialize_all_variables() )
13     a, b = sess.run([z1,z2], {X:np.ones((4,3))} )
14     print a
15     print b
```

Feed data into  
computation graph

Fetch values from  
computation graph



Computation Graph

# Save and Restore

```
1 import tensorflow as tf
2 import numpy as np
3
4 x = tf.placeholder(tf.float32, [2,2], name = 'x')
5
6 w1 = tf.Variable(tf.random_uniform([2,3]), name='w1')
7 w2 = tf.Variable(tf.random_uniform([3,1]), name='w1')
8
9 z1 = tf.matmul(x,w1)    # (2,3)
10 z2 = tf.matmul(z1,w2)  # (2,1)
11
12 with tf.Session() as sess :
13     sess.run(tf.initialize_all_variables())
14     saver = tf.train.Saver()
15     for i in range(10):
16         print sess.run( z2, {x:np.random.rand(2,2)} )
17         path = saver.save(sess, 'model/my-model.ckpt')
18         print path
```

```
1 import tensorflow as tf
2 import numpy as np
3
4 x = tf.placeholder(tf.float32, [2,2], name = 'x')
5
6 w1 = tf.Variable(tf.random_uniform([2,3]), name='w1')
7 w2 = tf.Variable(tf.random_uniform([3,1]), name='w1')
8
9 z1 = tf.matmul(x,w1)    # (2,3)
10 z2 = tf.matmul(z1,w2)  # (2,1)
11
12 with tf.Session() as sess :
13     saver = tf.train.Saver()
14     saver.restore(sess, "model/my-model.ckpt")
15     for i in range(10):
16         print sess.run( z2, {x:np.random.rand(2,2)} )
17
18
```

# Important Operations

```
8 |
9 |
10 |     # Concatenation
11 | emb0 = tf.constant( np.random.rand(3,4) )      # 3X4
12 | emb1 = tf.constant( np.random.rand(3,3) )      # 3X3
13 | X = tf.concat(1, [emb0, emb2])                  # 3X7
14 |
15 |     # Operations wrt one index
16 | X = tf.constant( np.ones((3,4)) )              # 3X4
17 | y = tf.reduce_sum( X, 1 )                      # Output: [4,4,4]
18 | z = tf.reduce_max(X, 1)                        # Output: [1,1,1]
19 | a = tf.reduce_mean(X, 1)                       # Output: [1,1,1]
20 | b = tf.argmax(X, 1)                            # Output: [0,0,0]
21 |
22 |
23 |     # Reshape the tensor
24 | x = tf.constant( np.random.rand(3,4,5) )       # 3X4X5
25 | y = tf.reshape( x, [3,20] )                   # 3X20
26 | z = tf.reshape( y, [-1] )                      # 60
27 | a = tf.reshape( z, [-1, 5, 2] )                # 6X5X2
28 | b = tf.expand_dims(a, -1)                      # 6X5X2X1
29 | c = tf.squeeze(b)                             # 6X5X2
30 |
31 |
32 |
```

# Part-2

---

## IMPLEMENTATIONS

# Multi Layer Neural Network (1)

nnEx.py

```
1 import tensorflow as tf
2
3 i_dim = 10
4 h1_dim = 5
5 num_classes = 3
6 #Symbolic or Place holder for input and output
7 X = tf.placeholder(tf.float32, [None, i_dim], name='input') # 100 X 10
8 input_y = tf.placeholder(tf.float32, [None, num_classes], name='input_y') # 100 X 3
9 #Initialize parameters
10 W1 = tf.get_variable('W_1', shape = [i_dim, h1_dim], initializer = tf.random_normal_initializer() ) # 10 X 5
11 b1 = tf.Variable(tf.constant(0.1, shape=[h1_dim]), name='b1') # 5
12 W2 = tf.get_variable('W_2', shape = [h1_dim, num_classes], initializer = tf.random_normal_initializer() ) # 5X3
13 b2 = tf.Variable(tf.constant(0.1, shape=[num_classes]), name='b2') # 3
14
15 #First layer operations
16 H1 = tf.nn.xw_plus_b(X, W1, b1, name="H2") # 100 X 5
17 Z1 = tf.sigmoid(H1) # 100 X 5
18 #Second layer operations
19 H2 = tf.nn.xw_plus_b(H1, W2, b2, name="H2") # 100 X 3
20 #Loss function
21 losses = tf.nn.softmax_cross_entropy_with_logits(H2, input_y) # 100
22 loss = tf.reduce_mean(losses) + 0.001 * (tf.nn.l2_loss(W1) + tf.nn.l2_loss(W2)) # 1
23 #Predictions of the batch
24 predictions = tf.argmax(H2, 1, name="predictions")
25 #Accuracy of correct prediction in batch
26 correct_predictions = tf.equal(predictions, tf.argmax(input_y, 1) )
27 accuracy = tf.reduce_mean(tf.cast(correct_predictions, "float"), name="accuracy")
28 #Optimization
29 optimizer = tf.train.AdamOptimizer(1e-2)
30 grads_and_vars = optimizer.compute_gradients(loss)
31 global_step = tf.Variable(0, name="global_step", trainable=False)
32 train_op = optimizer.apply_gradients(grads_and_vars, global_step = global_step)
33
34 with tf.Session() as sess :
35
36     sess.run(tf.initialize_all_variables())
37     # Create Dataset
38     D = np.asarray( np.random.rand(1000, 10), dtype='float32')
39     Y = np.asarray( np.zeros((1000, 3)), dtype='float32' )
40     for i in range(1000) :
41         k = np.random.randint(3)
42         Y[i][k] = 1.0
43     X_train = D[0:800]; Y_train = Y[0:800]
44     X_test = D[800:]; Y_test = Y[800:]
45     # Training
```

Create Tensors

Define  
Operations

Environment  
Setup

Execution

# Multi Layer Neural Network (2)

nnEx.py

```
1 import tensorflow as tf
2
3 i_dim = 10
4 h1_dim = 5
5 num_classes = 3
6 #Symbolic or Place holder for input and output
7 X = tf.placeholder(tf.float32, [None, i_dim], name='input') # 100 X 10
8 input_y = tf.placeholder(tf.float32, [None, num_classes], name="input_y") # 100 X 3
9 #Initialize parameters
10 W1 = tf.get_variable('W_1', shape = [i_dim, h1_dim], initializer = tf.random_normal_initializer() ) # 10 X 5
11 b1 = tf.Variable(tf.constant(0.1, shape=[h1_dim]), name="b1") # 5
12 W2 = tf.get_variable('W_2', shape = [h1_dim, num_classes], initializer = tf.random_normal_initializer() ) # 5X3
13 b2 = tf.Variable(tf.constant(0.1, shape=[num_classes]), name="b2") # 3
14
15 #First layer operations
16 H1 = tf.nn.xw_plus_b(X, W1, b1, name="H2") # 100 X 5
17 Z1 = tf.sigmoid(H1) # 100 X 5
18 #Second layer operations
19 H2 = tf.nn.xw_plus_b(H1, W2, b2, name="H2") # 100 X 3
20 #Loss function
21 losses = tf.nn.softmax_cross_entropy_with_logits(H2, input_y) # 100
22 loss = tf.reduce_mean(losses) + 0.001 * (tf.nn.l2_loss(W1) + tf.nn.l2_loss(W2)) # 1
23 #Predictions of the batch
24 predictions = tf.argmax(H2, 1, name="predictions")
25 #Accuracy of correct prediction in batch
26 correct_predictions = tf.equal(predictions, tf.argmax(input_y, 1))
27 accuracy = tf.reduce_mean(tf.cast(correct_predictions, "float"), name="accuracy")
28 #Optimization
29 optimizer = tf.train.AdamOptimizer(1e-2)
30 grads_and_vars = optimizer.compute_gradients(loss)
31 global_step = tf.Variable(0, name="global_step", trainable=False)
32 train_op = optimizer.apply_gradients(grads_and_vars, global_step = global_step)
33
34 with tf.Session() as sess :
35
36     sess.run(tf.initialize_all_variables())
37     # Create Dataset
38     D = np.asarray( np.random.rand(1000, 10), dtype='float32')
39     Y = np.asarray ( np.zeros((1000, 3)), dtype='float32' )
40     for i in range(1000) :
41         k = np.random.randint(3)
42         Y[i][k] = 1.0
43     X_train = D[0:800]; Y_train = Y[0:800]
44     X_test = D[800:]; Y_test = Y[800:]
45     # Training
```

```
5
6 i_dim = 10
7 h1_dim = 5
8 num_classes = 3
9
10 #Symbolic or Place holder for input and output
11 X = tf.placeholder(tf.float32, [None, i_dim], name='input') # 100 X 10
12 input_y = tf.placeholder(tf.float32, [None, num_classes], name="input_y") # 100 X 3
13
14 #Initialize parameters
15 W1 = tf.get_variable('W_1', shape = [i_dim, h1_dim], initializer = tf.random_normal_initializer() ) # 10 X 5
16 b1 = tf.Variable(tf.constant(0.1, shape=[h1_dim]), name="b1") # 5
17 W2 = tf.get_variable('W_2', shape = [h1_dim, num_classes], initializer = tf.random_normal_initializer() ) # 5X3
18 b2 = tf.Variable(tf.constant(0.1, shape=[num_classes]), name="b2") # 3
19
```

# Multi Layer Neural Network (3)

```
1 import tensorflow as tf
2
3 i_dim = 10
4 h1_dim = 5
5 num_classes = 3
6 #Symbolic or Place holder for input and output
7 X = tf.placeholder(tf.float32, [None, i_dim], name='input') # 100 X 10
8 input_y = tf.placeholder(tf.float32, [None, num_classes], name='input_y') # 100 X 3
9 #Initialize parameters
10 W1 = tf.get_variable('W1', shape = [i_dim, h1_dim], initializer = tf.random_normal_initializer() ) # 10 X 5
11 b1 = tf.Variable(tf.constant(0.1, shape=[h1_dim]), name='b1') # 5
12 W2 = tf.get_variable('W2', shape = [h1_dim, num_classes], initializer = tf.random_normal_initializer() ) # 5X3
13 b2 = tf.Variable(tf.constant(0.1, shape=[num_classes]), name='b2') # 3
14
15 #First layer operations
16 H1 = tf.nn.xw_plus_b(X, W1, b1, name='H1') # 100 X 5
17 Z1 = tf.sigmoid(H1) # 100 X 5
18 #Second layer operations
19 H2 = tf.nn.xw_plus_b(H1, W2, b2, name='H2') # 100 X 3
20 #Loss function
21 losses = tf.nn.softmax_cross_entropy_with_logits(H2, input_y) # 100
22 loss = tf.reduce_mean(losses) + 0.001 * (tf.nn.l2_loss(W1) + tf.nn.l2_loss(W2)) # 1
23 #Predictions of the batch
24 predictions = tf.argmax(H2, 1, name='predictions')
25 #Accuracy of correct prediction in batch
26 correct_predictions = tf.equal(predictions, tf.argmax(input_y, 1) )
27 accuracy = tf.reduce_mean(tf.cast(correct_predictions, 'float'), name='accuracy')
28 #Optimization
29 optimizer = tf.train.AdamOptimizer(1e-2)
30 grads_and_vars = optimizer.compute_gradients(loss)
31 global_step = tf.Variable(0, name='global_step', trainable=False)
32 train_op = optimizer.apply_gradients(grads_and_vars, global_step = global_step)
33
34 with tf.Session() as sess :
35
36     sess.run(tf.initialize_all_variables())
37     # Create Dataset
38     D = np.asarray( np.random.rand(1000, 10), dtype='float32')
39     Y = np.asarray ( np.zeros((1000, 3)), dtype='float32' )
40     for i in range(1000) :
41         k = np.random.randint(3)
42         Y[i][k] = 1.0
43     X_train = D[0:800]; Y_train = Y[0:800]
44     X_test = D[800:]; Y_test = Y[800:]
45     # Training
21
22 #First layer operations
23 H1 = tf.nn.xw_plus_b(X, W1, b1, name='H1') # 100 X 5
24 Z1 = tf.sigmoid(H1) # 100 X 5
25
26 #Second layer operations
27 H2 = tf.nn.xw_plus_b(Z1, W2, b2, name='H2') # 100 X 3
28
29 #Loss function
30 losses = tf.nn.softmax_cross_entropy_with_logits(H2, input_y) # 100
31 loss = tf.reduce_mean(losses) + 0.001 * (tf.nn.l2_loss(W1) + tf.nn.l2_loss(W2))
32
33 #Predictions of the batch
34 predictions = tf.argmax(H2, 1, name='predictions')
35
36 #Accuracy of correct prediction in batch
37 correct_predictions = tf.equal(predictions, tf.argmax(input_y, 1) )
38 accuracy = tf.reduce_mean(tf.cast(correct_predictions, 'float'), name='accuracy')
39
40 #Optimization
41 optimizer = tf.train.AdamOptimizer(1e-2)
42 grads_and_vars = optimizer.compute_gradients(loss)
43 global_step = tf.Variable(0, name='global_step', trainable=False)
44 train_op = optimizer.apply_gradients(grads_and_vars, global_step = global_step)
45
```



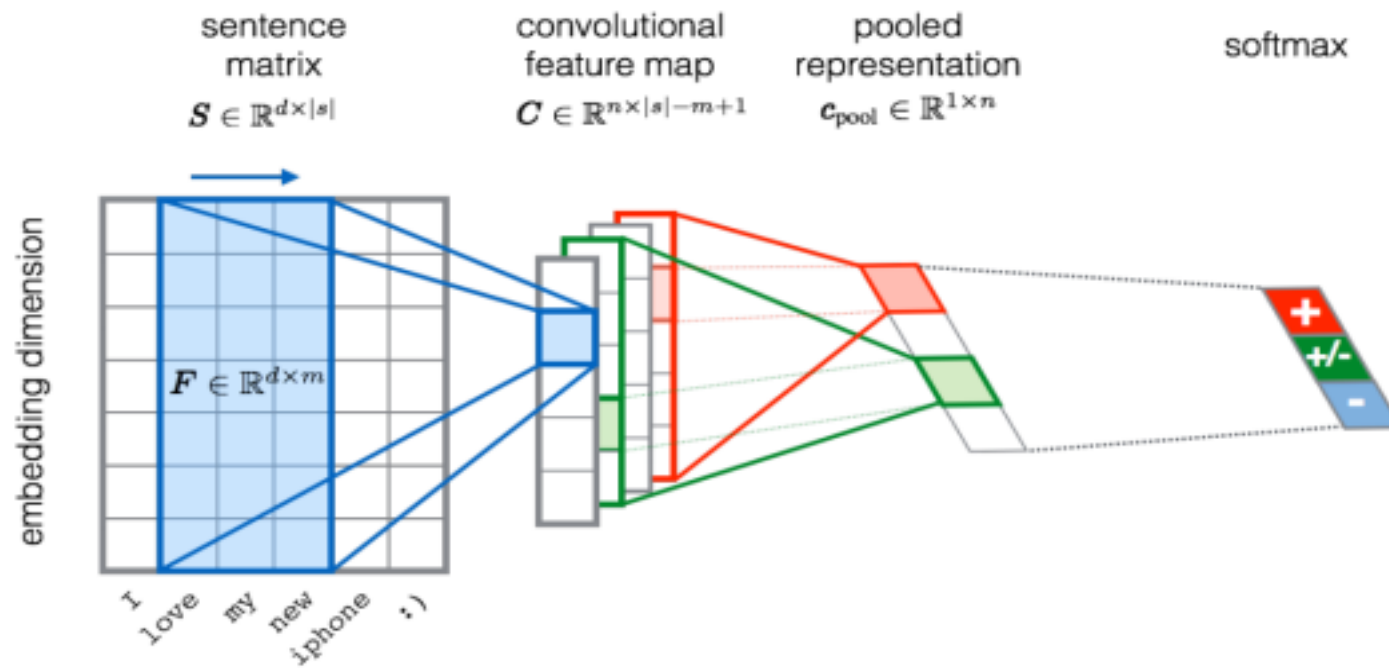
# Multi Layer Neural Network (4)

```
1 import tensorflow as tf
2
3 i_dim = 10
4 h1_dim = 5
5 num_classes = 3
6 #Symbolic or Place holder for input and output
7 X = tf.placeholder(tf.float32, [None, i_dim], name='input') # 100 X 10
8 input_y = tf.placeholder(tf.float32, [None, num_classes], name='input_y') # 100 X 3
9 #Initialize parameters
10 W1 = tf.get_variable('W1', shape = [i_dim, h1_dim], initializer = tf.random_normal_initializer() ) # 10 X 5
11 b1 = tf.Variable(tf.constant(0.1, shape=[h1_dim]), name='b1') # 5
12 W2 = tf.get_variable('W2', shape = [h1_dim, num_classes], initializer = tf.random_normal_initializer() ) # 5X3
13 b2 = tf.Variable(tf.constant(0.1, shape=[num_classes]), name='b2') # 3
14
15 #First layer operations
16 H1 = tf.nn.xw_plus_b(X, W1, b1, name='H2') # 100 X 5
17 Z1 = tf.sigmoid(H1) # 100 X 5
18 #Second layer operations
19 H2 = tf.nn.xw_plus_b(H1, W2, b2, name='H2') # 100 X 3
20 #Loss function
21 losses = tf.nn.softmax_cross_entropy_with_logits(H2, input_y) # 100
22 loss = tf.reduce_mean(losses) + 0.001 * (tf.nn.l2_loss(W1) + tf.nn.l2_loss(W2)) # 1
23 #Predictions of the batch
24 predictions = tf.argmax(H2, 1, name='predictions')
25 #Accuracy of correct prediction in batch
26 correct_predictions = tf.equal(predictions, tf.argmax(input_y, 1))
27 accuracy = tf.reduce_mean(tf.cast(correct_predictions, 'float'), name='accuracy')
28 #Optimization
29 optimizer = tf.train.AdamOptimizer(1e-2)
30 grads_and_vars = optimizer.compute_gradients(loss)
31 global_step = tf.Variable(0, name='global_step', trainable=False)
32 train_op = optimizer.apply_gradients(grads_and_vars, global_step = global_step)
33
34 with tf.Session() as sess :
35
36     sess.run(tf.initialize_all_variables())
37     # Create Dataset
38     D = np.asarray( np.random.rand(1000, 10), dtype='float32')
39     Y = np.asarray ( np.zeros((1000, 3)), dtype='float32' )
40     for i in range(1000):
41         k = np.random.randint(3)
42         Y[i][k] = 1.0
43     X_train = D[0:800]; Y_train = Y[0:800]
44     X_test = D[800:]; Y_test = Y[800:]
45     # Training
```

```
48
49 with tf.Session() as sess :
50
51     sess.run(tf.initialize_all_variables())
52
53     # Create Dataset
54     D = np.asarray( np.random.rand(1000, 10), dtype='float32')
55     Y = np.asarray ( np.zeros((1000, 3)), dtype='float32' )
56     for i in range(1000):
57         k = np.random.randint(3)
58         Y[i][k] = 1.0
59
60     X_train = D[0:800]; Y_train = Y[0:800]
61     X_test = D[800:]; Y_test = Y[800:]
62
63
64     # Training
65     for k in range(100):
66         _, l, acc = sess.run([train_op, loss, accuracy], {X:X_train, input_y:Y_train} )
67         print 'loss and accuracy', l, acc
68
69     # Testing
70     acc, pred = sess.run( [accuracy, predictions], {X:X_test, input_y:Y_test} )
71     print "Accuracy in test set", acc
72
```



# Convolution Neural Network (1)



Severyn 2015

# Convolution Neural Network (2)

---

*tf.nn.conv2d(input, filter, strides, padding, name)*

## Arguments :

- **input** : A 4-d tensor of shape [batch\_size, len\_sent, dim\_we, num\_channels]
- **filter** : A 4-d tensor of shape [filter\_size, dim\_we, num\_channels, num\_filters]
- **strides** : A list of *int* of size 4, [ 1, 1, 1, 1] / [1,2,1,1] / [1,3,1,1]
- **padding**: "VALID"/ "SAME"
- **name** : Name of the node in computation graph

## Return :

- A 4-d tensor of shape [batch\_size, out\_high, out\_width, num\_filters]  
where  $\text{out\_high} = (\text{len\_sent} - \text{filter\_size} + 1) / \left( \left\lceil \frac{\text{len\_sent}}{\text{strides}[1]} \right\rceil \right)$   
 $\text{out\_width} = 1$

# Convolution Neural Network (3)

cnnEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 word_dict_size = 100; sentMax=5; emb_size=4; filter_size=3; num_filters=8
5
6 w = tf.placeholder(tf.int32, [None, sentMax], name="x")          # word index matrix (N, 5)
7 W_wemb = tf.Variable(tf.random_uniform([word_dict_size, emb_size], -1.0, +1.0)) # word embedding matrix (100, 4)
8 emb0 = tf.nn.embedding_lookup(W_wemb, w)                       # (N, 5, 4)
9 X = tf.expand_dims(emb0, -1)                                   # (N, 5, 4, 1)
10
11 # Convolution layer
12 W_conv = tf.Variable(tf.truncated_normal([filter_size, emb_size, 1, num_filters], stddev=0.1), name="W_conv") # ( 3, 4, 1, 8 )
13 b_conv = tf.Variable(tf.constant(0.1, shape=[num_filters]), name="b_conv") # ( 8 )
14
15 conv = tf.nn.conv2d(X, W_conv, strides=[1, 1, 1, 1], padding="VALID", name="conv") # (N, 3, 1, 8)
16 conv1 = tf.nn.bias_add(conv, b_conv, name='conv1')             # add bias value
17 h1 = tf.nn.relu(conv1, name="relu")                            # apply activation function (N, 3, 1, 8)
18 pooled = tf.nn.max_pool(h1, ksize=[1, sentMax-filter_size+1, 1, 1], strides=[1, 1, 1, 1], padding='VALID', name="pool") # ( N, 1, 1, 8)
19 p1 = tf.squeeze(pooled) # (N,8)
20
21 # Running
22 with tf.Session() as sess :
23
24     sess.run( tf.initialize_all_variables() )
25     D = np.asarray( np.random.randint(0, high=word_dict_size, size = [10, sentMax] ), dtype='int32') # 1000 : size of dataset
26     p = sess.run(p1, {w:D} )
27     print np.shape(p) # (10, 8)
28
29
```

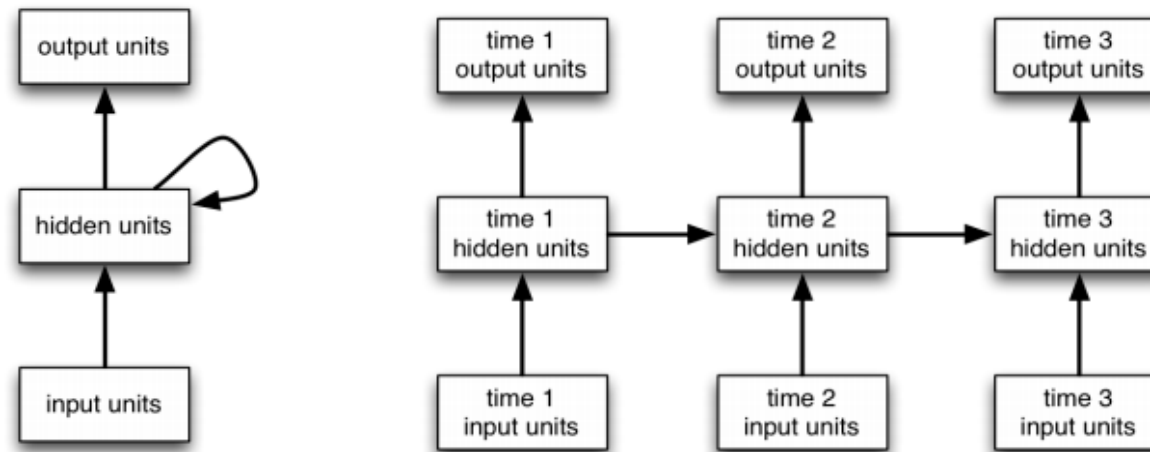
# Convolution Neural Network (4)

---

- ❖ `cnnEx1.py` : Multiple fixed length filter used
- ❖ `cnnEx2.py` : Multiple variable length filter used [3,4,5]

# Recurrent Neural Network (0)

- RNN is a special kind of NN which utilize sequential information and maintains history through its intermediate layer



- $$h^{(t)} = \tanh ( U \cdot x^{(t)} + W \cdot h^{(t-1)} )$$

# Recurrent Neural Network (1)

---

**`tf.nn.rnn_cell.BasicRNNCell(num_units)`**

*Argument:*

- *num\_units:* *int, size of hidden layer [H]*

*Return:*

- *cell :* *we will use this in `dynamtic_rnn()` or `bidirectional_dynamtic_rnn()`*

**`tf.nn.dynamic_rnn(inputs, cell, dtype, sequence_length )`**

*Arguments:*

- *inputs:* *input tensor [N,M,D]*
- *cell:* *Cell of particular RNN*
- *dtype:* *float32/float64*
- *sequence\_length:* *Sequence length of each sequence*

*Return:*

- *output :* *output for every word [N,M,H]*
- *last\_state :* *output of last word in every sequence [N,H]*

# Recurrent Neural Network (2)

rnnEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 X = tf.placeholder(tf.float64, [None, None, 8] )
5 X_lengths = tf.placeholder(tf.int32, [None] )
6
7 cell1 = tf.nn.rnn_cell.BasicRNNCell(num_units=4)
8 outputs, last_states = tf.nn.dynamic_rnn(
9     cell = cell1,
10    dtype = tf.float64,
11    sequence_length = X_lengths,
12    inputs = X
13 )
14
15 # Create input data
16 x = np.random.randn(6, 5, 8)
17 x_lengths = [4, 3, 4, 5, 5, 4]
18
19 with tf.Session() as sess:
20     sess.run( tf.initialize_all_variables() )
21     out, sta = sess.run([outputs, last_states], {X:x, X_lengths:x_lengths} )
22     print 'outputs', out
23     print 'final_states', sta
24
25 --
```

# Recurrent Neural Network (3)

birnnEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 X = tf.placeholder(tf.float64, [None, None, 8] )
5 X_lengths = tf.placeholder(tf.int32, [None] )
6
7 cell1 = tf.nn.rnn_cell.BasicRNNCell(num_units=4)
8 cell2 = tf.nn.rnn_cell.BasicRNNCell(num_units=4)
9 outputs, last_states = tf.nn.bidirectional_dynamic_rnn(
10     cell_fw = cell1,
11     cell_bw = cell2,
12     dtype = tf.float64,
13     sequence_length = X_lengths,
14     inputs = X
15 )
16
17 output_fw, output_bw = outputs
18 output_state_fw, output_state_bw = last_states
19 |
20 # Create input data
21 x = np.random.randn(6, 5, 8)
22 x_lengths = [4, 3, 4, 5, 5, 4]
23
24 with tf.Session() as sess:
25     sess.run( tf.initialize_all_variables() )
26     out_fw, out_bw, sta_fw, sta_bw = sess.run([output_fw, output_bw, output_state_fw, output_state_bw], {X:x, X_lengths:x_lengths} )
27     print 'forward output', out_fw
28     print 'forward final_outputs', sta_fw
29     print 'backward output', out_bw
30     print 'backward final_outputs', sta_bw
31
32
```



# Recurrent Neural Network (4)

lstmEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 X = tf.placeholder(tf.float64, [None, None, 8] )
5 X_lengths = tf.placeholder(tf.int32, [None] )
6
7 cell = tf.nn.rnn_cell.LSTMCell(num_units=4)
8 outputs, last_states = tf.nn.dynamic_rnn(
9     cell = cell,
10    dtype = tf.float64,
11    sequence_length = X_lengths,
12    inputs = X
13 )
14
15
16 # Create input data
17 x = np.random.randn(6, 5, 8)
18 x_lengths = [4, 3, 4, 5, 5, 4]
19
20 with tf.Session() as sess:
21     sess.run( tf.initialize_all_variables() )
22     out, sta = sess.run([outputs, last_states], {X:x, X_lengths:x_lengths} )
23     print 'output', out
24     print 'cell_states', sta.c
25     print 'final_outputs', sta.h
26
27 |
```

# Recurrent Neural Network (5)

gruEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 X = tf.placeholder(tf.float64, [None, None, 8] )
5 X_lengths = tf.placeholder(tf.int32, [None] )
6
7 cell1 = tf.nn.rnn_cell.GRUCell(num_units=4)
8 cell2 = tf.nn.rnn_cell.GRUCell(num_units=4)
9 outputs, last_states = tf.nn.bidirectional_dynamic_rnn(
10     cell_fw = cell1,
11     cell_bw = cell2,
12     dtype = tf.float64,
13     sequence_length = X_lengths,
14     inputs = X
15 )
16
17 output_fw, output_bw = outputs
18 output_state_fw, output_state_bw = last_states
19
20 # Create input data
21 x = np.random.randn(6, 5, 8)
22 x_lengths = [4, 3, 4, 5, 5, 4]
23
24 with tf.Session() as sess:
25     sess.run( tf.initialize_all_variables() )
26     out_fw, out_bw, sta_fw, sta_bw = sess.run([output_fw, output_bw, output_state_fw, output_state_bw], {X:x, X_lengths:x_lengths} )
27     print 'forward output', out_fw
28     print 'final_outputs', sta_fw
29     print 'backward output', out_bw
30     print 'backward final_outputs', sta_bw
31
32 |
```

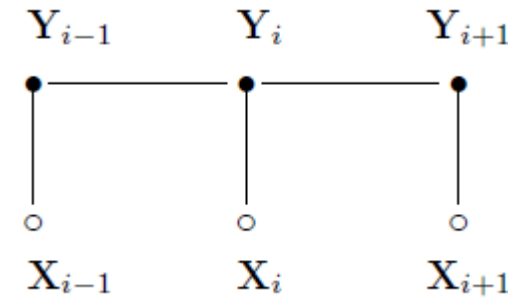
# Recurrent Neural Network (6)

---

More concrete example is there in [\*rnn1Ex.py\*](#)

# Conditional Random Field (1)

---



$$\text{➤ } P([t]_1^{|s|} | [w]_1^{|s|}) = \frac{\exp(\text{Score}([w]_1^{|s|}, [t]_1^{|s|}))}{\sum_{x \in A} \exp(\text{Score}([w]_1^{|s|}, [x]_1^{|s|}))}$$

$$\text{Score}([w]_1^{|s|}, [t]_1^{|s|}) = \sum_{1 \leq i \leq |s|} (W_{t_{i-1}, t_i}^{trans} + Z_{t_i}^{(i)})$$

# Conditional Random Field

---

`tf.contrib.crf.crf_log_likelihood( US, True_labels, lengths )`

*Arguments:*

1. `US` : *( N,M, c), tensor of unary potential scores for batch*
2. `True_labels` : *(N,M), matrix of true label for every word*
3. `lengths` : *(N) , actual length of every sentence*

*Return:*

1. `log_like` : *(N), log likelihood of every label sequence*
2. `W_pair` : *(c,c), Pairwise parameter matrix*

`tf.contrib.crf.viterbi_decode(us, W_pair)`

*Arguments:*

1. `us` : *(c ) vector of unary potential score for a pattern*
2. `W_pair` : *(c,c) Pair wise potential*

*Return:*

1. `seq` : *(M) highest probability label sequence*
2. `score` : *A float containing containing score of viterbi*

# Conditional Random Field (2)

crfEx.py

```
1 import tensorflow as tf
2 import numpy as np
3
4 sentMax = 10          # Max length of sentence
5 num_classes = 3       # number of classes
6 num_features = 15      # number of features
7
8 X = tf.placeholder(tf.float32, [None, sentMax, num_features] ) # (6, 10, 15)
9 input_y = tf.placeholder(tf.int32, [None, sentMax] )           # (6, 10)
10 X_lengths = tf.placeholder(tf.int32, [None] )                  # (6)
11 X1 = tf.reshape( X, [-1, num_features] )                       # (6*10, 15)
12
13 #Fully connected layer operations
14 W_ff = tf.Variable(tf.random_uniform([num_features, num_classes], -1.0, +1.0), name="W") # (15,3)
15 b_ff = tf.Variable(tf.constant(0.1, shape=[num_classes]), name="b")
16 H1 = tf.nn.xw_plus_b(X1, W_ff, b_ff, name="H1")                # ( 6*10, 3)
17
18 Z1 = tf.reshape(H1, [-1, sentMax, num_classes] )              # ( 6, 10, 3)
19
20 #CRF layer
21 log_likelihood, transition_params = tf.contrib.crf.crf_log_likelihood( Z1, input_y, X_lengths )
22 loss = tf.reduce_mean(-log_likelihood)
23
24 # Create input data
25 x = np.asarray( np.random.randn(6, sentMax, num_features), dtype='float32' ) # dataset: #sentence=6, #words = 10, #features = 15
26 y = np.random.randint(3, size=[6, sentMax])                        # 3 number of labels ( B, I, O )
27 x_lengths = [8, 5, 7, 9, 10, 4]                                     # actual length of each sentence
28
29 with tf.Session() as sess:
30     sess.run( tf.initialize_all_variables() )
31     l, tp = sess.run([loss, transition_params], {X:x, X_lengths:x_lengths, input_y:y } )
32     print l
33
34     #Decoding
35     us, ps = sess.run([Z1, transition_params], {X:x, X_lengths:x_lengths, input_y:y } )
36     viterbi_sequence,_ = tf.contrib.crf.viterbi_decode(us[1], ps)    # highest scoring sequence.
37     print 'true seq', y[1]
38     print 'pred seq', viterbi_sequence
```

# Conditional Random Field (3)

---

**crfEx1.py** : *Bi-RNN + CRF model for sequence labeling with dummy data*

---

**Thanks**

**Code used here are available : <https://github.com/sunilitggu/tensorflow-tutorial>**