



CS 20SI:  
TensorFlow for Deep Learning  
Lecture 2 summary

3/21/2017

# outline

1. TensorBoard
2. Randomly Generated Constants
3. Variables
4. Control Dependencies

# TensorBoard

```
import tensorflow as tf
a = tf.constant(2)
b = tf.constant(3)
x = tf.add(a, b)
with tf.Session() as sess:
    # add this line to use TensorBoard.

    (**) writer = tf.summary.FileWriter('./graphs', sess.graph)

    print sess.run(x)
writer.close()      # close the writer when you're done using it
```

# TensorBoard

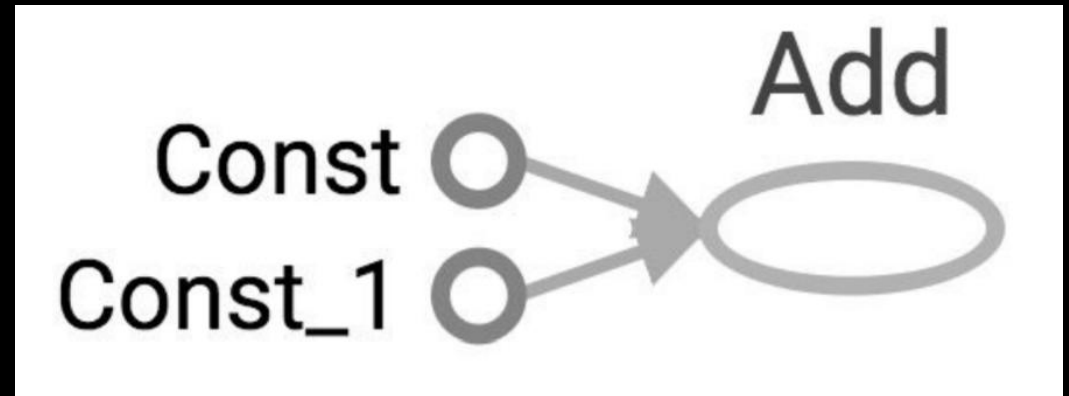
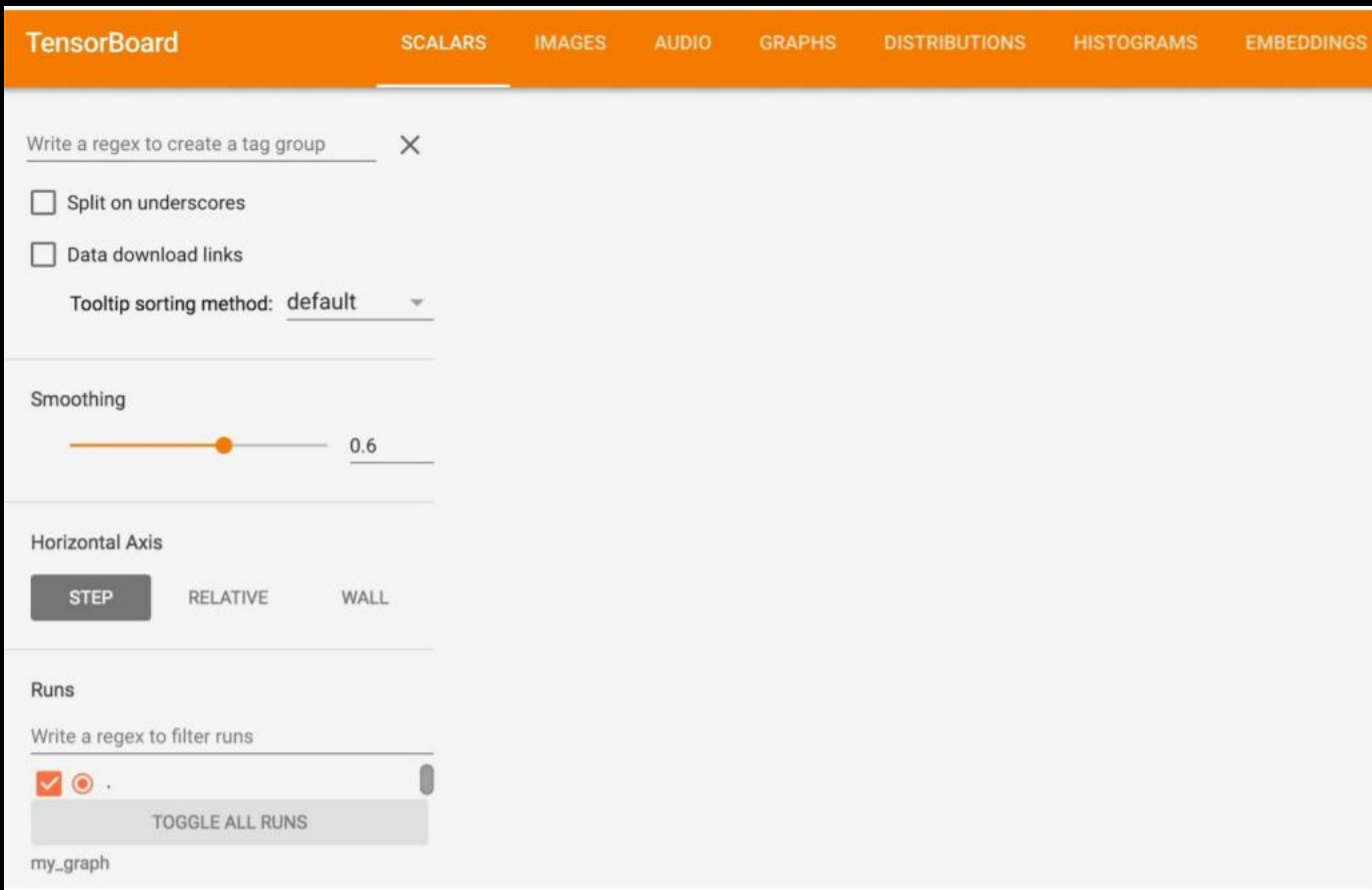
Go to terminal, run:

```
$ python [yourprogram].py
```

```
$ tensorboard --logdir="./graphs" --port 6006
```

Then open your browser and go to: <http://localhost:6006/>

# TensorBoard



TF automatically names the nodes when you don't explicitly name them. More about this next lecture!

Nodes: operators, variables, and constants

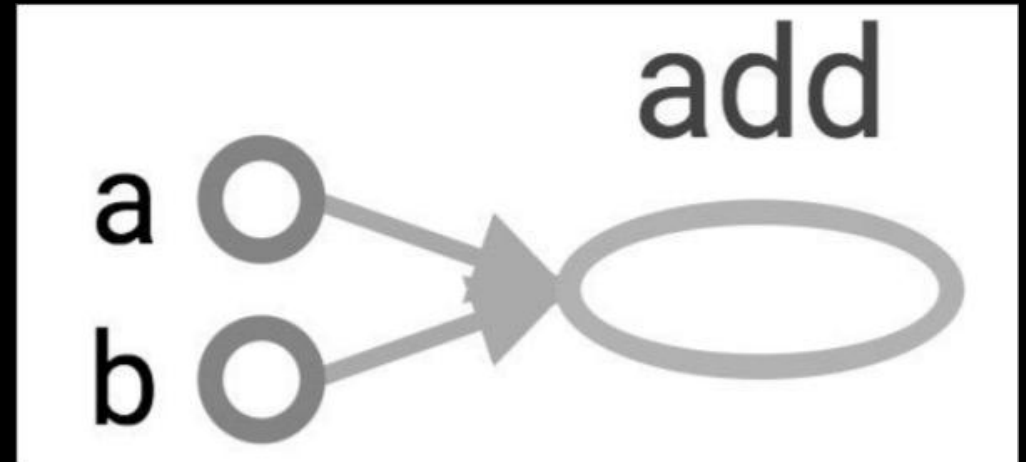
Edges: tensors

Tensors are data.

# TensorBoard

```
a = tf.constant(2, name="a")  
b = tf.constant(3, name="b")  
x = tf.add(a, b, name="add")
```

```
writer = tf.summary.FileWriter("./graphs", sess.graph)  
with tf.Session() as sess:  
    print sess.run(x)    # >> 5
```



# Randomly Generated Constants

```
tf.set_random_seed(seed)
```

# Randomly Generated Constants

To generate different sequences across sessions, set neither graph-level nor op-level seeds:

```
a = tf.random_uniform([1])
b = tf.random_normal([1])
with tf.Session() as sess1:
    print(sess1.run(a))    # generates 'A1'
    print(sess1.run(a))    # generates 'A2'
    print(sess1.run(b))    # generates 'B1'
    print(sess1.run(b))    # generates 'B2'
```

```
with tf.Session() as sess2:
    print(sess2.run(a))    # generates 'A3'
    print(sess2.run(a))    # generates 'A4'
    print(sess2.run(b))    # generates 'B3'
    print(sess2.run(b))    # generates 'B4'
```



# Randomly Generated Constants

To generate the same repeatable sequence for an op across sessions, set the seed for the op:

```
a = tf.random_uniform([1], seed=1)
```

```
b = tf.random_normal([1])
```

```
# Repeatedly running this block with the same graph will generate the same
```

```
# sequence of values for 'a', but different sequences of values for 'b'.
```

```
with tf.Session() as sess1:
```

```
    print(sess1.run(a)) # generates 'A1'
```

```
    print(sess1.run(a)) # generates 'A2'
```

```
    print(sess1.run(b)) # generates 'B1'
```

```
    print(sess1.run(b)) # generates 'B2'
```

```
with tf.Session() as sess2:
```

```
    print(sess2.run(a)) # generates 'A1'
```

```
    print(sess2.run(a)) # generates 'A2'
```

```
    print(sess2.run(b)) # generates 'B3'
```

```
    print(sess2.run(b)) # generates 'B4'
```

# Randomly Generated Constants

To make the random sequences generated by all ops be repeatable across sessions, set a graph-level seed:

```
tf.set_random_seed(1234)
a = tf.random_uniform([1])
b = tf.random_normal([1])
# Repeatedly running this block with the same graph will generate different
# sequences of 'a' and 'b'.
```

with tf.Session() as sess1:

```
print(sess1.run(a))  # generates 'A1'
print(sess1.run(a))  # generates 'A2'
print(sess1.run(b))  # generates 'B1'
print(sess1.run(b))  # generates 'B2'
```

with tf.Session() as sess2:

```
print(sess2.run(a))  # generates 'A1'
print(sess2.run(a))  # generates 'A2'
print(sess2.run(b))  # generates 'B1'
print(sess2.run(b))  # generates 'B2'
```

# Variables

tf.Variable holds several ops:

(1) `x = tf.Variable(...)`

(2) `x.initializer`        # init op

(3) `x.value()`            # read op

(4) `x.assign(...)`        # write op

(5) `x.assign_add(...)` # and more

# Variables

The easiest way is initializing all variables at once:

```
init = tf.global_variables_initializer()  
with tf.Session() as sess:  
    sess.run(init)
```

Initialize only a subset of variables:

```
init_ab = tf.variables_initializer([a, b], name="init_ab")  
with tf.Session() as sess:  
    sess.run(init_ab)
```

Initialize a single variable:

```
W = tf.Variable(tf.zeros([784,10]))  
with tf.Session() as sess:  
    sess.run(W.initializer)
```

# Variables

```
W = tf.Variable(10)
assign_op = W.assign(100)
with tf.Session() as sess:
    sess.run(W.initializer)
    print W.eval() # >> 10
    sess.run(assign_op)
print W.eval() # >> 100
```

# Control Dependencies

```
tf.Graph.control_dependencies(control_inputs)
# defines which ops should be run first
# your graph g have 5 ops: a, b, c, d, e

with g.control_dependencies([a, b, c]):
    # 'd' and 'e' will only run after 'a', 'b', and 'c' have executed.
    d = ...
    e = ...
```

谢谢大家！