TensorBoard Basic

To learn how to use TensorBoard, based on <u>notes</u> (<a href="https://github.com/fluxcapacitor/pipeline/blob/master/myapps/jupyter/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamentals/TensorFlow/Fundamenta

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In [ ]: # TensorBoard Helper Functions and Constants
        # Directory to export TensorBoard summary statistics, graph data, etc.
        TB DIR = '/tmp/tensorboard/helloworld'
        def _start_tb(d):
            Private function that calls `tensorboard` shell command
            args:
              d: The desired directory to launch in TensorBoard
            !tensorboard --port=6006 --logdir=$d
        def start_tensorboard():
            Starts TensorBoard from the notebook in a separate thread.
            Prevents Jupyter Notebook from halting while TensorBoard runs.
            import threading
            threading.Thread(target= start tb, args=(TB DIR,)).start()
            del threading
        def stop_tensorboard():
            Kills all TensorBoard processes
            !ps -aef | grep "tensorboard" | tr -s ' ' | cut -d ' ' -f2 | xargs kill ·
        def reset tensorboard():
            stop_tensorboard()
            start tensorboard()
```

```
In [ ]: # Import core TensorFlow libraries
   import tensorflow as tf
   import numpy as np
```

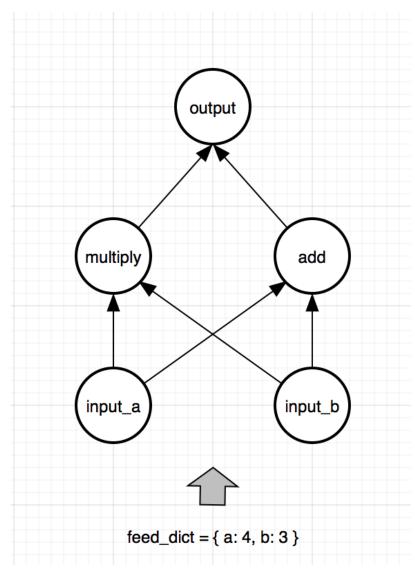
The first graph for TensorFlow

- nodes represent computation
- links represent data flow, which can have n-diamension, 0=scale, 1=vector, 2=matrix.

TensorFlow provides a placeholder operation that must be fed with data on execution. See <u>doc</u> (<u>https://www.tensorflow.org/versions/r0.9/api_docs/python/io_ops.html</u>) for details. We can set default value for placeholder too.

1-Define a Computational Graph

output =
$$c + d = (a + b) + (a * b)$$



```
In [ ]: # `tf.placeholder` creates an "input" node- we will give it value when we rus
    a = tf.placeholder(tf.int32, name="input_a")
    b = tf.placeholder(tf.int32, name="input_b")

# there are two nodes, named input_a and input_b, datatype is int32. They wis

In [ ]: # `tf.add` is operation "add", and we call/name the node "add", may be using
    c = tf.add(a, b, name="add") # this give us c = a + b

In [ ]: # `tf.mul` creates a multiplication node
    d = tf.mul(a, b, name="multiply") # this gives us d = a * b
In [ ]:
```

```
# Add up the results of the previous two nodes
out = tf.add(c, d, name="output") # this gives us
```

```
In [ ]: # OPTIONAL
# Create a scalar summary, which will log the value we tell it to when execut
# In this case, we'll tell it to save our output value from `out`
# This works in tandem with our SummaryWriter below
# To create the summary, we pass in two parameters:
# 1. A 'tag', which gives a label to the data
# 2. The value(s) we'd like to save
# We also give a `name` to the summary itself (does not affect behavior)
out_summary = tf.scalar_summary("output", out, name="output_summary")
```

2-Run the Graph

- Start a tf.Session to launch the graph
- Setup any necessary input values
- Use a tf.train.SummaryWriter to write information for TensorBoard (Recommended)
- Use Session.run() to compute values from the graph

```
In [ ]: # Start a session
        sess = tf.Session()
In [ ]: # Create a "feed dict" dictionary to define input values
        # Keys to dictionary are handles to our placeholders
        # Values to dictionary are values we'd like to feed in
        feed dict = { a: 4, b: 3 }
In [ ]: # OPTIONAL
        # Opens a `SummaryWriter` object, which can write stats about the graph to d.
        # We pass in two parameters into the SummaryWriter constructor
        # The first is a string, specifies a directory to write to.
          (Note: `TB DIR` was specified earlier. "TB" stands for TensorBoard
        # The second parameter passes in our graph. This allows us to visualize our
        writer = tf.train.SummaryWriter(TB DIR, graph=sess.graph)
In [ ]: # Execute the graph using `sess.run()`, passing in two parameters:
        # The first parameter, `fetches` lists which node(s) we'd like to receive as
        # The second parameter, `feed dict`, feeds in key-value pairs
        # to input or override the value of nodes
        # In this case, we run both the output value, as well as its scalar summary
        result, summary = sess.run([out, out summary], feed dict=feed dict)
        # Print output with fun formatting
        print("(({0}*{1}) + ({0}+{1})) = ".format(feed dict[a], feed dict[b]) + str(1)
```

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In []: # We add the summary to our SummaryWriter, which writes the data to disk:
    # Normally, these summaries are used to generate statistics over time
    # TensorBoard doesn't do well visualizing single points, so we fake a "globa.
    # With two points, it will generate a line
    writer.add_summary(summary, global_step=0)
    writer.add_summary(summary, global_step=100)
In []: # We're done! Close down our Session and SummaryWriter to tidy up.
    sess.close()
    writer.close()
```

3-Visualize with Tensorboard

```
In [ ]: # Start TensorBoard
start_tensorboard()
In [ ]: # Once you are done, stop TensorBoard
# stop_tensorboard()
In [ ]:
```