



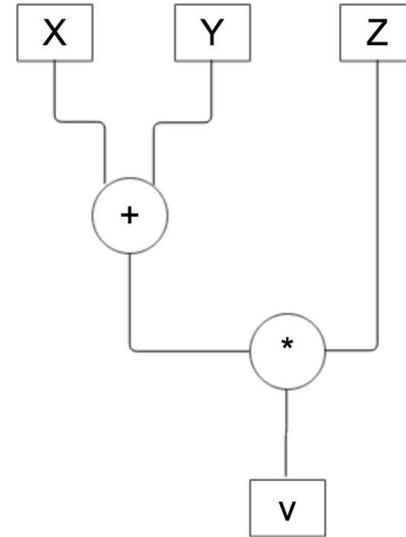
Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

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Background

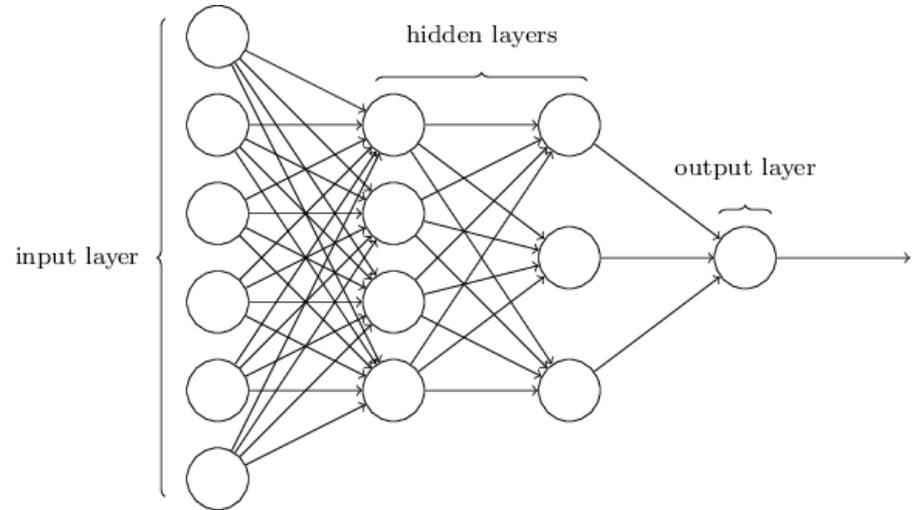
- Dataflow graphs
 - Represent data dependencies between operations
 - **Node**: a unit of computation
 - **Edges**: Data flowing in/out of computation node



$$V = (X+Y) * Z$$

Background

- Deep learning models
 - State of the art for many problems in ML
 - Can get very big!
 - Interpretability is a big issue



TensorFlow

- An API for building neural networks
 - Simplifies the coding process
 - Built around the idea of a dataflow graph
 - Edges are tensors
 - Nodes are mathematical operations

Goals

- Provide an intuitive overview
- Recognize similarities and differences between graph components
- Examine nested structure of a high level component
- Inspect details of low level operations
- Examine quantitative data

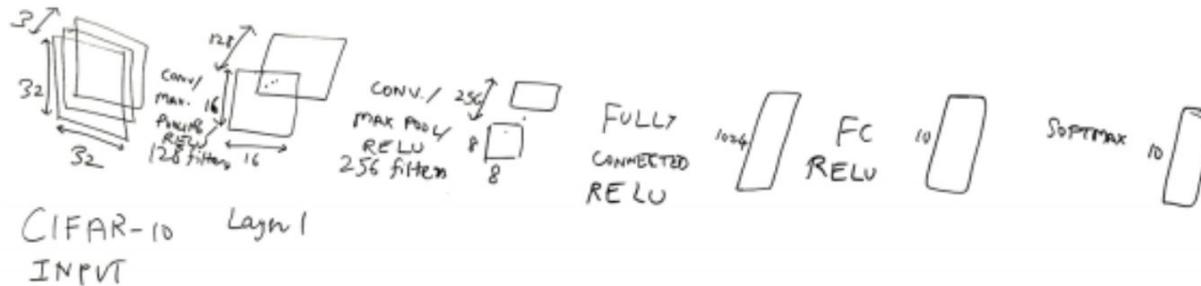
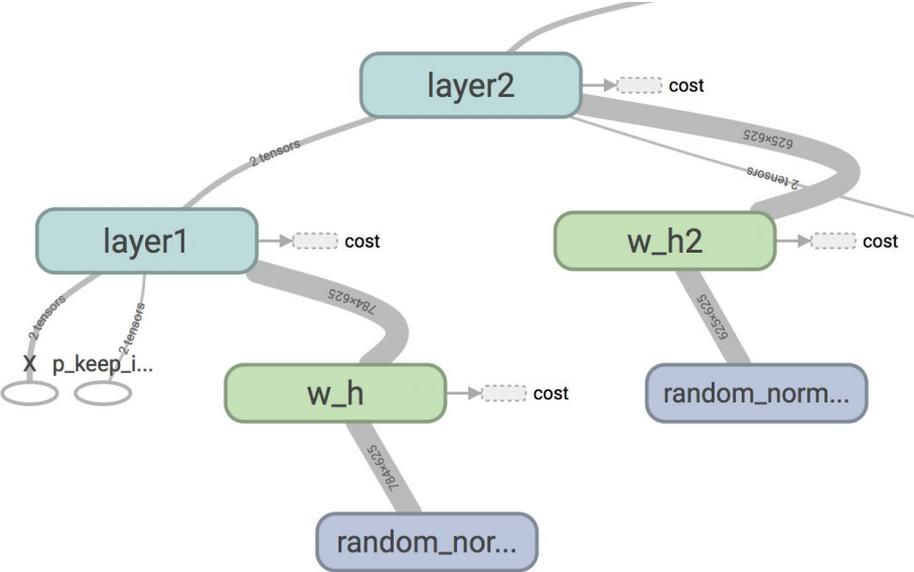


Fig 2. From [Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow](#) Kanit Wongsuphasawat et al. *EEE Trans. Visualization & Comp. Graphics (Proc. VAST)*, 2018

Basic Encoding



Problem

- Standard graph layout techniques don't work well out of the box
 - Mismatch between graph topology and semantics
 - Graph heterogeneity = Not all nodes are created equal
 - Interconnected nodes

Graph Transformations

Remove non-critical nodes

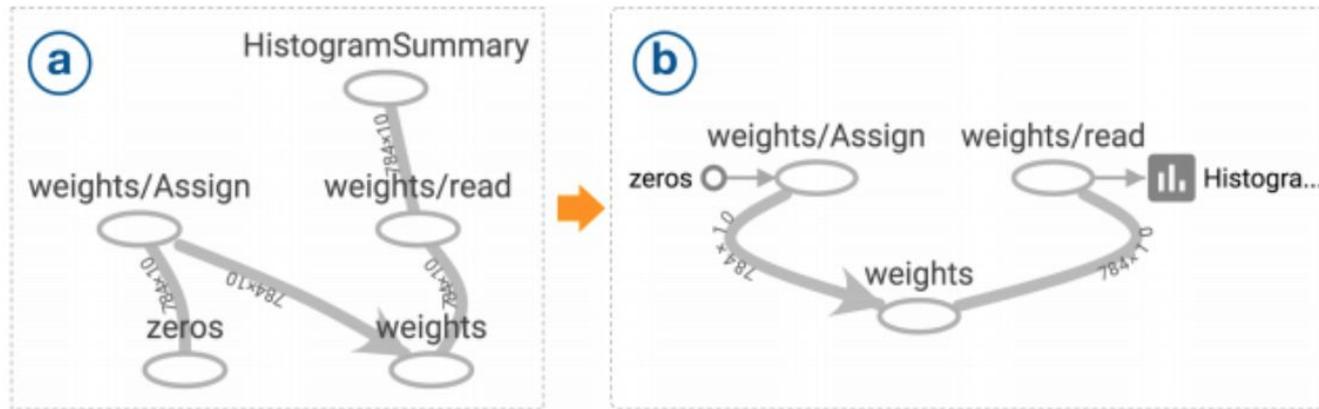


Fig 3. From [Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow](#) Kanit Wongsuphasawat et al. *EEE Trans. Visualization & Comp. Graphics (Proc. VAST)*, 2018

Graph Transformations

- Build a hierarchically clustered graph
 - Group operations based on namespaces
 - Bundle edges between groups

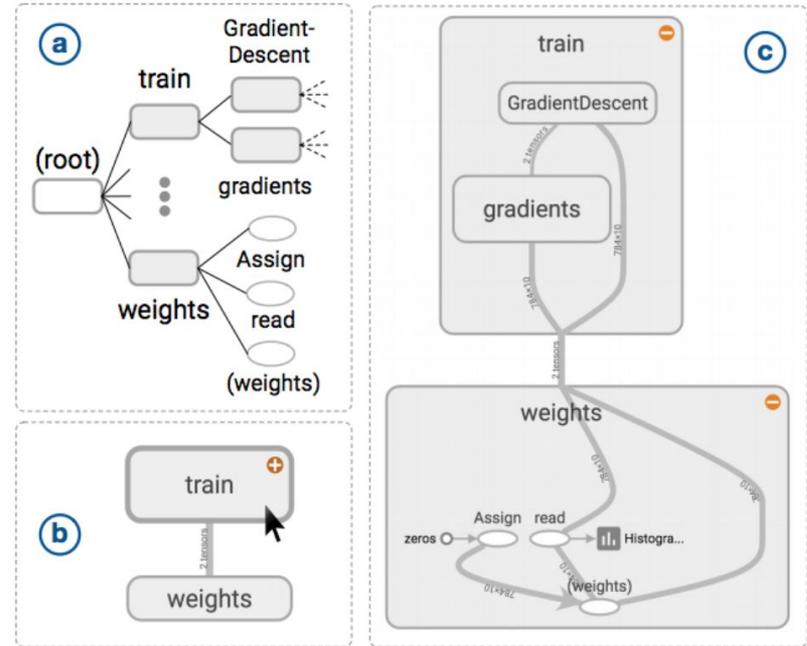


Fig 4. From [Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow](#)
 Kanit Wongsuphasawat *et al.* *EEE Trans. Visualization & Comp. Graphics*
 (Proc. VAST), 2018

Namespaces

```
with tf.name_scope("xent"):  
    xent = tf.reduce_mean(  
        tf.nn.softmax_cross_entropy_with_logits(logits=logits, labels=y))  
  
with tf.name_scope("train"):  
    train_step = tf.train.AdamOptimizer(1e-4).minimize(xent)  
  
with tf.name_scope("accuracy"):  
    correct_prediction = tf.equal(tf.argmax(logits, 1), tf.argmax(y, 1))  
    accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
```

[Google Developers]. (2017, Feb 15). Hands-on Tensorboard (TensorFlow Dev Summit 2017) [Video File]. Retrieved from <https://www.youtube.com/watch?v=eBbEDRsCmv4>

Namespaces

```
# Setup placeholders, and reshape the data
x = tf.placeholder(tf.float32, shape=[None, 784], name="x")
x_image = tf.reshape(x, [-1, 28, 28, 1])
y = tf.placeholder(tf.float32, shape=[None, 10], name="labels")

conv1 = conv_layer(x_image, 1, 32, "conv1")
conv2 = conv_layer(conv1, 32, 64, "conv2")

flattened = tf.reshape(conv2, [-1, 7 * 7 * 64])
fc1 = fc_layer(flattened, 7 * 7 * 64, 1024, "fc1")
logits = fc_layer(fc1, 1024, 10, "fc2")
```

[Google Developers]. (2017, Feb 15). Hands-on Tensorboard (TensorFlow Dev Summit 2017) [Video File]. Retrieved from <https://www.youtube.com/watch?v=eBbEDRsCmv4>

Graph Transformations

Extract Auxiliary nodes from the graph

- Many high-degree nodes turn out to be not that important
 - NoOp nodes
 - Declaring/initializing variables
 - Nodes that compute statistics
- Nodes extracted and put to the right of the whole graph
- Small proxy icons to embedded in neighbouring nodes

Graph Transformations

Deep Learning models have highly repeated structure

- Group nodes with identical subgraphs have the same color
- Finding these subgraphs is NP-hard

VAD Analysis

- **What: Data**
 - Network
- **What: Derived**
 - Cluster hierarchy
 - Edges bundled
 - Nodes removed or embedded
- **Scale:**
 - 1000s of nodes
- **How: Encode**
 - Ellipse/Rectangle marks for nodes, connection marks for links
- **Why: Tasks**
 - Make graph understandable



Demo

Critique

Positive

- Provides an intuitive overview of the dataflow graph
- Makes it easier to debug deep learning models
- Infinitely better than sketches

Negative

- No two-way editing
- Can be hard to tell quickly how the data flows through the network
- Debugging could be improved if integrated with tests
- Diff-viewer would also aid debugging