

# TensorFlow 101

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# Outline

- Basics in TensorFlow
- Basics in TensorBoard

# Basic Operations in TF

1. Operations
2. Tensors

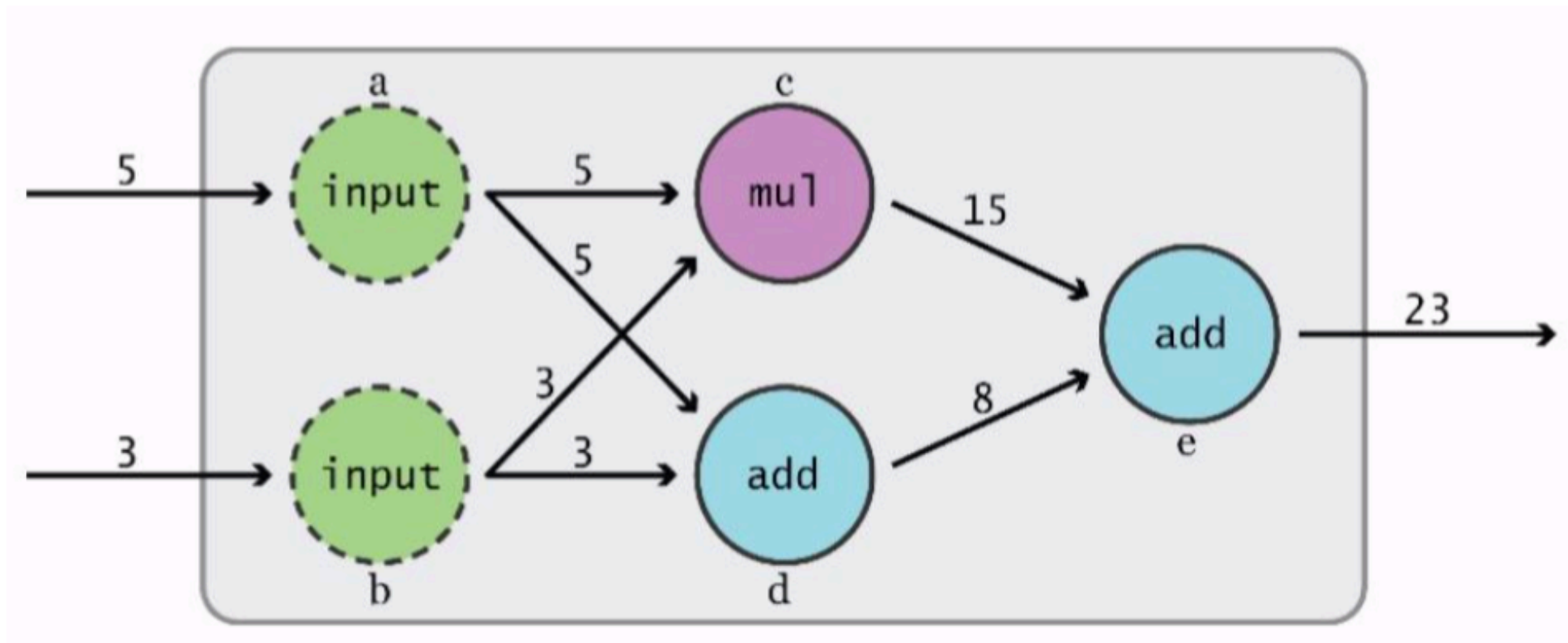
## **Idea:**

- Create a model that consists of a set of operations
- Feed the data into the models
- Tensors will flow between operations to get an ‘output’ tensor !!!

# Data Flow Graphs

- TensorFlow separates definitions of computations from their execution
- Its execution is decided into 2 phases *i.e.*
  - Phase 1: assemble a graph
  - Phase 2: Use a session to execute operations in the graph

# Data Flow Graphs



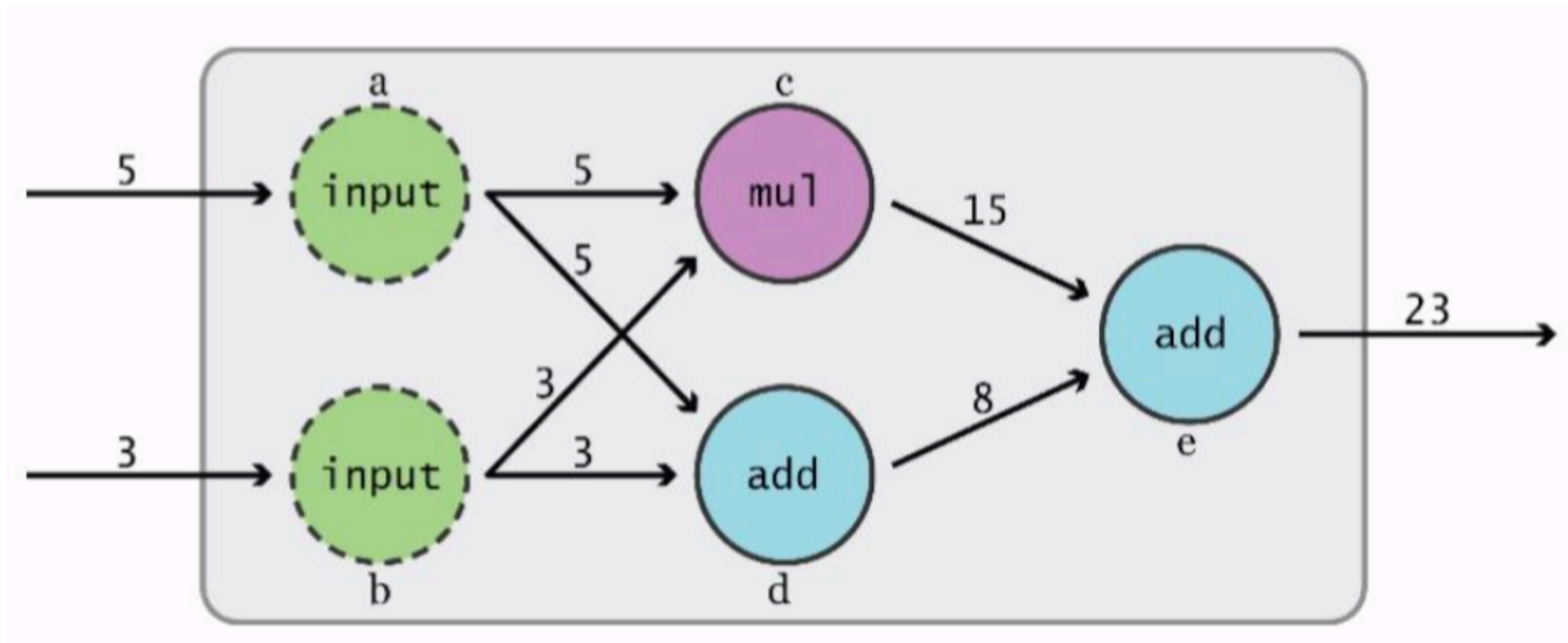
**Nodes:** operators, variables, constants, and placeholders

**Edges:** tensors

# What's a tensor?

- ▶ An n-dimensional array
- ▶ 0-d tensor: scalar (number)
- ▶ 1-d tensor: vector
- ▶ 2-d tensor: matrix
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# Data Flow Graphs



Tensors are data !

∴ TensorFlow  $\longleftrightarrow$  DataFlow



# Why graph?

- Save computation *i.e.* only subgraphs corresponding to an executed operation are run !
- Facilitate distributed computation, spread the work across multiple CPUs, GPUs, or devices
- Many common machine learning models are commonly taught and visualized as directed graphs.



Let's do our 1<sup>st</sup> lab !

# What's TensorBoard?

- TensorBoard is a visualization software that comes with any standard TensorFlow installation.
- Google's word: *The computations you'll use TensorFlow for many things (like training a massive deep neural network) and they can be complex and confusing. To make it easier to understand, debug, and optimize TensorFlow programs, we've included a suite of visualization tools called TensorBoard.*

# What to visualize?

1. Visualizing the graph
2. Writing summaries to visualize learning

To use TensorBoard, we need to **create event log files** using the code:

```
writer = tf.summary.FileWriter([logdir], [graph])
```

# Visualizing a graph

To use TensorBoard, we need to **create event log files** using the code:

```
writer = tf.summary.FileWriter([logdir], [graph])
```

There are 2 ways to get the graph:

1. Call the graph using **tf.get\_default\_graph()**, which returns the default graph of the program
2. Set it as **session.graph** which returns the session's graph. This requires us to have a session created (**This is more common**).

**Either way, make sure to create a writer only after defining the graph.**

**Otherwise, the graph visualized on TensorBoard would be incomplete !!!**

# Types of summary

1. Scalars
2. Images
3. Audio
4. Histograms
5. Graphs

**A summary is a special operation that TensorBoard takes in a regular tensor and outputs to your disk !**

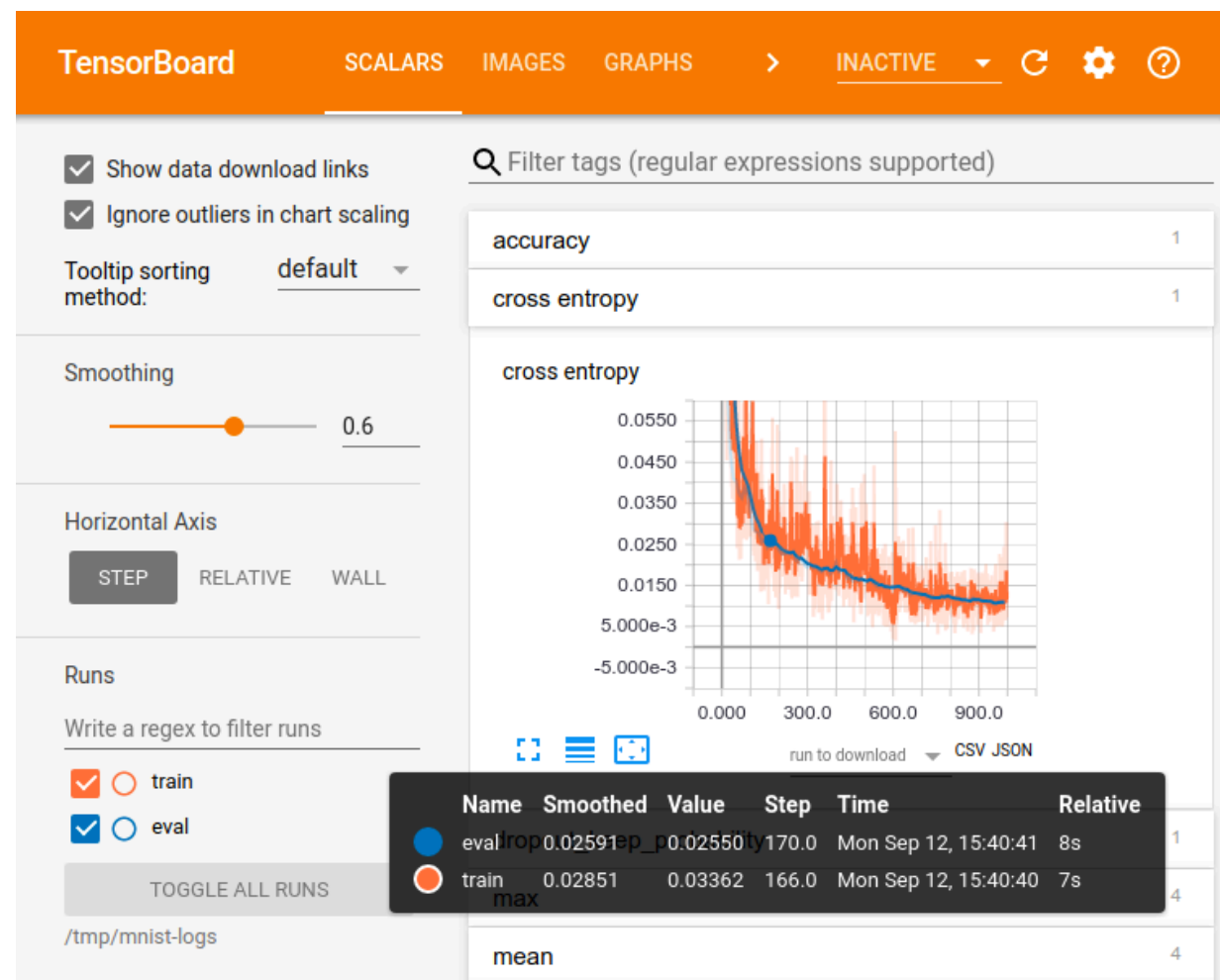


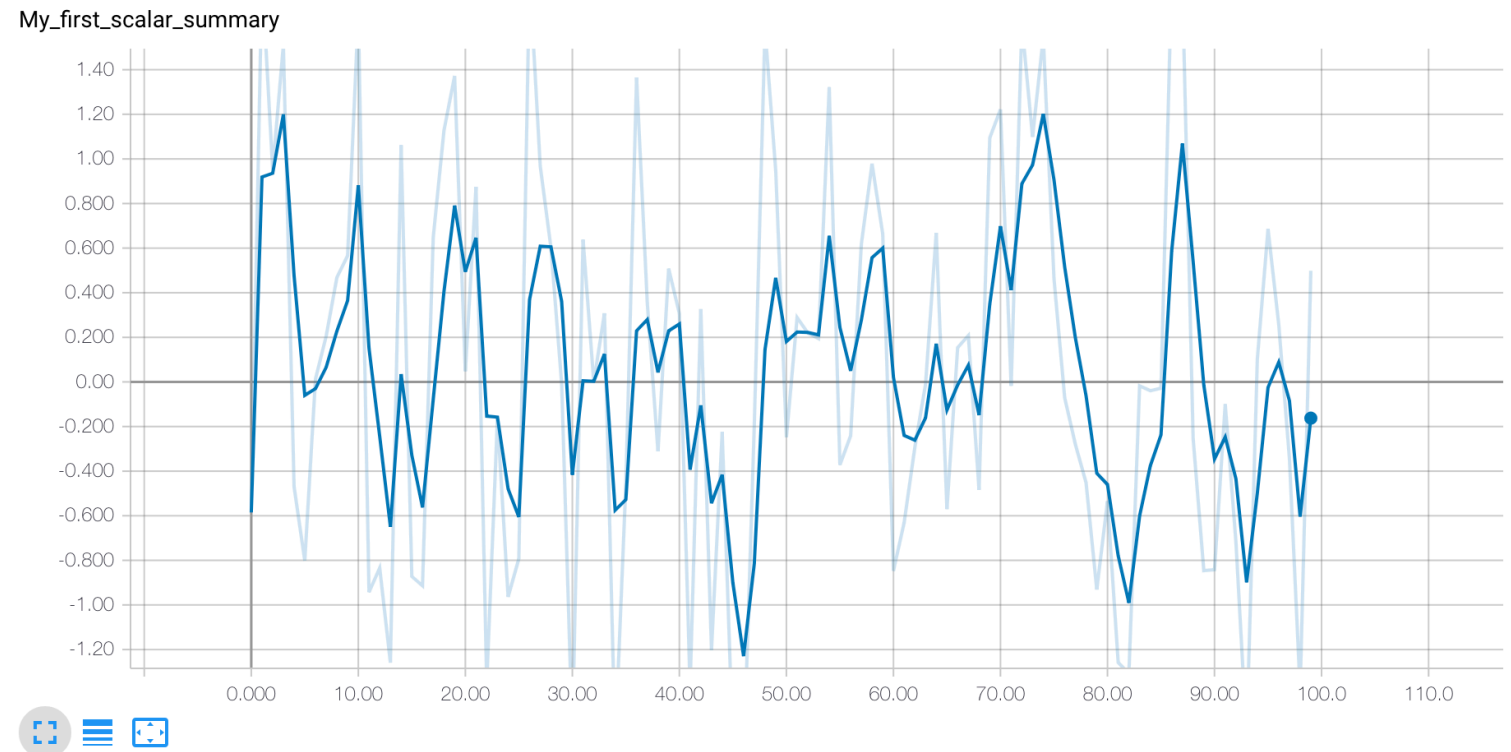
Figure: TensorBoard's appearance

# tf.summary.scalar

- Write the values of a scalar tensor that changes over time or iterations.
- It's usually used to monitor the changes of loss function or classification accuracy.

# Example 1

- Provide a code to randomly pick 100 values from a standard normal distribution,  $\mathcal{N}(0,1)$ , and plot them on TensorBoard.
- Hint: *create a variable and initialize it from a normal distribution (with  $\mu = 0$  and  $\sigma = 1$ ), then run a for loop in the session and initialize it 100 times.*



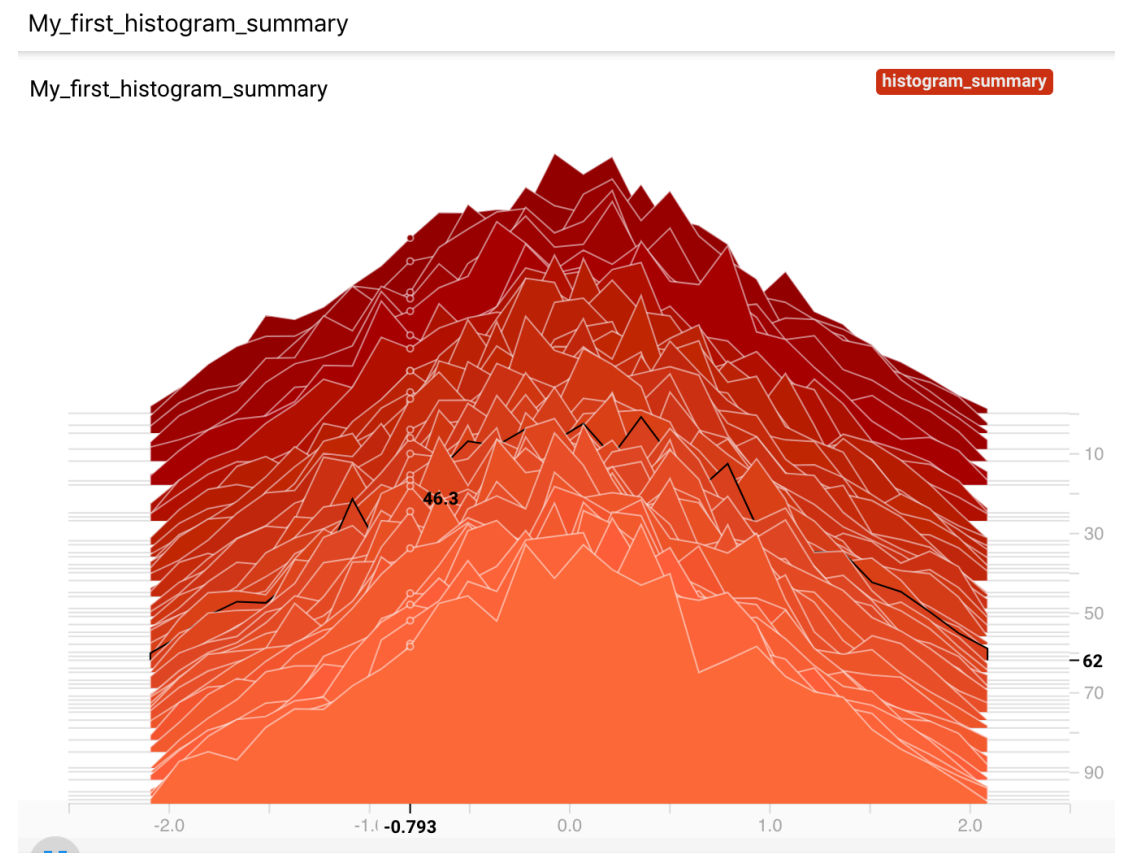
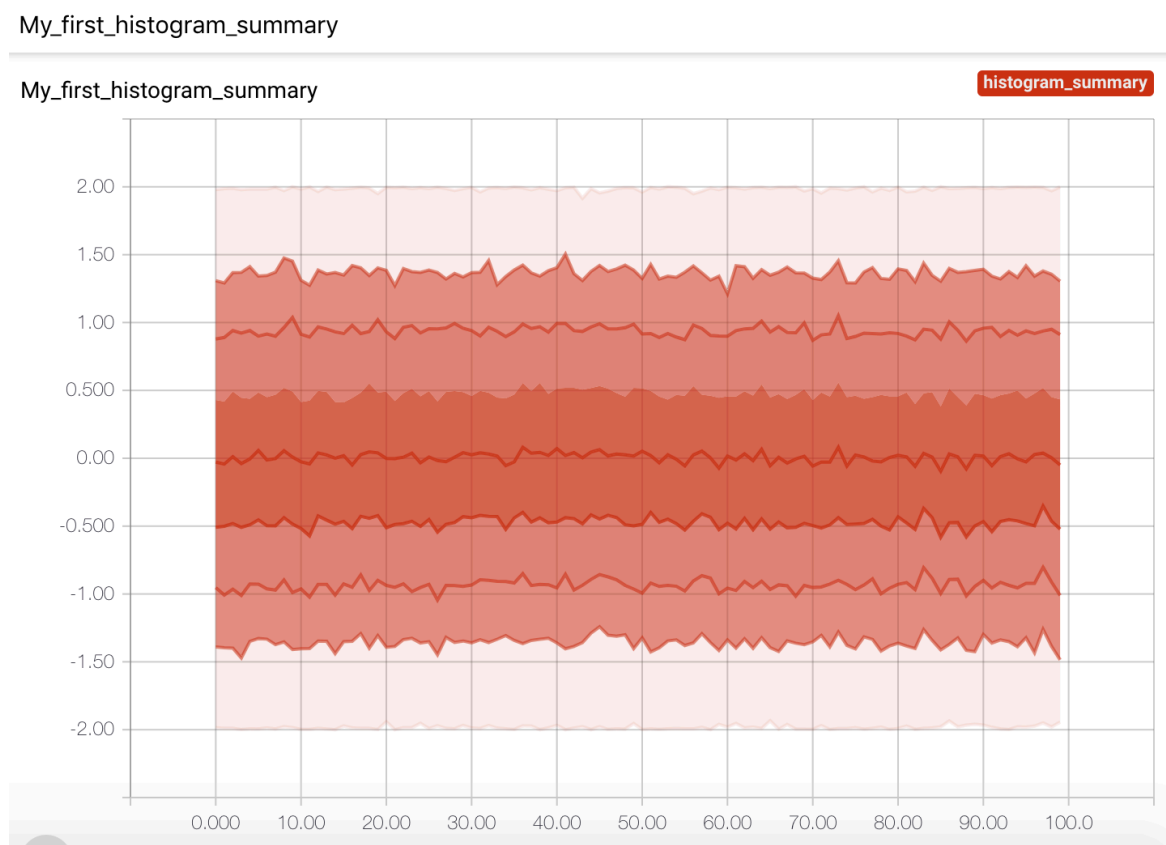
# tf.summary.histogram

- Observe the change of a value over time or iterations.
- Used for plotting the histogram (and the distribution) of the values of a non-scalar tensor.
- In NN, it's commonly used to monitor the changes of weight and bias distributions.
- Helps detecting irregular behavior of the network parameters *e.g.* when our weights explode or shrink abnormally.



# Example 2

- Continue the previous example by adding a matrix of size 30x40, whose entries come from a standard normal distribution. Initialize this matrix 100 times and plot the distribution of its entries over time.



# Writing summaries

- In practice, we can use any number of summaries to track different parameters in our model.
  - ▶ This makes running and writing the summaries extremely insufficient.
- To handle this, we have to merge all summaries in our graph and run them at once inside your session.
  - ▶ use **`tf.summary.merge_all()`**

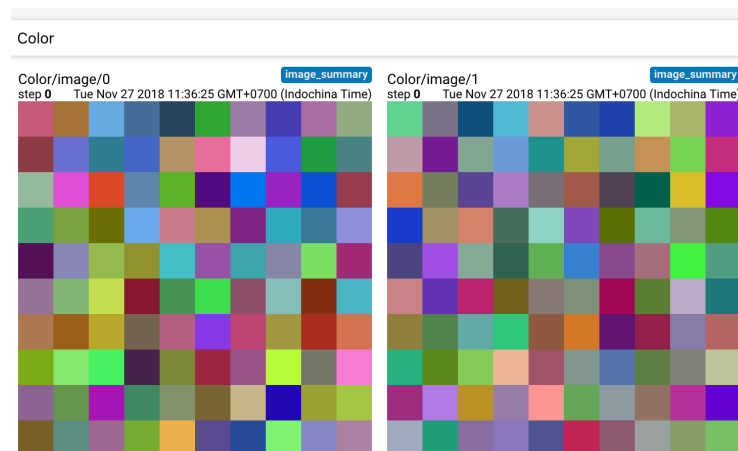
# tf.summary.image

- Used for writing and visualizing tensors as images
- In NN, usually used for tracking images that are fed to the network (*e.g.* in each batch) or the images generated in the output.
- In general, this can be used to plot any tensors.

# Example 3

Let's define two variables:

1. Of size 30x10 as 3 grayscale images of size 10x10
  2. Of size 50x30 as 5 color images of size 10x10
- and plot them as images in TensorBoard.



# Further reading

- TensorBoard Github: <https://github.com/tensorflow/tensorboard>
- TensorBoard Graph Visualization: [https://www.tensorflow.org/guide/graph\\_viz](https://www.tensorflow.org/guide/graph_viz)
- TensorBoard Visualized Learning: [https://www.tensorflow.org/guide/summaries\\_and\\_tensorboard](https://www.tensorflow.org/guide/summaries_and_tensorboard)
- <https://itnext.io/how-to-use-tensorboard-5d82f8654496>