

TensorFlow

Tom Rochette <tom.rochette@coreteks.org>

August 30, 2025 — [861fb9d0](#)

0.1 Context

TensorFlow has rapidly grown in popularity due to the fact that it is developed/supported by Google. As more and more developers move to the platform, it becomes essential to learn how it works and have a general idea of the various concepts it makes use of. This is a short article about some of these concepts.

0.2 Learned in this study

0.3 Things to explore

1 Overview

- Computations are represented as graphs
- Graphs are executed in the context of `Sessions`

2 Building a graph

- Start with ops that do not need any input (called `source ops`), such as `Constant`

3 Session

- Graphs are executed within a session (context)
`session = tf.session()`
- Sessions are given one or many tensors to resolve
`session.run([tensorA, tensorB])`
- Once we're done with a session, it should be closed
`session.close()`

4 Tensors

A tensor is simply a multidimensional array of data. A scalar is a 0-D tensor, a vector is a 1-D tensor, a matrix is a 2-D tensor and anything over 3-D is called an n-D tensor.

Rank: The number of dimensions of a tensor.

Rank	Math entity	Example
0	Scalar	$s = 483$
1	Vector	$v = [1.1, 2.2, 3.3]$
2	Matrix	$m = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$
3	3-Tensor	$t = \begin{bmatrix} [[2], [4], [6]], [[8], [10], [12]], [[14], [16], [18]] \end{bmatrix}$

Shape: A vector describing the number of elements at each point within a dimension.

Rank	Shape	Dimension number	Example
0	<code>[]</code>	0-D	A 0-D tensor. A scalar.
1	<code>[D0]</code>	1-D	A 1-D tensor with shape <code>[5] = [1, 2, 3, 4, 5]</code> .
2	<code>[D0, D1]</code>	2-D	A 2-D tensor with shape <code>[3, 4] = [[1, 2, 3, 4], [1, 2, 3, 4], [1, 2, 3, 4]]</code> .
3	<code>[D0, D1, D2]</code>	3-D	A 3-D tensor with shape <code>[1, 4, 3] = [[[1, 2, 3], [1, 2, 3], [1, 2, 3], [1, 2, 3]]]</code> .
n	<code>[D0, D1, ..., Dn]</code>	n-D	A tensor with shape <code>[D0, D1, ..., Dn]</code> .

Type: Type of the data contained within the tensor.

Data type	Description
<code>DT_FLOAT</code>	32 bits floating point.
<code>DT_DOUBLE</code>	64 bits floating point.
<code>DT_INT64</code>	64 bits signed integer.
<code>DT_INT32</code>	32 bits signed integer.
<code>DT_INT16</code>	16 bits signed integer.
<code>DT_INT8</code>	8 bits signed integer.
<code>DT_UINT8</code>	8 bits unsigned integer.
<code>DT_STRING</code>	Variable length byte arrays. Each element of a Tensor is a byte array.
<code>DT_BOOL</code>	Boolean.
<code>DT_COMPLEX64</code>	Complex number made of two 32 bits floating points: real and imaginary parts.
<code>DT_QINT32</code>	32 bits signed integer used in quantized Ops.
<code>DT_QINT8</code>	8 bits signed integer used in quantized Ops.
<code>DT_QUINT8</code>	8 bits unsigned integer used in quantized Ops.

5 Variables

- Variables must be initialized (`tf.initialize_all_variables()`)
- Initialization is an operation, and thus must be executed within a session

6 Fetches

- All the ops needed to produce the values of requested tensors are run once (not once per requested tensor)

7 Feeds

- Temporarily replaces the output of an operation with a tensor value (act as a placeholder)
- The feed data is provided as an argument to a `session.run()` call
`sess.run([output], feed_dict={input1:[7.], input2:[2.]})`

8 Operations/Functions of interest

8.1 CNN

- `tf.nn.conv2d(input, kernel, strides, padding)`: apply a convolution using kernel
- `tf.nn.relu(input)`: rectifier linear unit, every negative value is set to 0, and positive values are kept the same
- `tf.nn.sigmoid(input)`: returns a value in the range [0.0, 1.0]
- `tf.nn.tanh(input)`: returns a value in the range [-1.0, 1.0]
- `tf.nn.dropout(input, keep_prob)`: set the output to 0.0 based on a given probability. The output is multiplied by 1/keep_prob in order to keep the expected sum unchanged
- `tf.nn.max_pool(input, kernel, strides, padding)`: take the maximum value found within a certain kernel size
- `tf.nn.avg_pool(input, kernel, strides, padding)`: averages out all the values at each depth found within a kernel size
- `tf.nn.local_response_normalization`

8.2 RNN

- `tf.nn.rnn_cell.BasicRNNCell(num_neurons)`: declares a recurrent neural network cell
- `tf.nn.dynamic_rnn(network, input)`: simulate the given RNN
- `tf.nn.rnn_cell.LSTMCell(num_neurons)`: declares a long short-term memory neural network cell
- `tf.nn.rnn_cell.GRUCell(num_neurons)`: declares a gated recurrent unit cell

9 CNN

- Used mostly to process high density matrices where the data surrounding a value is generally highly correlated with it
- Apply the convolution operator to a 2d matrix using a given kernel/filter

10 RNN

- Used to process sequential inputs (speech recognition, speech synthesis, connected handwriting recognition, time-series forecast, image caption generation, end-to-end translation)

11 See also

12 References

- <https://www.tensorflow.org/>
- <https://medium.com/jim-fleming/loading-tensorflow-graphs-via-host-languages-be10fd81876f>