

# TensorFlow

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December 21, 2025 — [77e1b28a](#)

## 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    | $t = \begin{bmatrix} \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix} & \begin{bmatrix} 8 \\ 10 \\ 12 \end{bmatrix} & \begin{bmatrix} 14 \\ 16 \\ 18 \end{bmatrix} \end{bmatrix}$ |

**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>