

With TF 1.0!



Lab I

TensorFlow Basics

Sung Kim <hunkim+ml@gmail.com>

Code: <https://github.com/hunkim/DeepLearningZeroToAll/>



Call for comments

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Other slides: <https://goo.gl/jPtVNt>



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An open-source software library for Machine Intelligence

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We're excited to announce the release of TensorFlow 1.0! Check out the migration guide to upgrade your code with ease.

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Dynamic graphs in TensorFlow

We've open-sourced TensorFlow Fold to make it easier than ever to work with input data with varying shapes and sizes.

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The 2017 TensorFlow Dev Summit

Thousands of people from the TensorFlow community participated in the first flagship event. Watch the keynote and talks.

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<https://www.tensorflow.org>

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With TF 1.0!

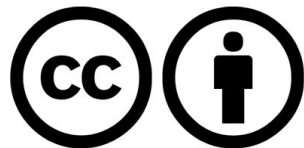


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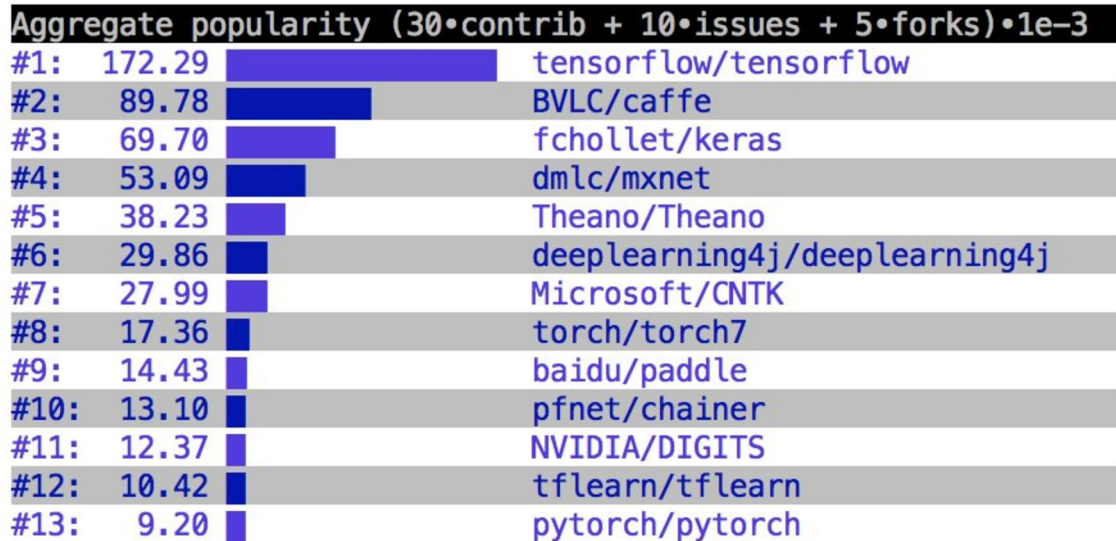
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TensorFlow

Deep learning libraries:
Accumulated GitHub metrics



Deep learning libraries: growth over past three months

new contributors from 2016-10-09 to 2017-02-10

#1:	192	tensorflow/tensorflow
#2:	89	dmlc/mxnet
#3:	78	fchollet/keras
#4:	42	baidu/paddle
#5:	29	Microsoft/CNTK
#6:	23	pfnet/chainer
#7:	21	Theano/Theano
#8:	20	deeplearning4j/deeplearning4j
#9:	20	tflearn/tflearn
#10:	19	BVLC/caffe
#11:	9	torch/torch7
#12:	3	NVIDIA/DIGITS

new forks from 2016-10-09 to 2017-02-10

#1:	6525	tensorflow/tensorflow
#2:	1822	BVLC/caffe
#3:	1316	fchollet/keras
#4:	999	dmlc/mxnet
#5:	909	deeplearning4j/deeplearning4j
#6:	887	Microsoft/CNTK
#7:	324	tflearn/tflearn
#8:	321	baidu/paddle
#9:	287	Theano/Theano
#10:	257	torch/torch7
#11:	175	NVIDIA/DIGITS
#12:	142	pfnet/chainer

new issues from 2016-10-09 to 2017-02-10

#1:	1563	tensorflow/tensorflow
#2:	979	fchollet/keras
#3:	871	dmlc/mxnet
#4:	646	baidu/paddle
#5:	486	Microsoft/CNTK
#6:	361	deeplearning4j/deeplearning4j
#7:	318	BVLC/caffe
#8:	217	NVIDIA/DIGITS
#9:	214	Theano/Theano
#10:	167	tflearn/tflearn
#11:	150	pfnet/chainer
#12:	90	torch/torch7

aggregate metrics growth from 2016-10-09 to 2017-02-10

#1:	54.01	tensorflow/tensorflow
#2:	18.71	fchollet/keras
#3:	16.38	dmlc/mxnet
#4:	12.86	BVLC/caffe
#5:	10.17	Microsoft/CNTK
#6:	9.32	baidu/paddle
#7:	8.75	deeplearning4j/deeplearning4j
#8:	4.21	Theano/Theano
#9:	3.89	tflearn/tflearn
#10:	3.14	NVIDIA/DIGITS
#11:	2.90	pfnet/chainer
#12:	2.46	torch/torch7



François Chollet @fchollet · Feb 11

Time for an update: what does the deep learning library landscape look like, seen from GitHub? [pic.twitter.com/QDZyVrYBd](https://twitter.com/QDZyVrYBd)

<https://twitter.com/fchollet/status/830499993450450944/>

TensorFlow

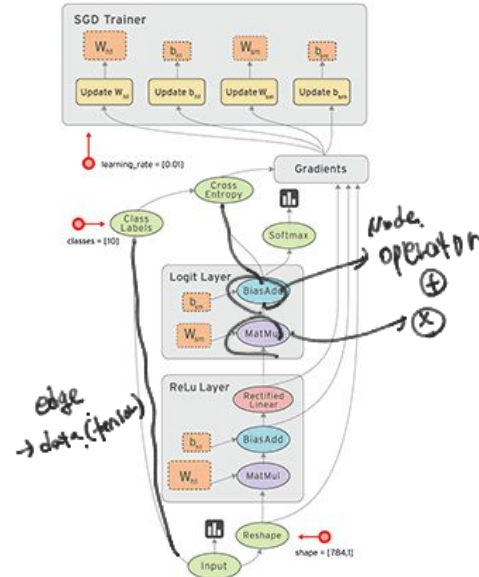
- TensorFlow™ is an open source software library for numerical computation using data flow graphs.
- Python!



<https://www.tensorflow.org/>

What is a Data Flow Graph?

- Nodes in the graph represent mathematical operations
- Edges represent the multidimensional data arrays (tensors) communicated between them.



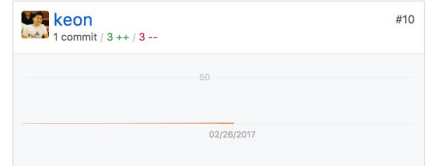
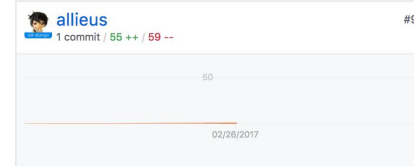
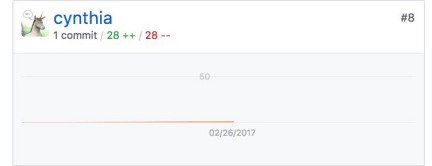
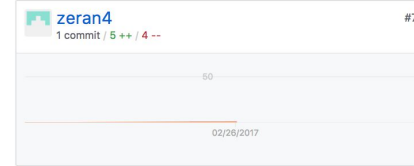
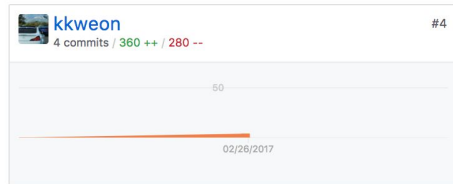
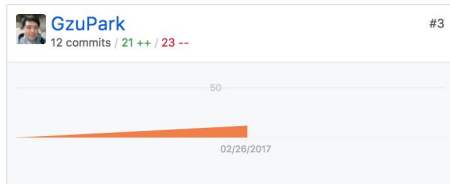
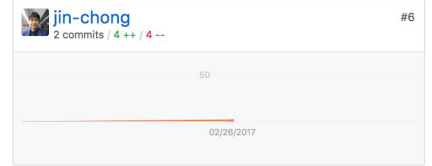
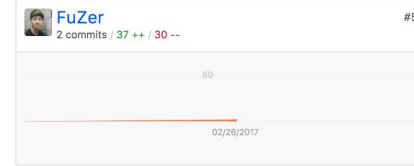
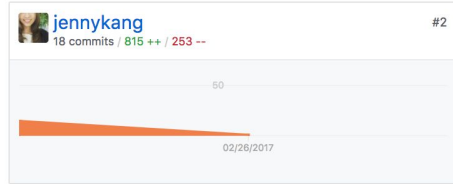
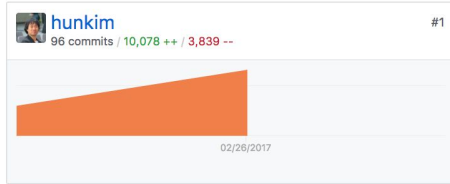
Installing TensorFlow

- Linux, Mac OSX, Windows
 - (sudo -H) pip install --upgrade tensorflow
 - (sudo -H) pip install --upgrade tensorflow-gpu
- From source
 - bazel ...
 - https://www.tensorflow.org/install/install_sources
- Google search/Community help
 - <https://www.facebook.com/groups/TensorFlowKR/>

Check installation and version

```
Sungs-MacBook-Pro:hunkim$ python3
Python 3.6.0 (v3.6.0:41df79263a11, Dec 22 2016, 17:23:13)
[GCC 4.2.1 (Apple Inc. build 5666) (dot 3)] on darwin
Type "help", "copyright", "credits" or "license" for more
information.
>>> import tensorflow as tf
>>> tf.__version__
'1.0.0'
>>>
```

<https://github.com/hunkim/DeepLearningZeroToAll/>



TensorFlow Hello World!

Hello TensorFlow!

In [2]:

```
# Create a constant op  
# This op is added as a node to the default graph
```

```
hello = tf.constant("Hello, TensorFlow!")
```

```
# start a TF session
```

```
sess = tf.Session()
```

```
# run the op and get result
```

```
print(sess.run(hello))
```

```
b'Hello, TensorFlow!'
```

b'String' 'b' indicates *Bytes literals*. <http://stackoverflow.com/questions/6269765/>

byte string 이라는 뜻 →

computational graph
사상사태
양상양상

Node 시작
값은 Hello, tensorflow
를 처리하고 있다.

① 노드 만들기 (22개의 노드)

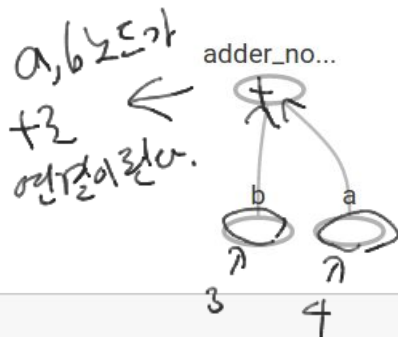
세션은 만들거고 세션이제만
computational graph를 실행할 수 있다.

② 세션 만들기

run하고 computational graph를
실행한다.

③ 22개의 노드 실행
(노드)

Computational Graph



In [4]: `node1 = tf.constant(3.0, tf.float32)`
`node2 = tf.constant(4.0) # also tf.float32 implicitly`
`node3 = tf.add(node1, node2)` → `node3 = node1 + node2`

Handwritten notes: "dtype을 정하지 않으면 float32 default" (If dtype is not specified, it defaults to float32). "노드를 만들 때 node3 = node1 + node2" (When creating nodes, node3 = node1 + node2).

In [5]: `print("node1:", node1, "node2:", node2)`
`print("node3:", node3)`

Handwritten note: "2개 출력" (2 outputs).

node1: Tensor("Const_1:0", shape=(), dtype=float32) node2: Tensor("Const_2:0", shape=(), dtype=float32)
node3: Tensor("Add:0", shape=(), dtype=float32)

In [6]: `sess = tf.Session()`
`print("sess.run(node1, node2): ", sess.run([node1, node2]))`
`print("sess.run(node3): ", sess.run(node3))`

Handwritten notes: "2개의 노드를 실행" (Execute 2 nodes). "run() ↑" (run() ↑). "한번 실행시키고 싶은 노드" (Node I want to run once).

sess.run(node1, node2): [3.0, 4.0]
sess.run(node3): 7.0

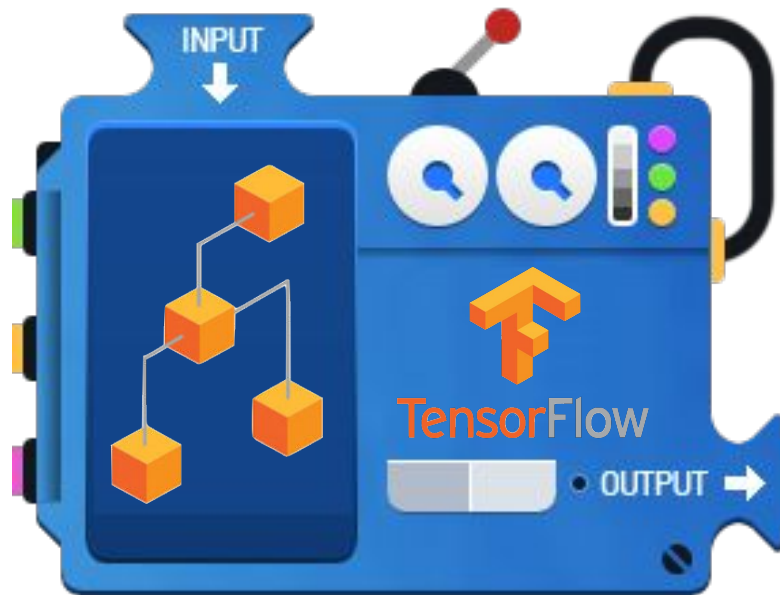
TensorFlow Mechanics

- 2 feed data and run graph (operation)
`sess.run (op)`

↳ graph를 실행

- 1 Build graph using
TensorFlow operations

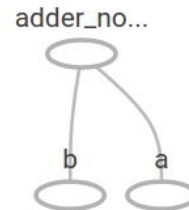
↳ ① tensorflow
tool로 graph
를 만들 Build



2 결과를 graph 속의 값이
update 되거나 어떤 값은 return
해줄

- 3 update variables
in the graph
(and return values)

Computational Graph



(1) Build graph (tensors) using TensorFlow operations

```
In [4]: node1 = tf.constant(3.0, tf.float32)
node2 = tf.constant(4.0) # also tf.float32 implicitly
node3 = tf.add(node1, node2)
```

(2) feed data and run graph (operation)

sess.run(op)

→ 세션 안들고
→ run()은 이동을 위한 operation(Node)
를 넘겨서 실행시킨다.

(3) update variables in the graph (and return values)

→ 값을 update
→ 값을 return

```
In [6]: sess = tf.Session()
print("sess.run(node1, node2): ", sess.run([node1, node2]))
print("sess.run(node3): ", sess.run(node3))
```

```
sess.run(node1, node2): [3.0, 4.0]
sess.run(node3): 7.0
```

Placeholder

placeholder에서 값을 넘겨주게 하는 것

type은 지정

placeholder의 값



그래프는 미리 만들어 놓은 실행시작은 런타임에서 값을 넘겨준다.

placeholder라는 특별한 node로 node를 만들어 준다.

In [7]:

```
a = tf.placeholder(tf.float32)
b = tf.placeholder(tf.float32)
add_node = a + b # + provides a shortcut for tf.add(a, b)
print(sess.run(add_node, feed_dict={a: 3, b: 4.5}))
print(sess.run(add_node, feed_dict={a: [1, 3], b: [2, 4]}))
```

```
7.5
[ 3.  7.]
```

array로 넘겨준다

placeholder는 둘레 값을 넘겨주게 함 (호환적)

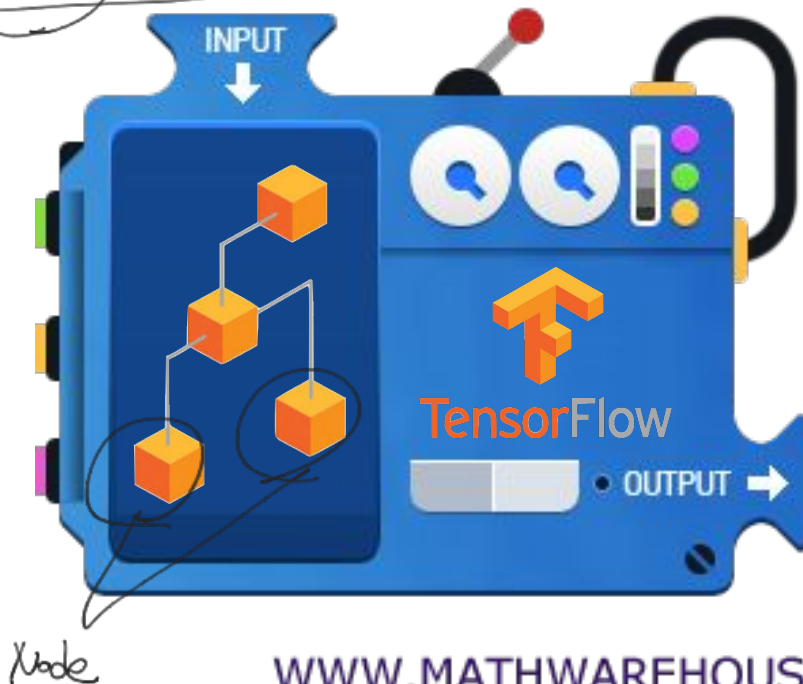
TensorFlow Mechanics

- 2 feed data and run graph (operation)
`sess.run(op, feed_dict={x: x_data})`

feed_dict
3개의 input

- 1 Build graph using
TensorFlow operations

graph을 graph
placeholder & node를
정의하는



- 3 update variables
in the graph
(and return values)

Everything is Tensor

Tensors

```
In [3]: 3 # a rank 0 tensor; this is a scalar with shape []  
        [1., 2., 3.] # a rank 1 tensor; this is a vector with shape [3]  
        [[1., 2., 3.], [4., 5., 6.]] # a rank 2 tensor; a matrix with shape [2, 3]  
        [[[1., 2., 3.]], [[7., 8., 9.]]] # a rank 3 tensor with shape [2, 1, 3]
```

```
Out[3]: [[[[1.0, 2.0, 3.0]], [[7.0, 8.0, 9.0]]]]
```

```
t = tf.Constant([1., 2., 3.])
```

Tensor Ranks, Shapes, and Types

```
t = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]
```

Rank	Math entity	Python example
0	<u>Scalar (magnitude only)</u>	<u>s = 483</u>
1	<u>Vector (magnitude and direction)</u>	v = [1.1, 2.2, 3.3]
2	<u>Matrix (table of numbers)</u>	m = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]
3	<u>3-Tensor (cube of numbers)</u>	t = [[[2], [4], [6]], [[8], [10], [12]], [[14], [16], [18]]]
n	<u>n-Tensor (you get the idea)</u>

Tensor Ranks, Shapes, and Types

$t = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]$

↪ 안쪽의 요소 3개

Rank	Shape	Dimension number	Example
0	$[]$ $[]$ 로 표현	0-D	A 0-D tensor. A scalar.
1	$[D_0]$	1-D	A 1-D tensor with shape $[5]$.
2	$[D_0, D_1]$ 2개 →	2-D	A 2-D tensor with shape $[3, 4]$.
3	$[D_0, D_1, D_2]$ 3개 →	3-D	A 3-D tensor with shape $[1, 4, 3]$.
n	$[D_0, D_1, \dots, D_{n-1}]$	n-D	A tensor with shape $[D_0, D_1, \dots, D_{n-1}]$.

Tensor Ranks, Shapes, and Types

```
t = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]
```

대부분의 경우

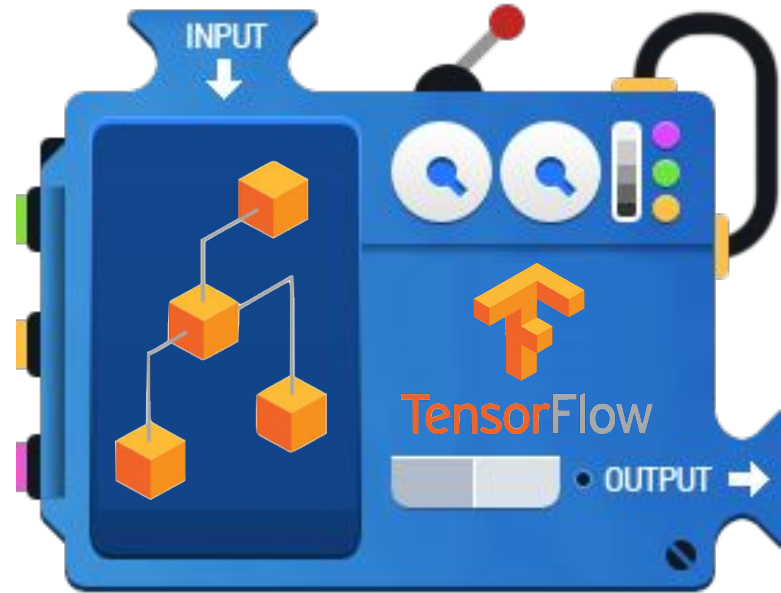
Data type	Python type	Description
<u>DT_FLOAT</u>	<u>tf.float32</u>	32 bits floating point.
DT_DOUBLE	<u>tf.float64</u>	64 bits floating point.
DT_INT8	tf.int8	8 bits signed integer.
DT_INT16	tf.int16	16 bits signed integer.
DT_INT32	<u>tf.int32</u>	32 bits signed integer.
DT_INT64	tf.int64	64 bits signed integer.

...

TensorFlow Mechanics

2 feed data and run graph (operation)
`sess.run (op, feed_dict={x: x_data})`

1 Build graph using
TensorFlow operations

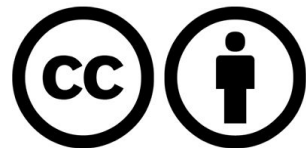


3 update variables
in the graph
(and return values)

Lab 2

Linear Regression

Sung Kim <hunkim+ml@gmail.com>





Variables

```
# Create two variables.
weights = tf.Variable(tf.random_normal([784, 200], stddev=0.35),
                      name="weights")
biases = tf.Variable(tf.zeros([200]), name="biases")
...
# Add an op to initialize the variables.
init_op = tf.global_variables_initializer()

# Later, when launching the model
with tf.Session() as sess:
    # Run the init operation.
    sess.run(init_op)
    ...
    # Use the model
    ...
```