## TensorFlow intro

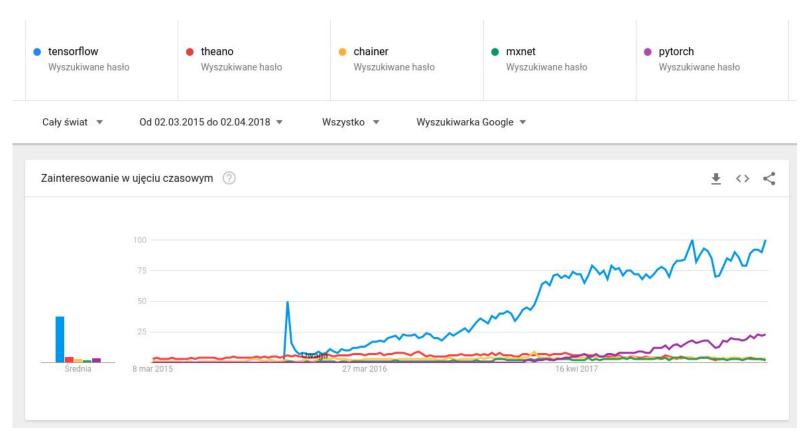


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#### What is Tensorflow

- TensorFlow<sup>™</sup> is an open source software library for high performance numerical computation (particularly GPUs)
- Strong support for machine learning and deep learning(like gradient computations)

# The most popular ML framework today





















#### Hello World

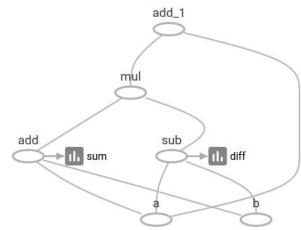
```
import tensorflow as tf
sess = tf.InteractiveSession()
a = tf.constant(5.0)
b = tf.constant(6.0)
c = a * b
print(c.eval())
```

#### Hello World

```
import tensorflow as tf
sess = tf.InteractiveSession() # What is this?
a = tf.constant(5.0)
b = tf.constant(6.0)
c = a * b
print(c.eval()) # Why do we have to do this, can't we just print(c),
to get result?
30.0
```

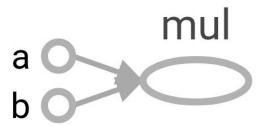
## Computation graphs

- "TensorFlow programs are usually structured into a construction phase, that assembles a graph, and an execution phase that uses a session to execute ops in the graph." -TensorFlow docs
- A computation graph is a DAG. It has "mathematical" operations as nodes(also called ops)
  - and tensor(multi-dimensional array as edges)
- Tensors are flowing through the computation graph, hence the name TensorFlow
- pip search tensorflow tensorflow (1.7.0) - TensorFlow helps the tensors flow



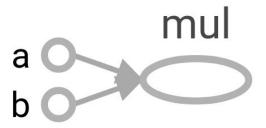
#### Hello World revisited

```
import tensorflow as tf
# Creating the computation graph phase
a = tf.constant(5.0, name='a')
b = tf.constant(6.0, name='b')
# type(a) == <class 'tensorflow.python.framework.ops.Tensor'>
# We are just creating a node in the computation graph here, no 5.0 * 6.0,
m = a * b # same as m = tf.multiply(a, b)
# type(m) == <class 'tensorflow.python.framework.ops.Tensor'>
```



#### Hello World revisited

```
import tensorflow as tf
# Creating the computation graph phase
a = tf.constant(5.0, name='a')
b = tf.constant(6.0, name='b')
# We are just creating a node in the computation graph here, no 5.0 * 6.0
m = a * b
# Running the computation graph phase
with tf.Session() as sess:
    print(m.eval()) # result = 30.0
    print(a.eval()) # result = 5.0
    print(sess.run([m, a])) # result = [30.0, 5.0]
```



#### Hello World 2.0

```
import tensorflow as tf
sess = tf.InteractiveSession()
a = tf.placeholder(dtype=tf.float32, name='a')
b = tf.placeholder(dtype=tf.float32, name='b')
c = a * b
feed_dict = {a: 5.0, b: 6.0}
print(sess.run(c, feed_dict))
```

- Use tf.placeholder variables (dummy nodes that provide entry points for data to computational graph).
- A feed\_dict is a python dictionary mapping from tf. placeholder vars (or their names) to data (numpy arrays, lists, etc.).

# What is wrong?

```
import tensorflow as tf
sess = tf.InteractiveSession()
a = tf.placeholder(dtype=tf.float32, name='a')
b = tf.placeholder(dtype=tf.float32, name='b')
X = [0.0, 1.0, 5.0, 5.0]
Y = [2.0, 3.0, 1.0, 8.0]
for x, y in zip(X, Y):
    feed dict = {a: x, b: y}
    print(sess.run(a + b, feed dict))
```

#### tf.Variable

```
import tensorflow as tf
sess = tf.InteractiveSession()
a = tf.Variable(0.0, dtype=tf.float32, name='a')
value to assign = tf.placeholder(dtype=tf.float32)
assign_op = a.assign(value_to_assign)
# print a.eval() # this would cause error because variable a is not initialized yet.
init_op = tf.global_variables_initializer()
sess.run(init op)
print(a.eval()) # result = 0.0
sess.run(assign_op, feed_dict={value_to_assign: 3.0})
print(a.eval()) # result = 3.0
```

#### tf.Variable

```
import tensorflow as tf
sess = tf.InteractiveSession()
b = tf.Variable(tf.random_uniform([2, 2], -0.1, 0.1), dtype=tf.float32, name='b')
c = tf.Variable(tf.random_normal([2, 2], stddev=0.1), dtype=tf.float32, name='c')
init_op = tf.global_variables_initializer()
sess.run(init_op)
print(b.eval())
print(c.eval())
[[-0.01472919 0.08135603]
 [ 0.07122798  0.03257368]]
[[0.03300044 0.1379751 ]
 [0.08122743 0.06315189]]
```

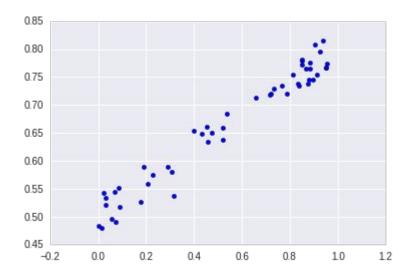
## tf.gradients

```
import tensorflow as tf
x = tf.placeholder(dtype=tf.float32)
f_x = x * x
df_dx = tf.gradients(xs=x, ys=f_x)

with tf.Session() as sess:
    feed_dict = {x: 3.0}
    print(sess.run([x, f_x, df_dx], feed_dict))

# Result: [array(3., dtype=float32), 9.0, [6.0]]
```

```
import numpy as np
import seaborn
import matplotlib.pyplot as plt
import random
f = lambda x: 0.3 * x + 0.5 # ground truth
g = lambda x: f(x) + random.gauss(0, 0.02) # a noisy version of f
X_data = [random.random() for _ in xrange(50)]
y data = map(g, X data)
plt.scatter(X data, y data)
X \text{ data} = \text{np.reshape}(X \text{ data}, (50, 1))
y data = np.reshape(y data, (50, 1))
```



```
# Creating placeholders for input
X = tf.placeholder(shape=(None, 1), dtype=tf.float32)
y_gt = tf.placeholder(shape=(None, 1), dtype=tf.float32)
# Creating model
W = tf.Variable(np.zeros((1, 1), dtype='float32'), name='W')
b = tf.Variable(0.0, name='b')
y_pred = tf.matmul(X, W) + b
loss = tf.reduce_mean((y_gt - y_pred) ** 2)
```

```
vars_to_train = [W, b]
gradients = tf.gradients(xs=vars_to_train, ys=loss)
assign_ops = []
for var, grad in zip(vars_to_train, gradients):
    assign_ops.append(var.assign(var - grad * 0.1))
gradient_step = assign_ops
```

```
vars_to_train = [W, b]
gradients = tf.gradients(xs=vars_to_train, ys=loss)
assign ops = []
for var, grad in zip(vars_to_train, gradients):
    assign ops.append(var.assign(var - grad * 0.1))
gradient step = assign ops
# OR using built-in GradientDescentOptimizer
optimizer = tf.train.GradientDescentOptimizer(learning rate=0.1)
gradient_step = optimizer.minimize(loss)
```

```
with tf.Session() as sess:
    sess.run(tf.global_variables_initializer())
    for epoch_idx in range(1000):
        feed_dict = {X: X_data, y_gt: y_data}
        _, loss_value = sess.run([gradient_step, loss], feed_dict)
        #print epoch_idx, loss_value
    print(sess.run(W), sess.run(b))
[[ 0.28261772]] 0.509359
```

#### **Tensorboard**

- Track progress of learning
- Compare experiments
- Save images/audio clips



# Questions?

# TensorFlow Lab



#### Labs

Work through the tutorials at your own pace:

https://www.tensorflow.org/versions/r1.4/get started/get started (don't read part about tf.estimator)
We are using tf version 1.4 because latter versions expose more high-level API, which hide how things actually work.

https://www.tensorflow.org/versions/r1.4/get\_started/mnist/beginners

Don't just read the code, try to run it simultaneously in the notebook/editor as you go through the text.

There are two task you can try solving:

- exc1.ipynb some introductory exercises
- mnist\_tf.py train your model on MNIST dataset

It is best to work in virtualenv.

To use jupyter with a virtualenv just install it in this virtualenv: 'pip install jupyter', and run 'jupyter notebook' having the virtualenv already activated.

virtualenvwrapper (<a href="https://virtualenvwrapper.readthedocs.io/en/latest">https://virtualenvwrapper.readthedocs.io/en/latest</a>) is very convenient for managing virtualenvs.

Exercises work in python3.