

Intro

- In this video we will overview basic principles of TensorFlow

TensorFlow DL framework

- We will use it in Jupyter Notebook with Python 3 kernel

```
import numpy as np  
import tensorflow as tf
```

- We will overview Python API for TensorFlow 1.2+
- APIs in other languages exist: Java, C++, Go
 - Python API is at present the most complete and the easiest to use
 - https://www.tensorflow.org/api_docs/



What is TensorFlow?

1. A tool to describe computational graphs
 - The foundation of computation in TensorFlow is the **Graph** object. This holds a network of nodes, each representing one **operation**, connected to each other as inputs and outputs.
2. A runtime for execution of these graphs
 - On CPU, GPU, TPU, ...
 - On one node or in distributed mode

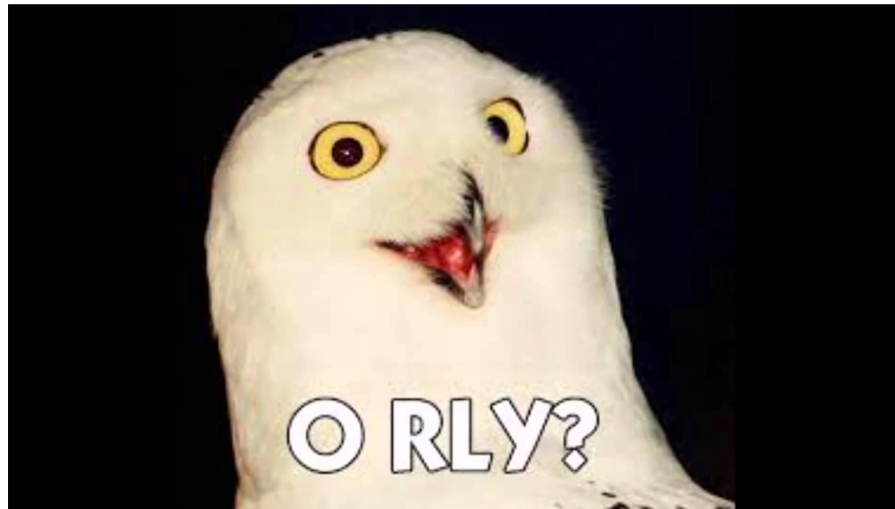


Why this name

- **Input** to any operation will be a collection of **tensors** (multi-dimensional arrays)
- **Output** will be a collection of tensors as well.
- We will have a graph of operations, each of which transforms tensors into another tensors, so it's a kind of a flow of tensors

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How the input looks like

- **Placeholder**

- This is placeholder for a tensor, which will be fed during graph execution (e.g. input features)
- `x = tf.placeholder(tf.float32, (None, 10))`

- **Variable**

- This is a tensor with some value that is updated during execution (e.g. weights matrix in MLP)
- `w = tf.get_variable("w", shape=(10, 20), dtype=tf.float32)`
- `w = tf.Variable(tf.random_uniform((10, 20)), name="w")`

- **Constant**

- This is a tensor with constant value, that cannot be changed
- `c = tf.constant(np.ones((4, 4)))`

Operation example

- Matrix product:

```
x = tf.placeholder(tf.float32, (None, 10))  
w = tf.Variable(tf.random_uniform((10, 20)), name="w")  
z = x @ w  
# z = tf.matmul(x, w)  
print(z)
```

- Output:

Tensor("matmul:0", shape=(?, 20), dtype=float32)

- We don't do any computations here, we just **define** the graph!

Computational graph

- TensorFlow creates a **default graph** after importing
 - All the operations will go there by default
 - You can get it with `tf.get_default_graph()`, which returns an instance of `tf.Graph`.
- You can **create your own** graph variable and define operations there:

```
g = tf.Graph()  
with g.as_default():  
    pass
```

- You can **clear** the default graph like this:

```
tf.reset_default_graph()
```


Jupyter Notebook cells

- If you run this cell **3 times**:

```
x = tf.placeholder(tf.float32, (None, 10))
```

- This is what you get in your **default graph**:

- Using `tf.get_default_graph().get_operations()`

```
[<tf.Operation 'Placeholder' type=Placeholder>,  
 <tf.Operation 'Placeholder_1' type=Placeholder>,  
 <tf.Operation 'Placeholder_2' type=Placeholder>]
```

- **Your graph is cluttered!**

- Clear your graph with `tf.reset_default_graph()`

Operations and tensors

- Every **node** in our graph is an **operation**:

```
x = tf.placeholder(tf.float32, (None, 10), name="x")
```

- Listing nodes with `tf.get_default_graph().get_operations()`:

```
[<tf.Operation 'x' type=Placeholder>]
```

- How to get **output tensors** of operation:

- `tf.get_default_graph().get_operations()[0].outputs`
- Output: `[<tf.Tensor 'x:0' shape=(?, 10) dtype=float32>]`

Running a graph

- A **tf.Session** object encapsulates the environment in which `tf.Operation` objects are executed, and `tf.Tensor` objects are evaluated.
- Create a session: `s = tf.InteractiveSession()`
- Defining a graph:
`a = tf.constant(5.0)`
`b = tf.constant(6.0)`
`c = a * b`
- Running a graph: `print(c)` *# here just looking at the type*
`print(s.run(c))` *# that's how you run the graph*
- Output:
`Tensor("mul:0", shape=(), dtype=float32)`
`30.0`

Running a graph

- Operations are written in C++ and executed on CPU or GPU.
- `tf.Session` owns necessary resources to execute your graph, such as `tf.Variable`, that occupy RAM.
- It is important to release these resources when they are no longer required with `tf.Session.close()`

Initialization of variables

- A variable has an initial value:
 - Tensor: `tf.Variable(tf.random_uniform((10, 20)), name="w")`
 - Initializer: `tf.get_variable("w", shape=(10, 20), dtype=tf.float32)`
- You need to run some code to **compute that initial value** in graph execution environment
- This is done with a call in your session `s`:

```
s.run(tf.global_variables_initializer())
```
- Without it you will get “Attempting to use uninitialized value” errors

Example

- Definition:

```
tf.reset_default_graph()
a = tf.constant(np.ones((2, 2), dtype=np.float32))
b = tf.Variable(tf.ones((2, 2)))
c = a @ b
```

- Running attempt:

```
s = tf.InteractiveSession()
s.run(c)
```

Output: **“Attempting to use uninitialized value”** error

- Running properly:

```
s.run(tf.global_variables_initializer())
s.run(c)
```

Output: array([[2., 2.], [2., 2.]], dtype=float32)

Feeding placeholder values

- Definition:

```
tf.reset_default_graph()
a = tf.placeholder(np.float32, (2, 2))
b = tf.Variable(tf.ones((2, 2)))
c = a @ b
```

- Running attempt:

```
s = tf.InteractiveSession()
s.run(tf.global_variables_initializer())
s.run(c)
```

Output: **“You must feed a value for placeholder tensor”** error

- Running properly:

```
s.run(tf.global_variables_initializer())
s.run(c, feed_dict={a: np.ones((2, 2))})
```

Output: array([[2., 2.], [2., 2.]], dtype=float32)

Summary

- TensorFlow: defining and running computational graphs
- Nodes of a graph are operations, that convert a collection of tensors into another collection of tensors
- In Python API you **define** the graph, you **don't execute** it along the way
 - In 1.5+ the latter mode is supported: eager execution
- You create a **session** to execute your graph (fast C++ code on CPU or GPU)
- Session **owns** all the resources (tensors eat RAM)