## Introduction to TensorFlow

Glenn Bruns CSUMB

Many figures in this deck from Géron, Hands-on Machine Learning with Scikit-Learn and TensorFlow

# Learning outcomes

After this lecture you should be able to:

- 1. List some of the features of TensorFlow
- Explain the basic concept behind TensorFlow programs
- 3. List the main parts of a TensorFlow program
- Explain the details of a simple TensorFlow program

Hey, where's the deep learning?

Patience, grasshopper, we will get to deep learning with TensorFlow soon.

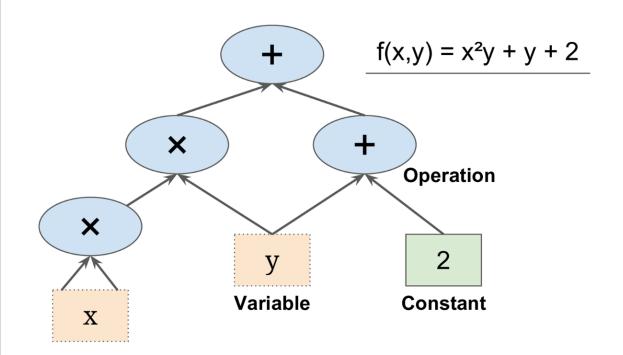
We're talking about TensorFlow first because it is a general-purpose library for numerical computing.

## What is TensorFlow?

It's an open source library for numerical computation.

## Principle:

- define a computation graph
- 2. execute the graph efficiently with C++ code



You can even break the graph into pieces and run them on separate CPUs or GPUs.

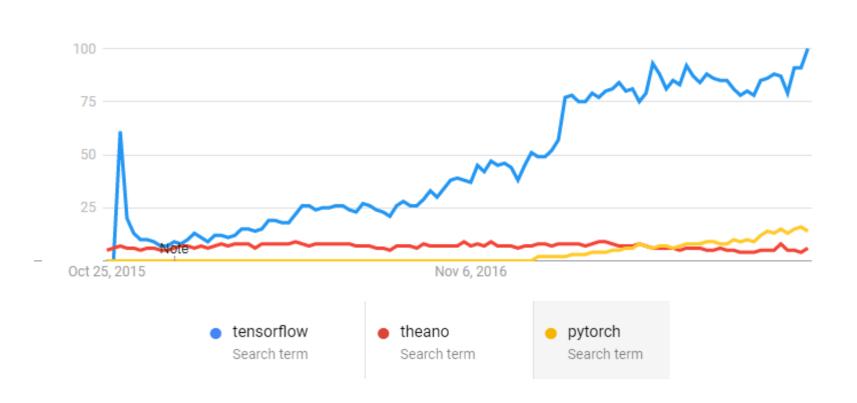
## Some features of TensorFlow

TensorFlow was released by Google in Nov. 2015

- □ Runs on Windows, macOS, and Linux
- ☐ Highly efficient C++ implementation
- Provides a very flexible Python API
- Comes with a visualization tool called
   TensorBoard
- □ Higher-level APIs have been built on top of it

# TensorFlow popularity

## Google Trends data:

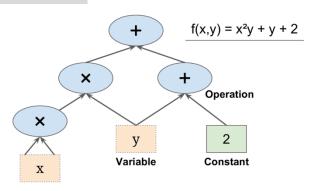


Some other deep learning libraries: Caffe, H2O, MXNet, Deeplearning4j

# A simple TensorFlow program

#### Construction phase: build a computation graph

```
x = tf.Variable(3, name="x")
y = tf.Variable(4, name="y")
f = x*x*y + y + 2
```



#### Execution phase: initialize and evaluate variables

```
sess = tf.Session()
sess.run(x.initializer)
sess.run(y.initializer)
result = sess.run(f)
print(result)
```

evaluation is done within a TensorFlow session

#### output:

42

# Alternatives for session handling

sess = tf.Session() sess.run(x.initializer) sess.run(y.initializer) 1 result = sess.run(f) print(result)

with tf.Session() as sess:

a pain to write sess.run a lot

x.initializer.run() 2 y.initializer.run()

result = f.eval()

sess is default session within block; session is automatically closed at end

init = tf.global variables initializer() with tf.Session() as sess: 3 init.run() result = f.eval()

initialize all variables at once; first statement is just a node

## Interactive sessions

#### third alternative from last slide

```
init = tf.global_variables_initializer()
with tf.Session() as sess:
    init.run()
    result = f.eval()
```

#### interactive session

```
sess = tf.InteractiveSession()
init.run()
result = f.eval()
print(result)
```

sess automatically becomes default session

handy in Jupyter notebooks or Spyder

need to close session manually

# Resetting a graph

When working in Jupyter notebooks or Spyder, you often rerun commands.

When nodes are created, they're added to the default graph automatically.

To avoid graphs with duplicate nodes, reset the graph:

```
tf.reset_default_graph()
```

Our text explains how to manage multiple graphs.

## Evaluating a node

```
w = tf.constant(3)
x = w + 2
y = x + 5
z = x * 3
with tf.Session() as sess:
    print(y.eval()) # 10
    print(z.eval()) # 15
```

### When z.eval() is run:

- the graph is examined for dependencies
- here, z depends on x, and x depends on w
- so w is evaluated, then x, then z

## Across graph runs:

- node values are dropped
- except variable values

# Linear Regression with TensorFlow

```
Where are the graph
import numpy as np
                                                constr. and execution
from sklearn.datasets import fetch california
                                                phases?
housing = fetch california housing()
                                                Is tf.transpose a TF or
m, n = housing.data.shape
                                                a scipy operation?
housing data plus bias = np.c [np.ones((m, 1)), housing.data|
                                                Where does variable
X = tf.constant(housing_data_plus_bias, dtype= X get its TF name?
y = tf.constant(housing.target.reshape(-1, 1), dtype=tf.float32,
name="y")
XT = tf.transpose(X)
theta = tf.matmul(tf.matmul(tf.matrix inverse(tf.matmul(XT, X)), XT), y)
with tf.Session() as sess:
    theta value = theta.eval()
```

We are computing the normal equation: this is not deep learning

# Summary

- □ TensorFlow is Google's general-purpose library for computing on graphs
- □ It is targeted at deep learning
- A graph can be run across multiple CPUs or GPUs
- □ In TensorFlow programs:
  - you first construct a graph
  - you then execute the graph

## Bonus content: what is a tensor?

- □ A tensor is an n-dimensional array it generalizes vectors and matrices
  - scalar = 0-dimensional array
  - vector = 1-dimensional array
  - matrix = 2-dimensional array
- □ In TensorFlow the number of dimensions is called the 'rank'
- □ The number of elements in each dimension is the 'shape'

See tensorflow.org/programmers\_guide/tensors