

Large-Scale Deep Learning With TensorFlow

Jeff Dean Google Brain team g.co/brain

In collaboration with **many** other people at Google

We can now store and perform computation on large datasets









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But what we really want is not just raw data, but computer systems that **understand** this data



What do I mean by understanding?

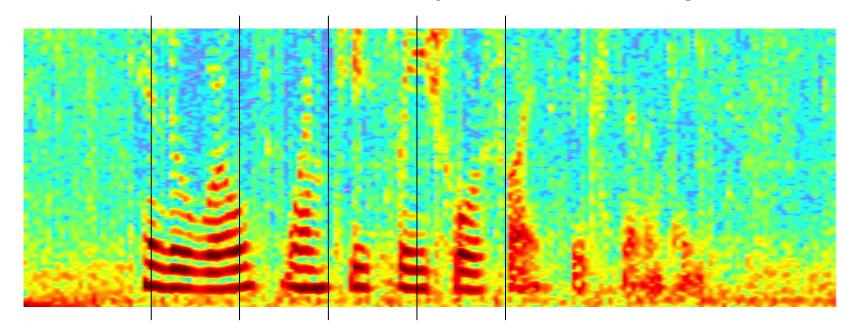




What do I mean by understanding?

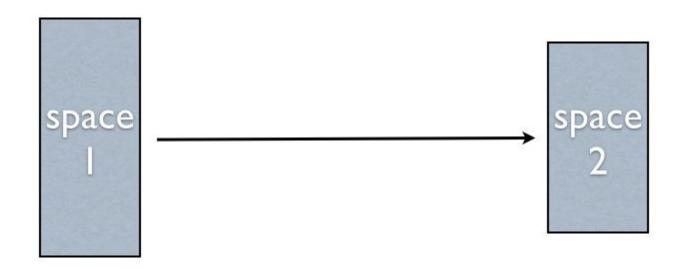


What do I mean by understanding?



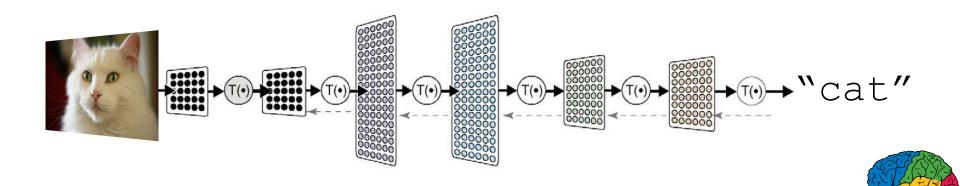
Neural Networks

Learn a complicated function from data

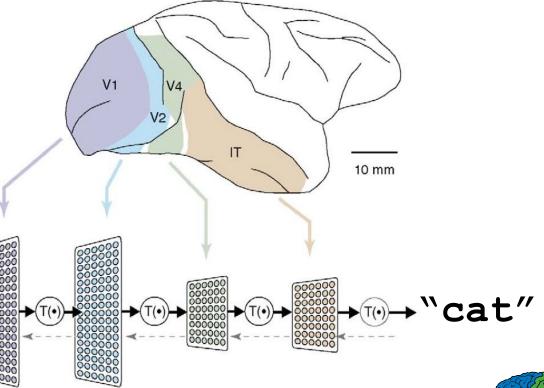




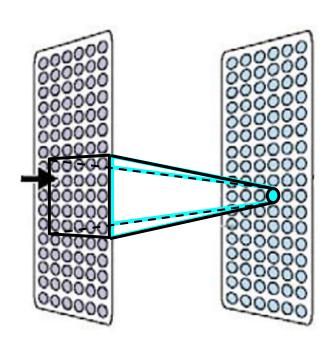
- A powerful class of machine learning model
- Modern reincarnation of artificial neural networks
- Collection of simple, trainable mathematical functions



 Loosely based on (what little) we know about the brain

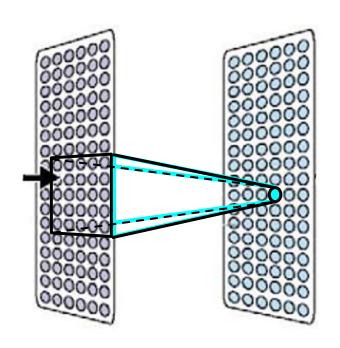






Commonalities with real brains:

- Each neuron is connected to a small subset of other neurons.
- Based on what it sees, it decides what it wants to say.
- Neurons learn to cooperate to accomplish the task.



Each neuron implements a relatively simple mathematical function.

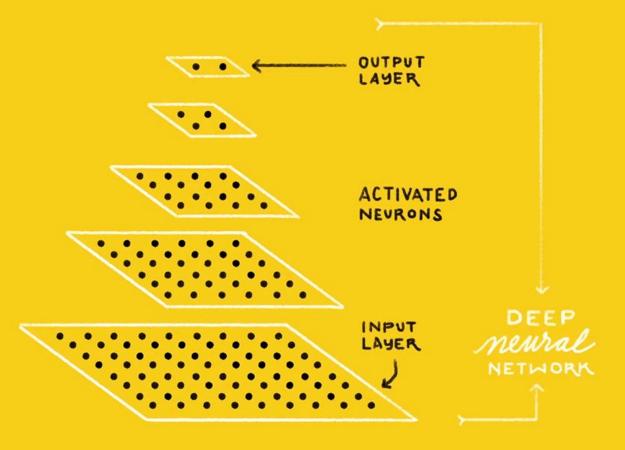
$$y = g(\vec{w} \cdot \vec{x} + b)$$

But the composition of 10⁶ - 10⁹ such functions is surprisingly powerful.

CAT DOG

CAT & DOG?





Important Property of Neural Networks

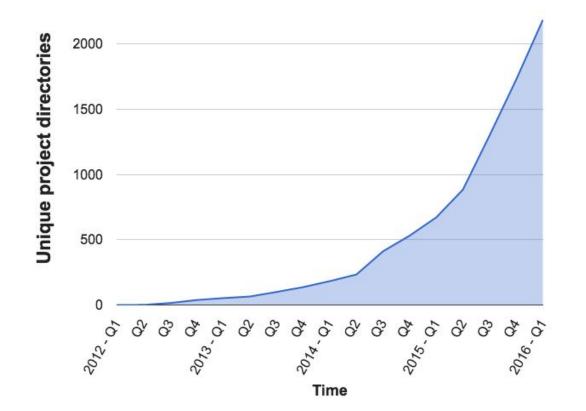
Results get better with

more data +
bigger models +
more computation

(Better algorithms, new insights and improved techniques always help, too!)

Growing Use of Deep Learning at Google

of directories containing model description files



Across many products/areas:

Android Apps drug discovery Gmail Image understanding Maps Natural language understanding Photos Robotics research Speech Translation YouTube ... many others ...





http://tensorflow.org/

and

https://github.com/tensorflow/tensorflow

Open, standard software for general machine learning

Great for Deep Learning in particular

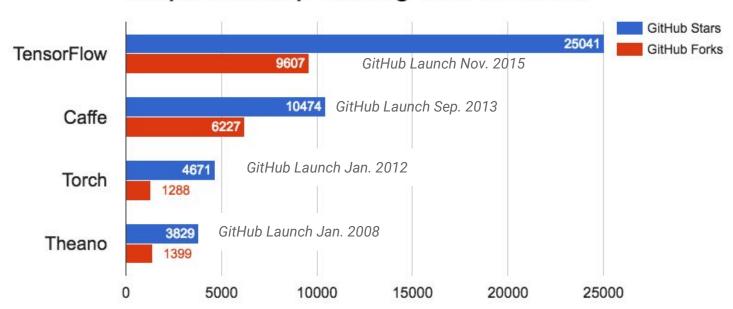
First released Nov 2015

Apache 2.0 license

Strong External Adoption



Adoption of Deep Learning Tools on GitHub



50,000+ binary installs in 72 hours, 500,000+ since Nov, 2015

Most forks of any GitHub repo in 2015, despite only being available starting in Nov, 2015

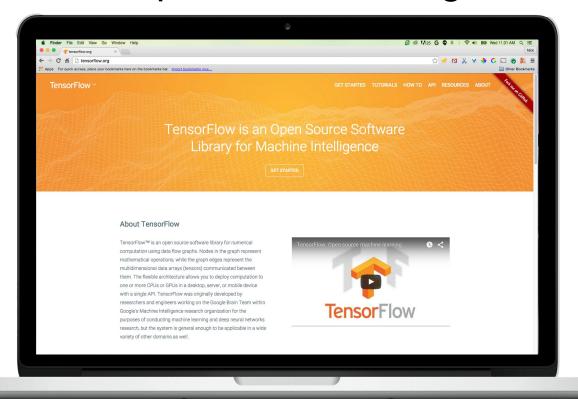
(source: http://donnemartin.com/viz/pages/2015)

Motivations

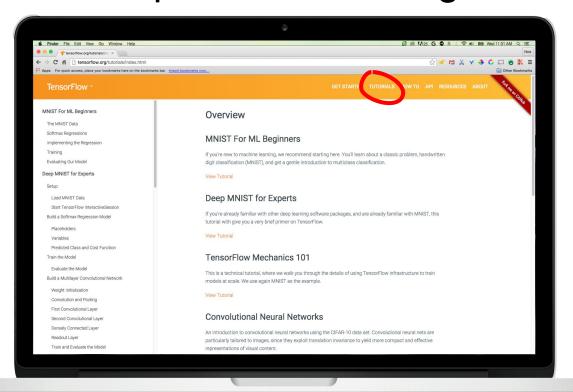


- DistBelief (our 1st system) was scalable and good for production deployment, but not as flexible as we wanted for research purposes
- Better understanding of problem space allowed us to make some dramatic simplifications
- Define a standard way of expressing machine learning ideas and computations
- Short circuit the MapReduce/Hadoop inefficiency

http://tensorflow.org/

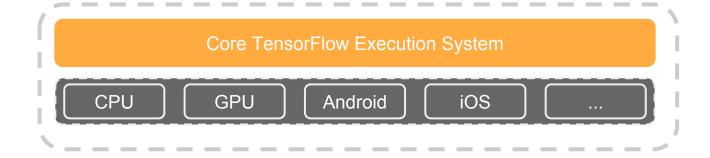


http://tensorflow.org/



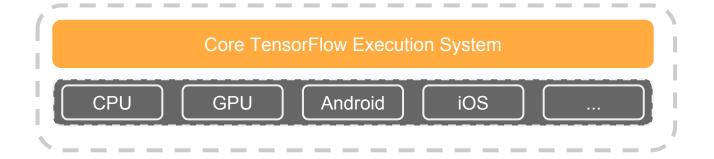
TensorFlow: Expressing High-Level ML Computations

- Core in C++
 - Very low overhead



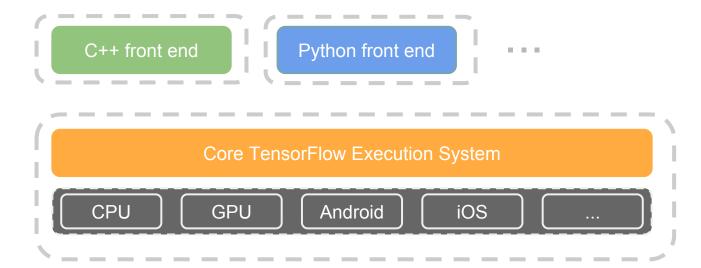
TensorFlow: Expressing High-Level ML Computations

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- Different front ends for specifying/driving the computation
 - Python and C++ today, easy to add more



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Example TensorFlow fragment

Build a graph computing a neural net inference.

```
import tensorflow as tf
from tensorflow.examples.tutorials.mnist import input_data

mnist = input_data.read_data_sets('MNIST_data', one_hot=True)
x = tf.placeholder("float", shape=[None, 784])
W = tf.Variable(tf.zeros([784,10]))
b = tf.Variable(tf.zeros([10]))
y = tf.nn.softmax(tf.matmul(x, W) + b)
```

Python API for Machine Learning

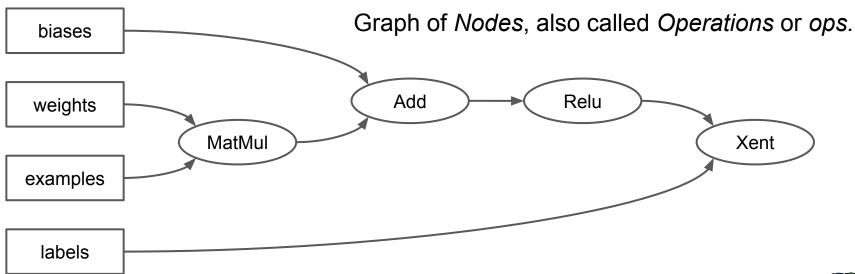
- Automatically add ops to calculate symbolic gradients of variables w.r.t. loss function.
- Apply these gradients with an optimization algorithm

```
y_ = tf.placeholder(tf.float32, [None, 10])
cross_entropy = -tf.reduce_sum(y_*tf.log(y))
opt = tf.train.GradientDescentOptimizer(0.01)
train_op = opt.minimize(cross_entropy)
```

Python API for Machine Learning

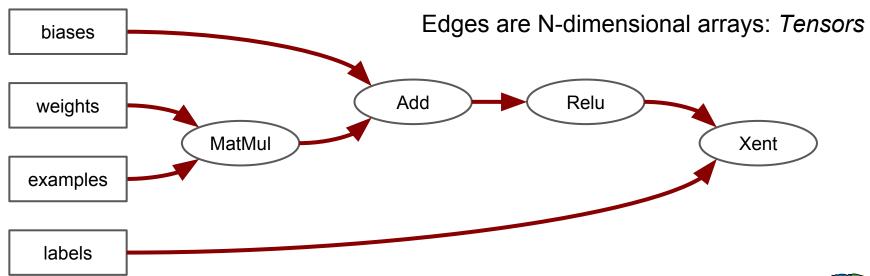
Launch the graph and run the training ops in a loop

```
init = tf.initialize_all_variables()
sess = tf.Session()
sess.run(init)
for i in range(1000):
   batch_xs, batch_ys = mnist.train.next_batch(100)
   sess.run(train_step, feed_dict={x: batch_xs, y_: batch_ys})
```



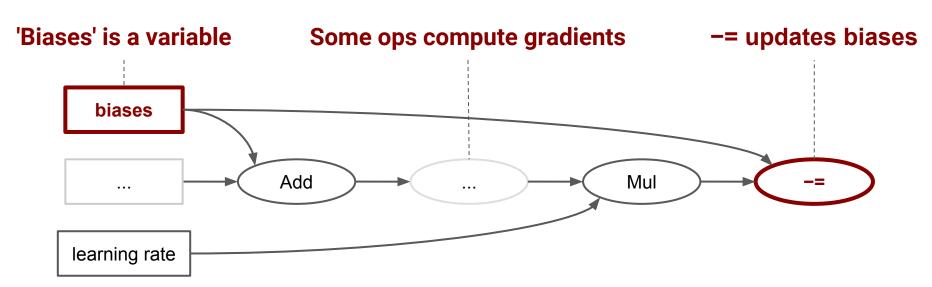






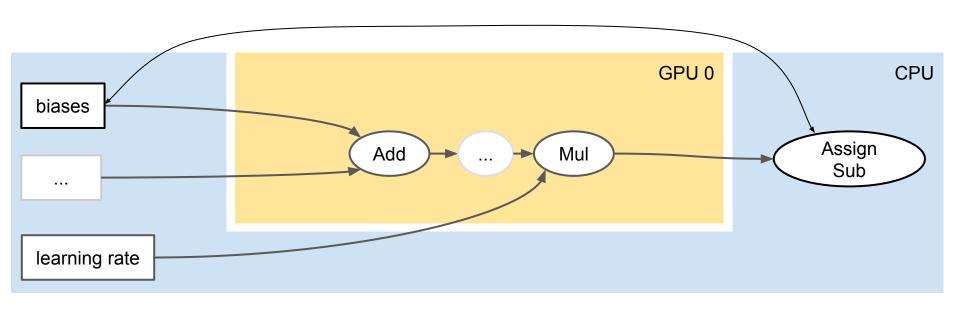






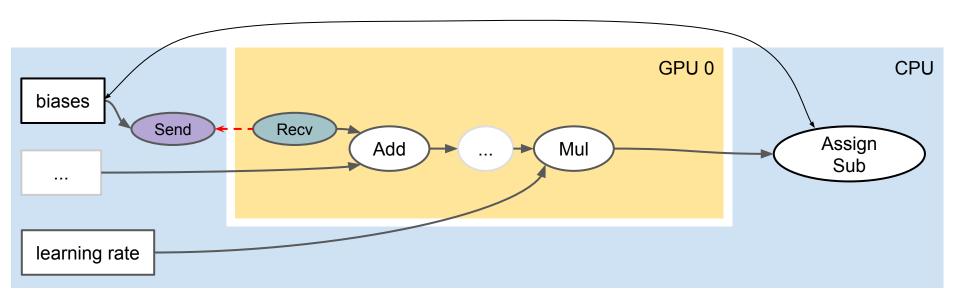






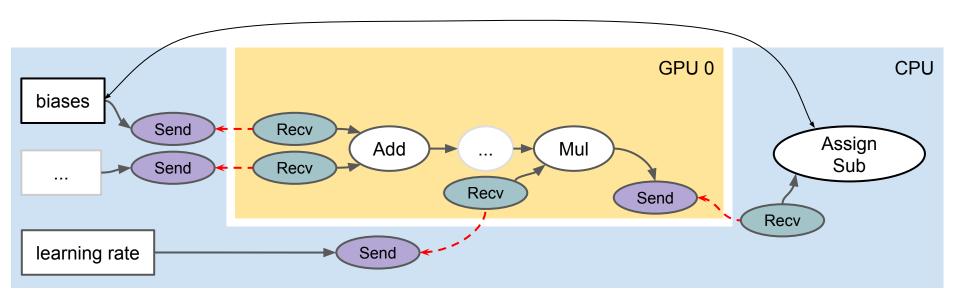
Assign Devices to Ops

- TensorFlow inserts Send/Recv Ops to transport tensors across devices
- Recv ops pull data from Send ops



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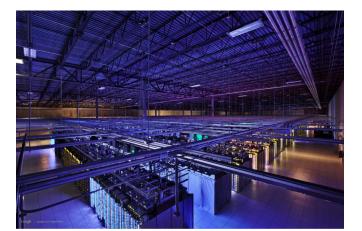


Automatically Runs on Variety of Platforms

phones



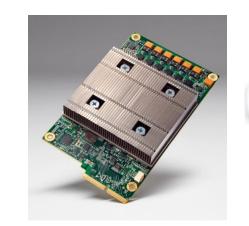
distributed systems of 100s of machines and/or GPU cards



single machines (CPU and/or GPUs) ...



custom ML hardware



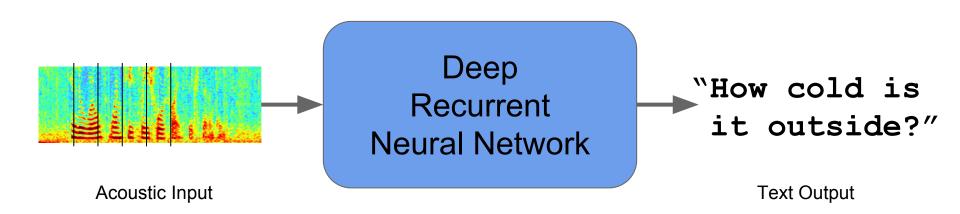




What are some ways that deep learning is having a significant impact at Google?



Speech Recognition

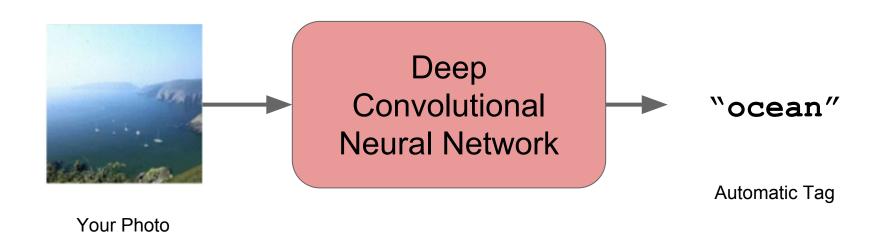


Reduced word errors by more than 30%



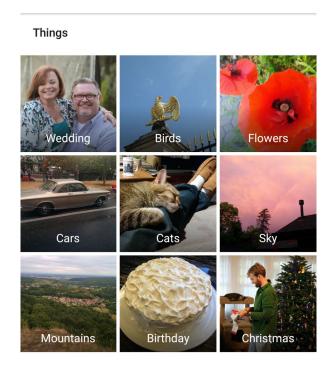


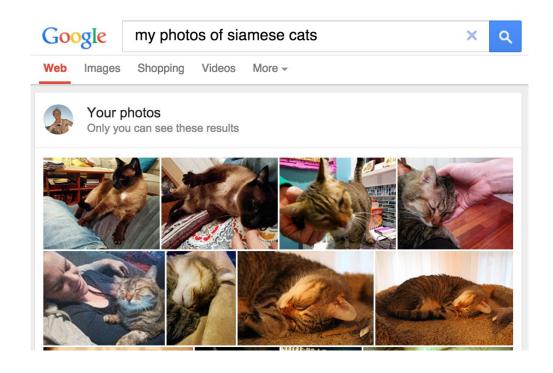
Google Photos Search



Search personal photos without tags.

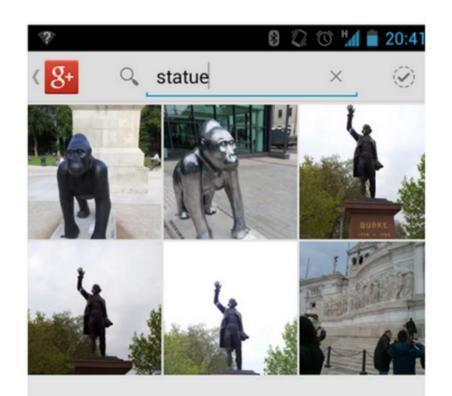
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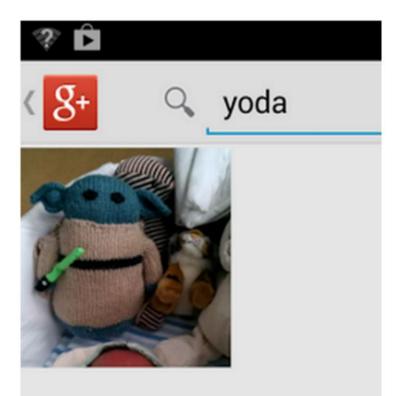




Google Photos Search

"Wow. The new Google photo search is a bit insane. I didn't tag those...:)"





1234 Bryant St, Palo Alto, CA 94301, USA







Analysis complete. Your roof has:



1,658 hours of usable sunlight per year Based on day-to-day analysis of weather patterns



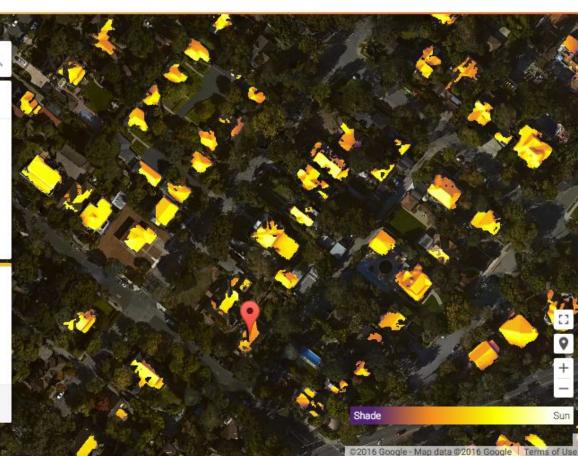
708 sq feet available for solar panels Based on 3D modeling of your roof and nearby trees

If your electric bill is at least \$175/month, leasing solar panels could reduce it.

FINE-TUNE ESTIMATE

SEE SOLAR PROVIDERS

Wrong roof? Drag the marker to the right one.

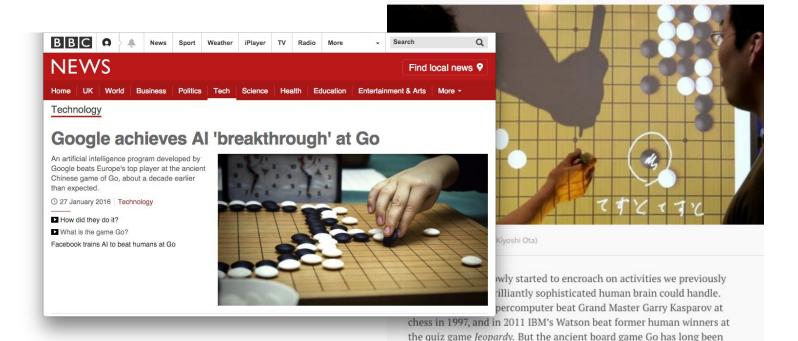


"Seeing" Go

Google's Al just cracked the game that supposedly no computer could beat

one of the major goals of artificial intelligence research. It's understood to be one of the most difficult games for computers to handle due to the sheer number of possible moves a player can make at any given point.

By Mike Murphy January 27, 2016



Until now, that is.



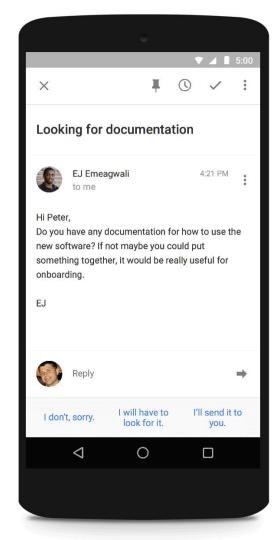


Smart Reply

April 1, 2009: April Fool's Day joke

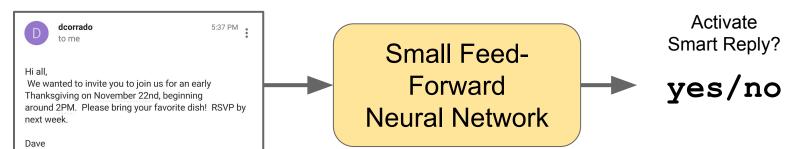
Nov 5, 2015: Launched Real Product

Feb 1, 2016: >10% of mobile Inbox replies



Google Research Blog - Nov 2015

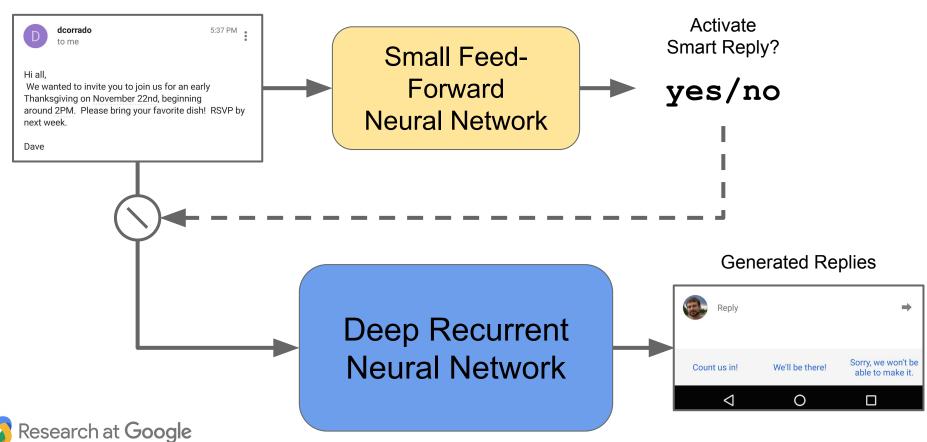
Incoming Email



Smart Reply

Google Research Blog - Nov 2015





Combined Vision + Translation

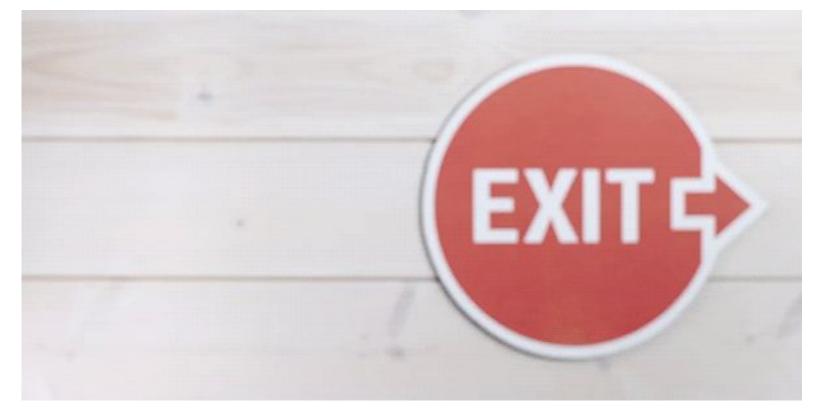


Image Captions Research



Human: A young girl asleep on the sofa cuddling a stuffed bear.

Model: A close up of a child holding a stuffed animal.

Model: A baby is asleep next to a teddy bear.



A man holding a tennis racquet on a tennis court.



A group of young people playing a game of Frisbee



Two pizzas sitting on top of a stove top oven



A man flying through the air while riding a snowboard



Experiment Turnaround Time and Research Productivity

• Minutes, Hours:

Interactive research! Instant gratification!

1-4 days

- Tolerable
- Interactivity replaced by running many experiments in parallel

1-4 weeks

- High value experiments only
- Progress stalls

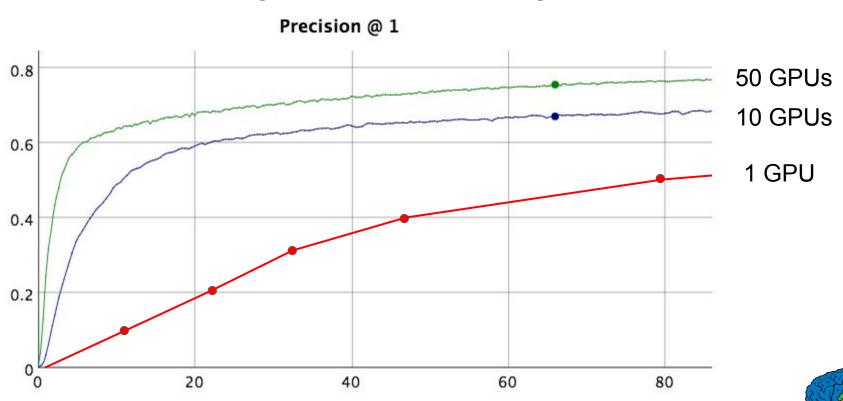
>1 month

Don't even try



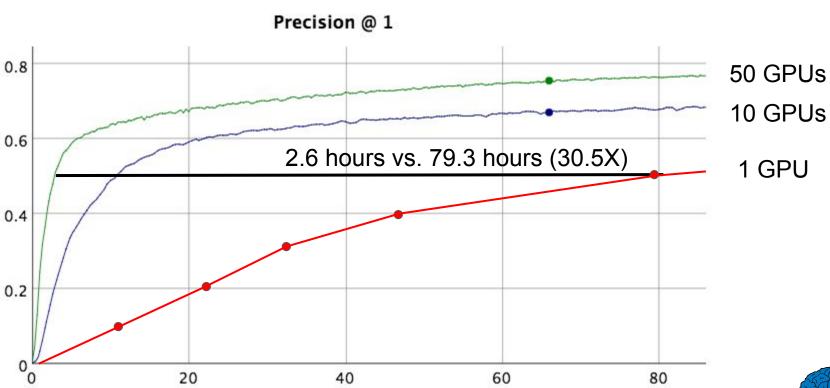


Image Model Training Time



Hours

Image Model Training Time



Hours



How Can You Get Started with Machine Learning?

Four ways, with varying complexity:

- (1) Use a Cloud-based API (Vision, Speech, etc.)
- (2) Run your own pretrained model
- (3) Use an existing model architecture, and retrain it or fine tune on your dataset
- (4) Develop your own machine learning models for new problems

More flexible, but more effort required



(1) Use Cloud-based APIs



GOOGLE TRANSLATE API

Dynamically translate between thousands of available language pairs

cloud.google.com/translate



CLOUD SPEECH API ALPHA

Speech to text conversion powered by machine learning

cloud.google.com/speech



CLOUD VISION API

Derive insight from images with our powerful Cloud Vision API

cloud.google.com/vision

CLOUD TEXT API ALPHA

Use Cloud Text API for sentiment analysis and entity recognition in a piece of text.

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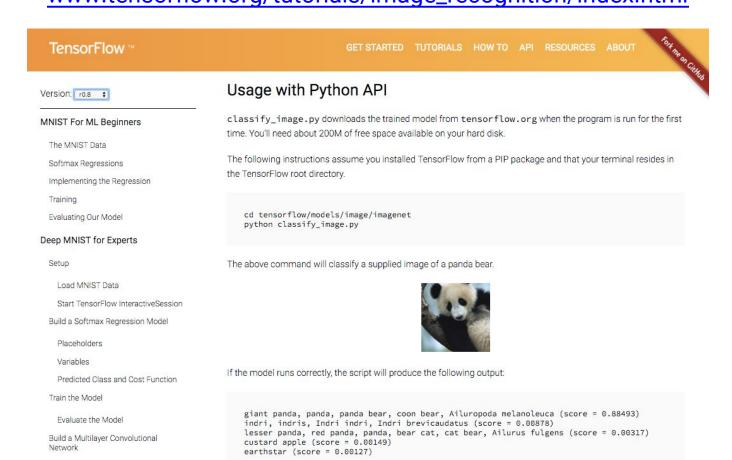
https://cloud.google.com/vision/



"running", "score": 0.99803412, "marathon", "score": 0.99482006 "joyLikelihood": "VERY_LIKELY"

,"description": "ABIERTO\n",
"local": "es"

(2) Using a Pre-trained Image Model yourself with TensorFlow www.tensorflow.org/tutorials/image_recognition/index.html



Woight Initialization

For training your own models (3 & 4), two choices:

Run open-source release on your own physical machines or virtual machines in a cloud hosting environment

or



cloud.google.com/ml

(3) Training a Model on Your Own Image Data

www.tensorflow.org/versions/master/how_tos/image_retraining/index.html

TensorFlow ™

GET STARTED TUTORIALS HOW TO API RESOURCES A

of the or Cithic

Version: master \$

Variables: Creation, Initialization, Saving, and Loading

Creation

Device placement

Initialization

Initialization from another Variable

Custom Initialization

Saving and Restoring

Checkpoint Files

Saving Variables

Restoring Variables

Choosing which Variables to Save and Restore

How to Retrain Inception's Final Layer for New Categories

Modern object recognition models have millions of parameters and can take weeks to fully train. Transfer learning is a technique that shortcuts a lot of this work by taking a fully-trained model for a set of categories like ImageNet, and retrains from the existing weights for new classes. In this example we'll be retraining the final layer from scratch, while leaving all the others untouched. For more information on the approach you can see this paper on Decaf.

Though it's not as good as a full training run, this is surprisingly effective for many applications, and can be run in as little as thirty minutes on a laptop, without requiring a GPU. This tutorial will show you how to run the example script on your own images, and will explain some of the options you have to help control the training process.

Contents

- How to Retrain Inception's Final Layer for New Categories
 - Training on Flowers

(4) Develop your own machine learning models

https://www.tensorflow.org/versions/master/get_started/basic_usage.html

TensorFlow ™

GET STARTED

Overview

TensorFlow is a programming system in which you represent computations as graphs. Nodes in the graph are called ops (short for operations). An op takes zero or more Tensors, performs some computation, and produces zero or more Tensors. A Tensor is a typed multi-dimensional array. For example, you can represent a mini-batch of images as a 4-D array of floating point numbers with dimensions [batch, height, width, channels].

A TensorFlow graph is a description of computations. To compute anything, a graph must be launched in a Session. A Session places the graph ops onto Devices, such as CPUs or GPUs, and provides methods to execute them. These methods return tensors produced by ops as numpy ndarray objects in Python, and as tensorflow::Tensor instances in C and C++.

The computation graph

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.

Example: Combining Vision with Robotics

"Deep Learning for Robots: Learning from Large-Scale Interaction", Google Research Blog, March, 2016

"Learning Hand-Eye Coordination for Robotic Grasping with Deep Learning and Large-Scale Data Collection", Sergey Levine, Peter Pastor, Alex Krizhevsky, & Deirdre Quillen, Arxiv, arxiv.org/abs/1603.02199



https://www.youtube.com/watch?v=iaF43Ze1oel

What Does the Future Hold?

Deep learning usage will continue to grow and accelerate:

- Across more and more fields and problems:
 - o robotics, self-driving vehicles, ...
 - health care
 - video understanding
 - dialogue systems
 - personal assistance
 - 0 ...



Conclusions

Deep neural networks are making significant strides in understanding:
In speech, vision, language, search, ...

If you're not considering how to use deep neural nets to solve your vision or understanding problems, you almost certainly should be



Further Reading

- Dean, et al., Large Scale Distributed Deep Networks, NIPS 2012, research.google. com/archive/large_deep_networks_nips2012.html.
- TensorFlow white paper, tensorflow.org/whitepaper2015.pdf (clickable links in bibliography)
- TensorFlow: A System for Large-Scale Machine Learning, http://arxiv.org/abs/1605.08695

tensorflow.org

research.google.com/people/jeff

We're Hiring! See g.co/brain



Questions?