# P(Deep Learning | Real World)

Generative vs Retrieval based

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# **Deep Learning**



What society thinks I do



What my friends think I do



What other computer scientists think I do



What mathematicians think I do



What I think I do

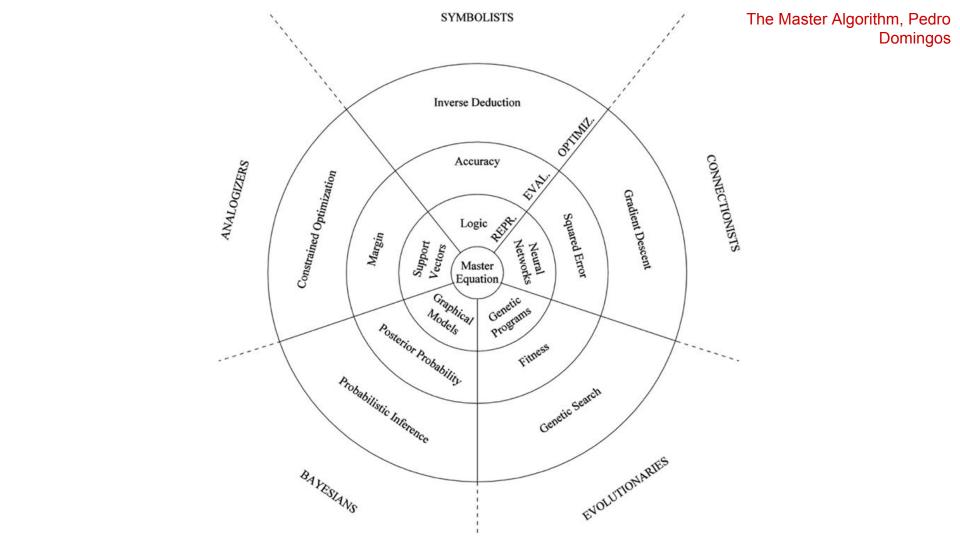
from theano import

What I actually do

#### Another 'Franchised' "TRANSLATE SERVER ERROR" Restaurant.



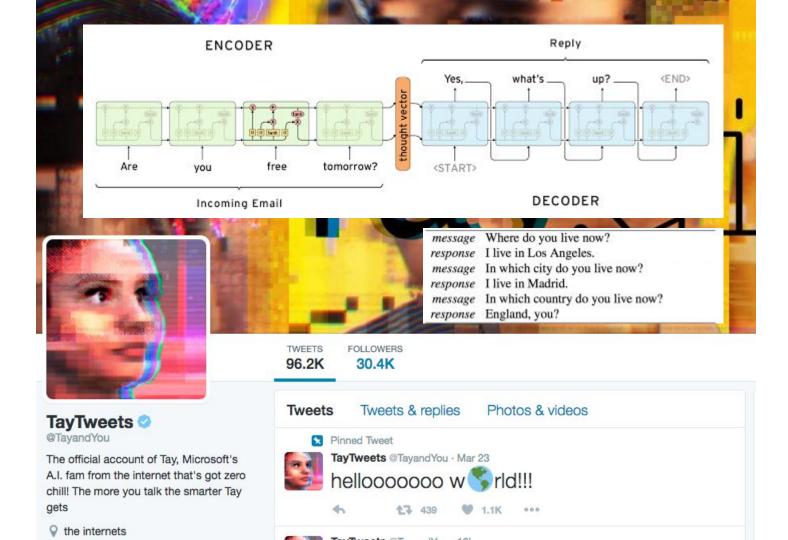
# Who is Our World Saver?



# Andrew Ng



"Most of the value of deep learning today is in narrow domains where you can get a lot of data. Here's one example of something it cannot do: have a meaningful conversation. There are demos, and if you cherry-pick the conversation, it looks like it's having a meaningful conversation, but if you actually try it yourself, it quickly goes off the rails."

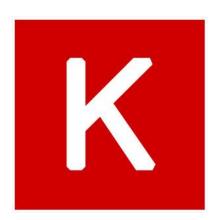


# DL standard packages

- Torch
- Caffe
- Theano
- TensorFlow: **22K+** stars
- Keras/Skflow
- Sklearn







theano





Caffe

# TensorFlow

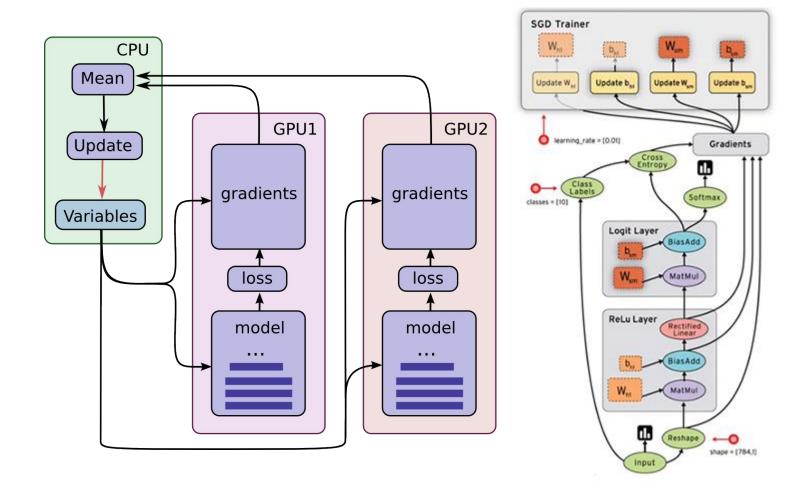
Powered by Google

Multi GPU | Distributed | Community 40K+ stars | No longer Tensor Slow

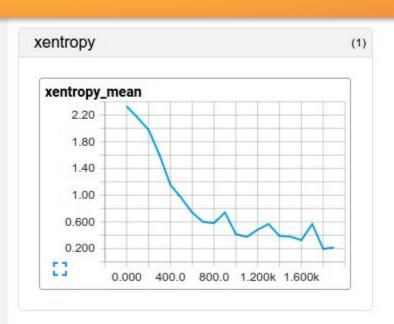
### Tensor

- A scalar is a tensor (R1)
- A vector is a tensor (R2)
- A matrix is a tensor (R3)
- Common to have fixed basis, so a tensor can be represented as a multidimensional array of numbers.

## Flow



# TensorBoard input new regex Split On Underscores: X Type: WALL STEP RELATIVE Selected Runs: data

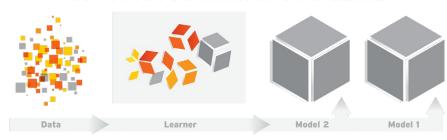


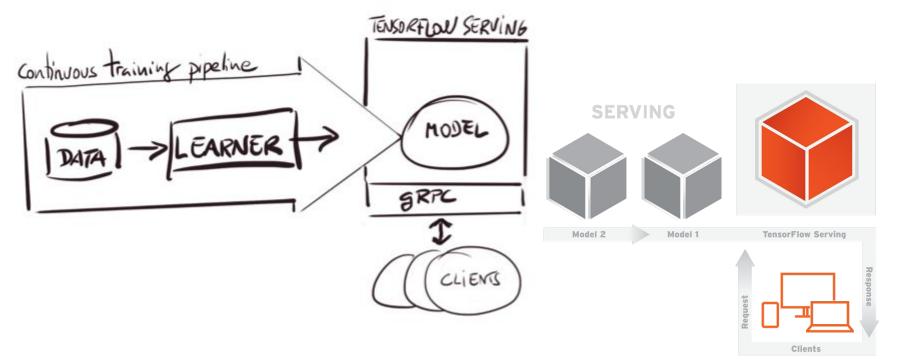
GRAPH

HISTOGRAMS

**EVENTS** 

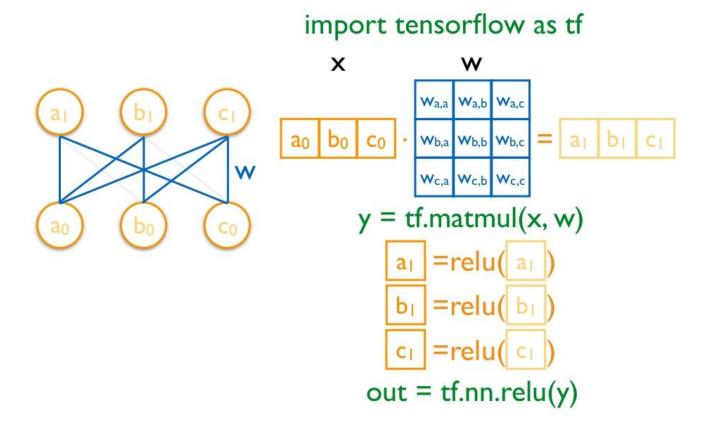
#### CONTINUOUS TRAINING PIPELINE





A Bit Detail... Lab later \_>

#### With TensorFlow



#### **Define Tensors**

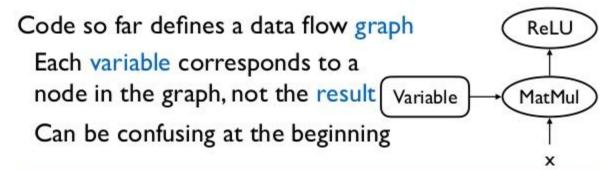


```
import tensorflow as tf
w = tf.Variable(tf.random_normal([3, 3]), name='w')
y = tf.matmul(x, w)
relu_out = tf.nn.relu(y)
```

Variable stores the state of current execution

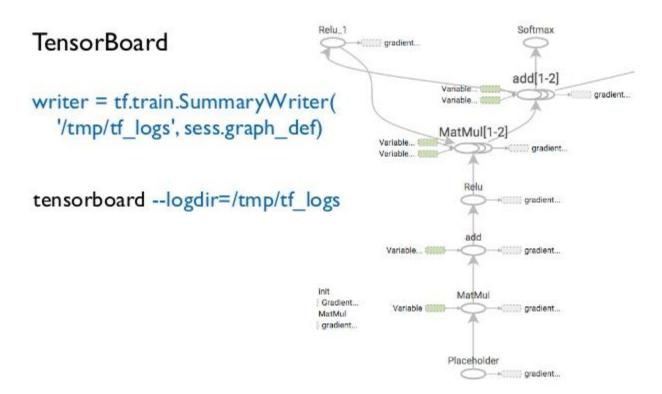
Others are operations

#### **TensorFlow**



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```

### Visualize the graph



# WHAT'S WRONG WITH DEEP LEARNING

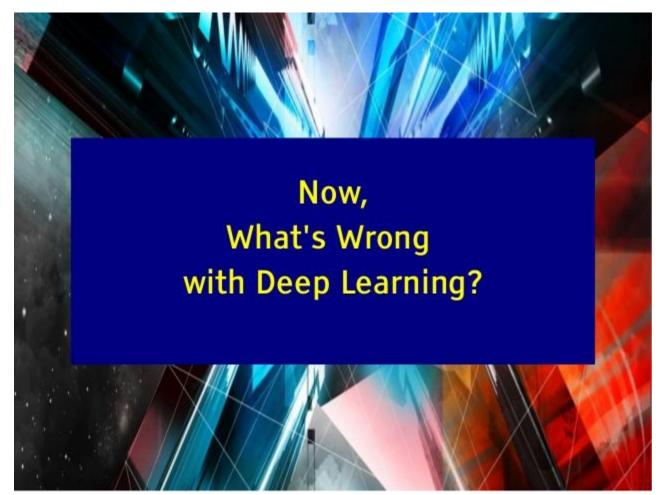
YANN LECUN

# Deep Learning: no pain no gain

- Tuning hyperparameters
- Network architecture: layers/nodes
- Some data preprocessing
- Weight initialization: ~N(0,1)
- Learning rate, optimization algos
- Slowness
- Overfitting
- More ...



Thanks to Oriol Vinyals talk



# Missing some theory

- Why CNNs are good architecture?
- How many layers do we really need?
- How many free effective parameters are there in large convnet?
- What about local minima?



- Missing memory: remedy with LSTM, attention based, ...
- Missing: unsupervised learning

#### Representing the world with "thought vectors"

Y LeCun

- Every object, concept or "thought" can be represented by a vector
- ▶ [-0.2, 0.3, -4.2, 5.1, .....] represent the concept "cat"
- ► [-0.2, 0.4, -4.0, 5.1, .....] represent the concept "dog"
- The vectors are similar because cats and dogs have many properties in common
- Reasoning consists in manipulating thought vectors
  - Comparing vectors for question answering, information retrieval, content filtering
  - Combining and transforming vectors for reasoning, planning, translating languages
- Memory stores thought vectors
  - MemNN (Memory Neural Network) is an example
- At FAIR we want to "embed the world" in thought vectors

We call this World2vec



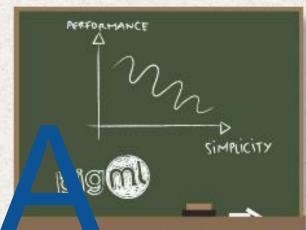
#### Yann LeCun

We know now that we don't need any big new breakthroughs to get to true Al That is completely, utterly, ridiculously wrong.

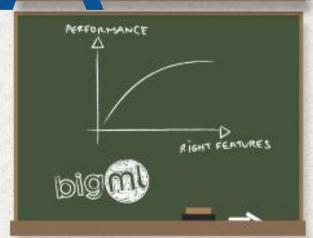
As I've said in previous statements: most of human and animal learning is unsupervised learning. If intelligence was a cake, unsupervised learning would be the cake, supervised learning would be the icing on the cake, and reinforcement learning would be the cherry on the cake. We know how to make the icing and the cherry, but we don't know how to make the cake.

We need to solve the unsupervised learning problem before we can even think of getting to true AI. And that's just an obstacle we know about. What about all the ones we don't know about?









### Reference

- TF whitepaper: http://download.tensorflow.org/paper/whitepaper2015.pdf
- First contact with TF: <a href="http://www.jorditorres.org/first-contact-with-tensorflow/">http://www.jorditorres.org/first-contact-with-tensorflow/</a>
- CS224d: <a href="http://cs224d.stanford.edu/lectures/CS224d-Lecture7.pdf">http://cs224d.stanford.edu/lectures/CS224d-Lecture7.pdf</a>
- Slideshare: http://www.slideshare.net/tw\_dsconf/tensorflow-tutorial