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TensorFlow for Java Developers

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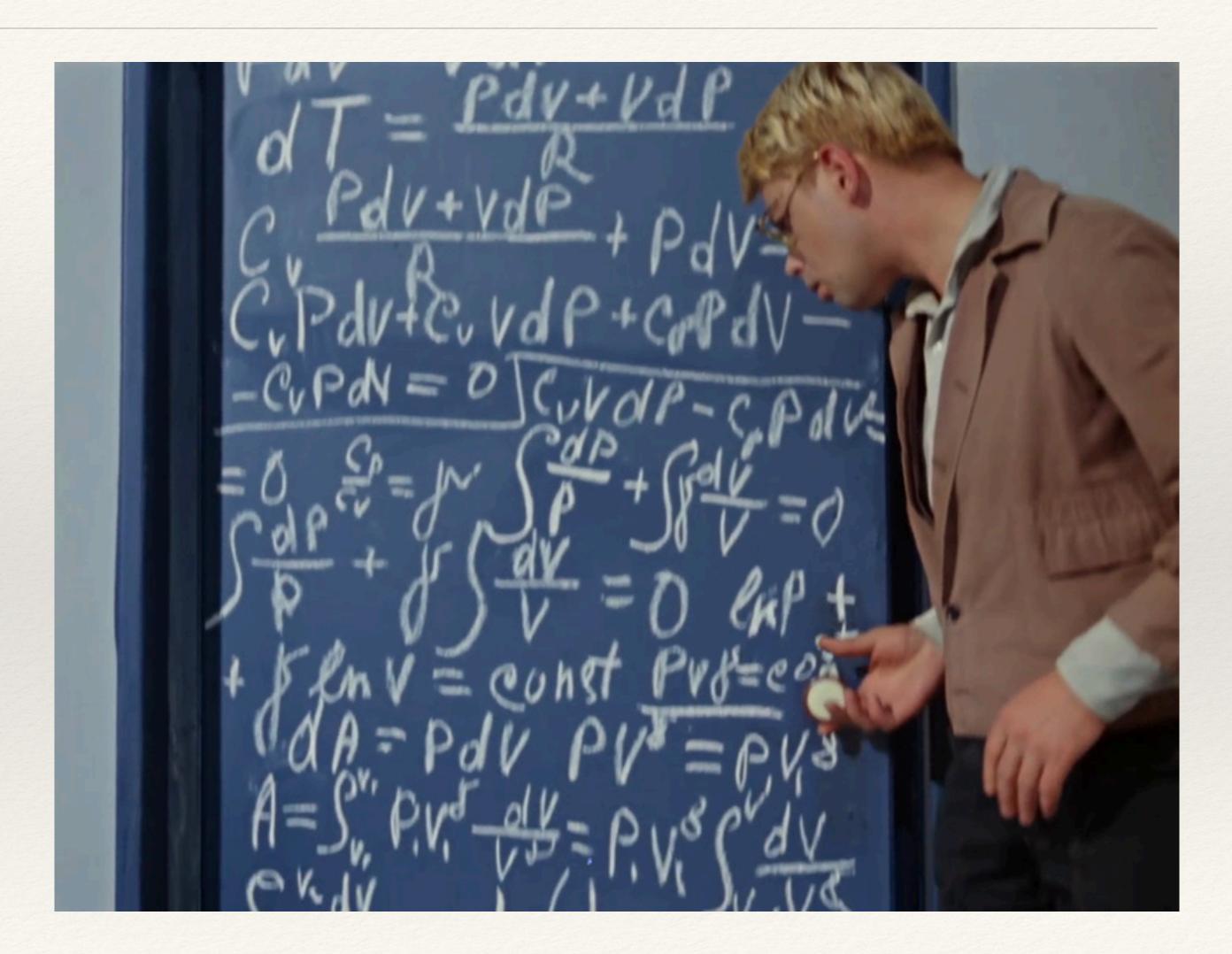
The Request

- * The Business has been reading technology sites
- * They want a machine learning solution
- * They're looking for you to find a suitable problem... then solve it



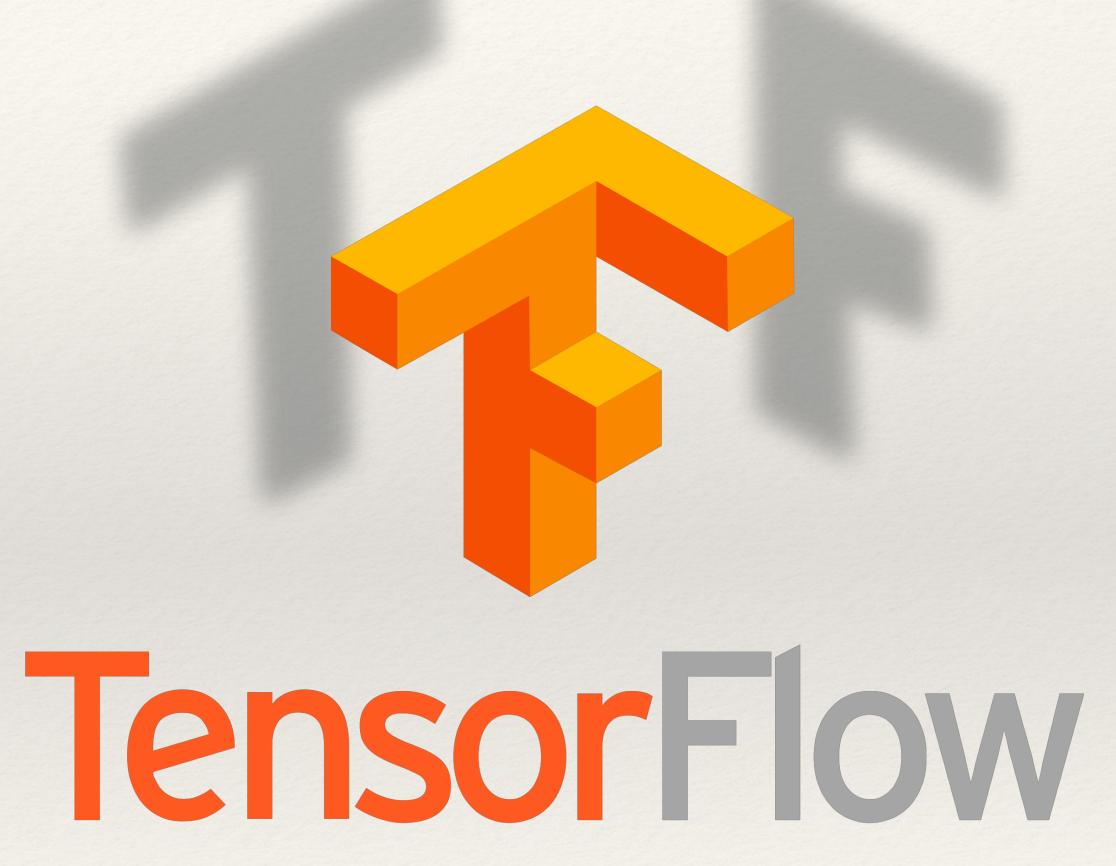
Your Starting Point

- * You have no prior background in machine learning
- * You have no prior background in advanced math / data



The Solution (aka "the Plan")

- * A very pragmatic, quick intro to machine learning
- * An equally pragmatic, quick intro to TensorFlow
- * 4 Java coding demonstrations

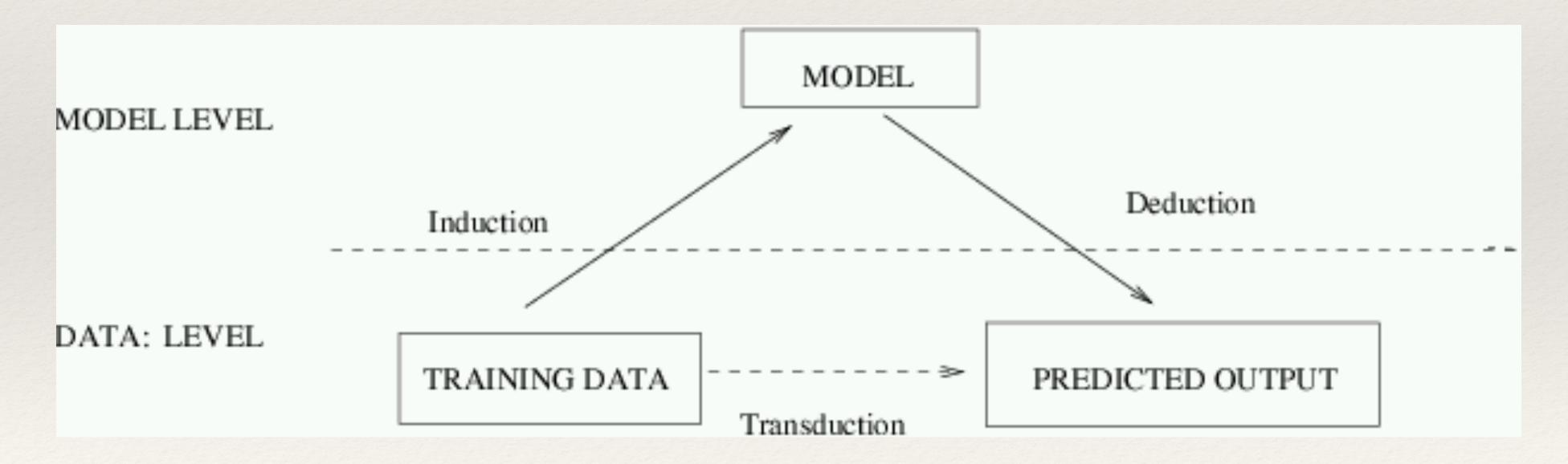


What is Machine Learning?

- * Teach a computer to solve 2 + 2 1 = = = arithmetic
- * Teach a computer to solve $2 + _ 1 = 4 <$ algebra
- * Teach a computer to solve $2 _ 3 _ 1 = 4 < ML$

Supervised Learning

- * Start with "tagged" training data
- * Use 2/3 of your data to evolve ("train") an algorithm called a Model
- * Use remaining 1/3 of your data to test your model with some predictions



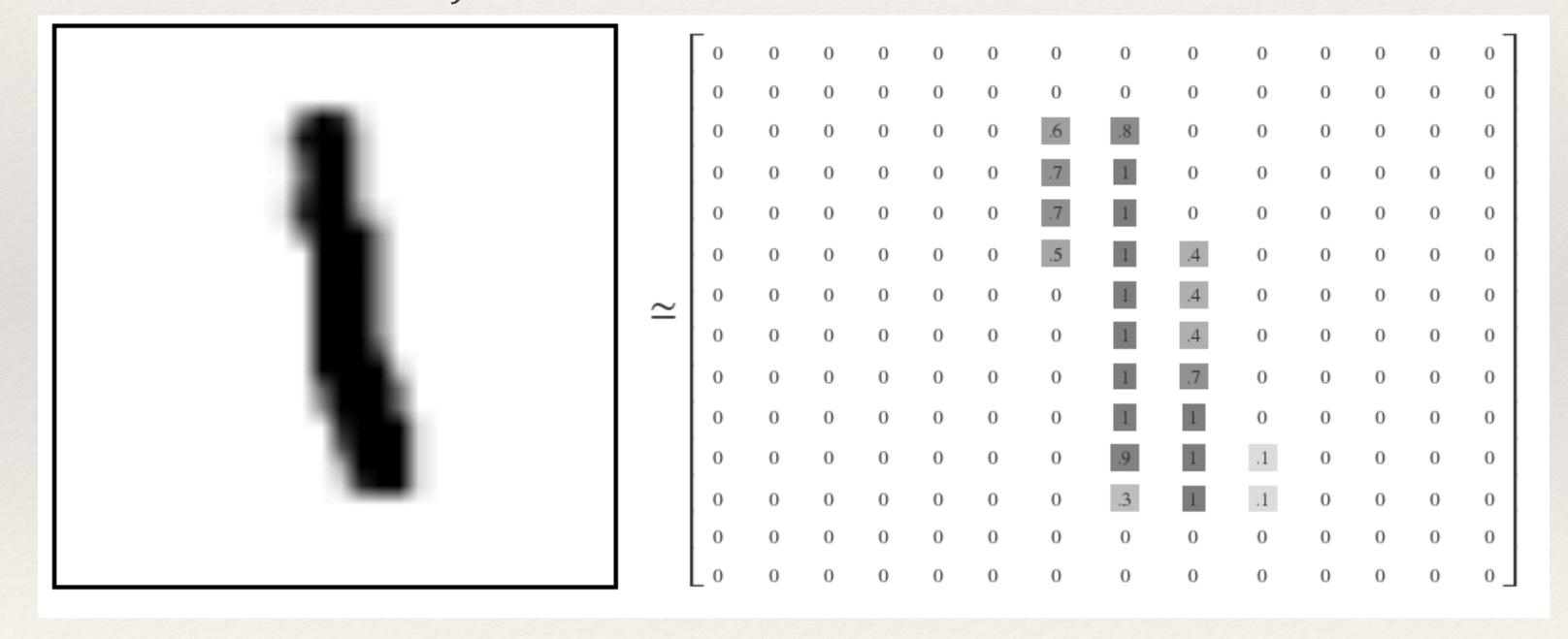
Induction





No Such Thing as Non-Numeric Data

- * Do you smoke? (0 = No, 1 = Yes)
- * Images and time data are just number matrices

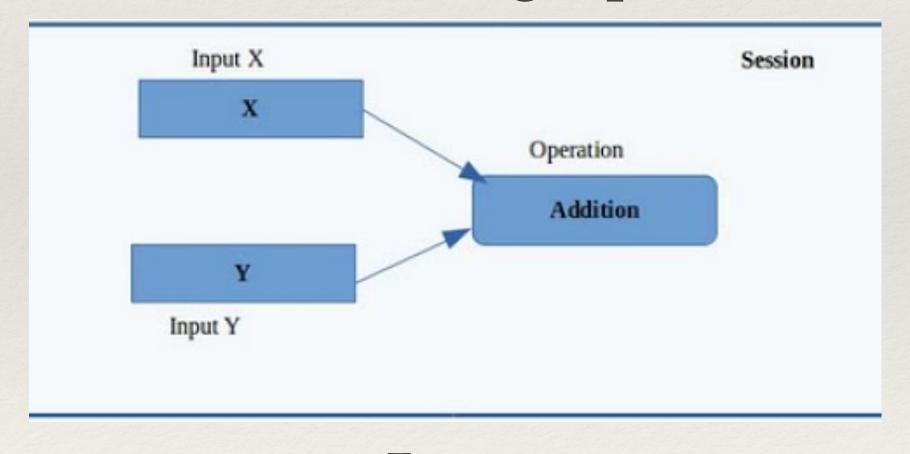


TensorFlow in 1 slide

- * TensorFlow is an Open Source, C++ library for ML
- * Its primary (think 90%+) wrapper is Python
- * Java integration requires... craftiness!



- * It provides containers called tensors for constants & variables & ops
- * Sequenced collections of tensors are called graphs
- * It provides execution environments for graphs to "flow" (train and predict)



Tensors

The TensorFlow Java SDK

- * Available as a maven dependency (and other formats)
- * Wraps maybe 20% of the TensorFlow classes



What's in the 20%?

- * Basic model creation
- * Model training
- * Inline predictions
- * Predictions across the network



Essential Classes

org.tensorflow.Tensor<T>

org.tensorflow.Output

org.tensorflow.Graph

org.tensorflow.Session



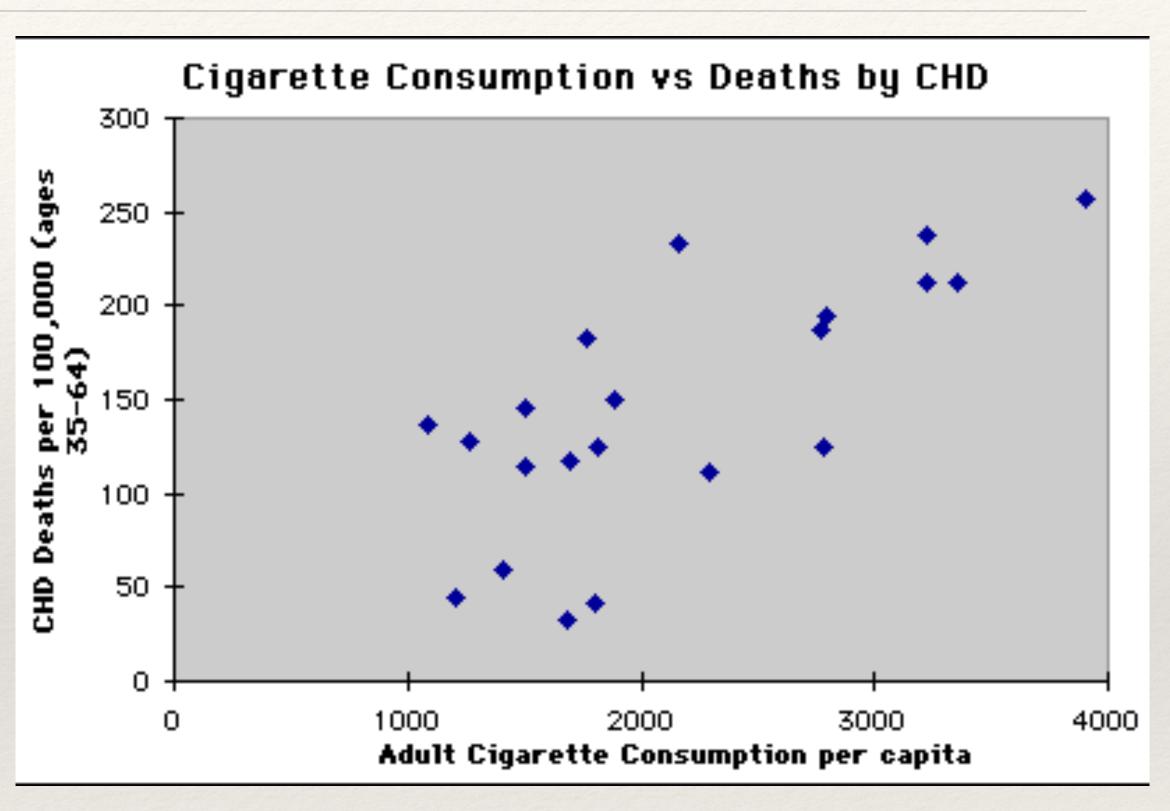
Hands On Lab

Demo 1

Adding 2 numbers

Start with Raw Data

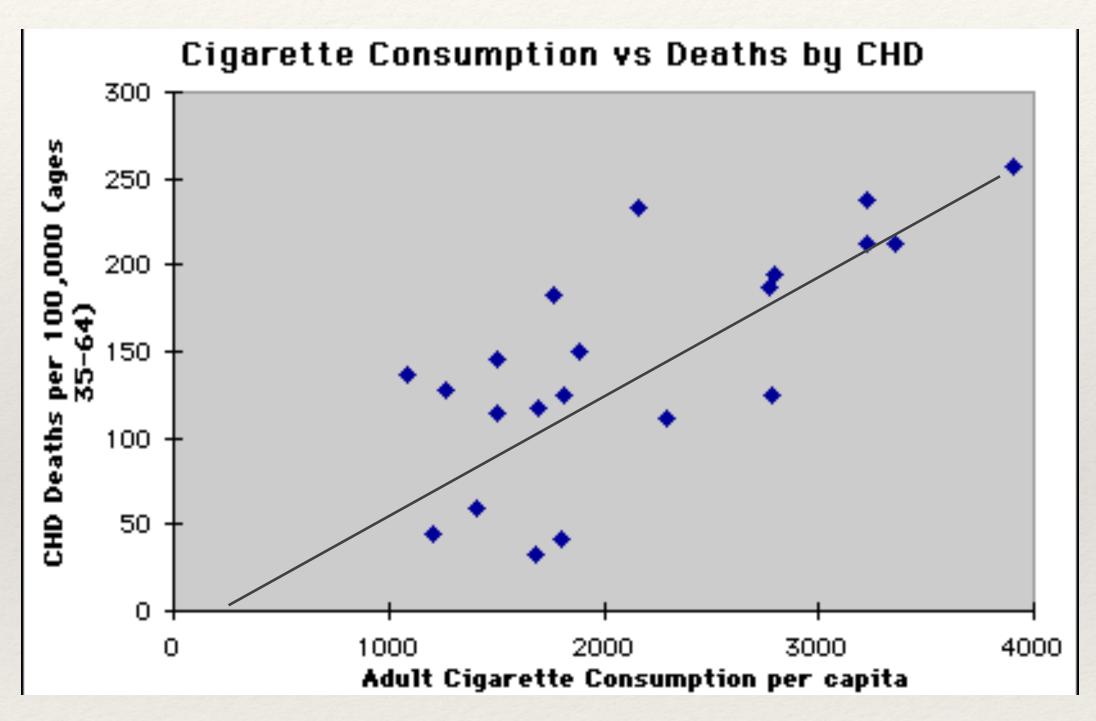
- * Let's go back to a prior example: smoking
- * We can represent this as cigarettes-percapita — X
- * We can also represent the number of deaths Y



Raw Data Points

Predict via Linear Regression

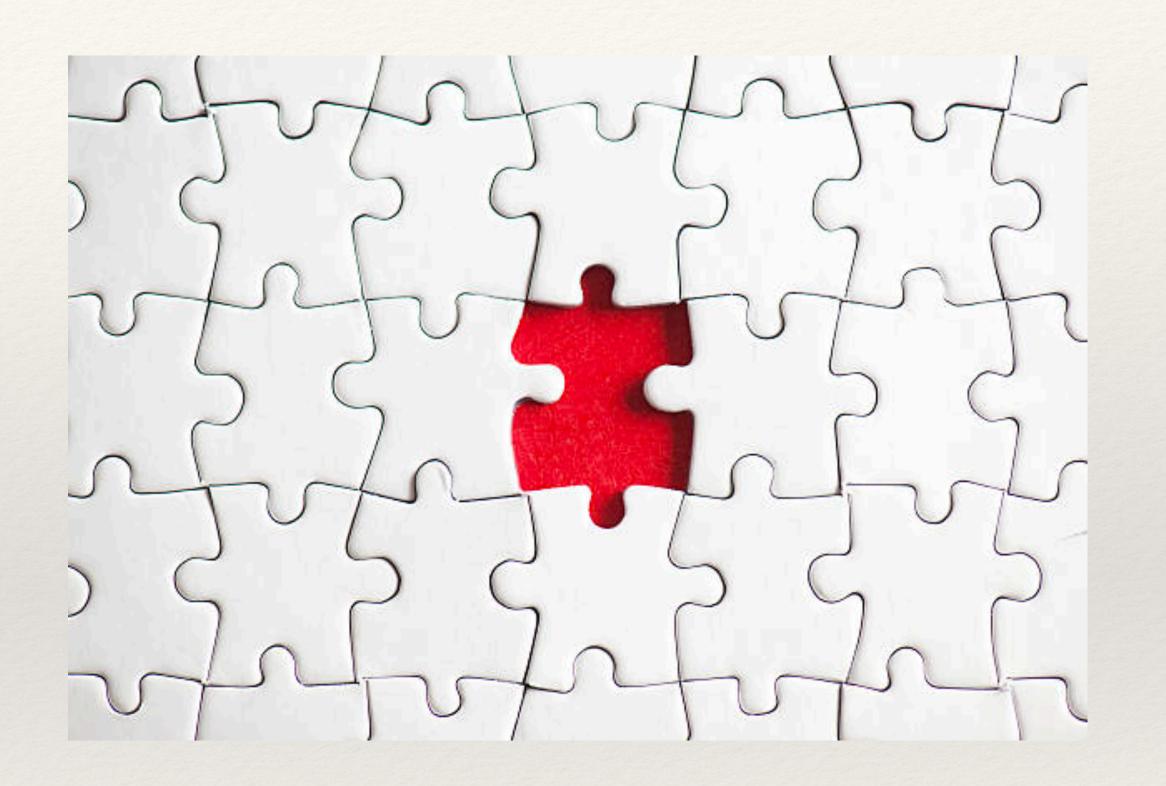
* y = mX + b (worst math we'll touch, I promise!)



Fit Line

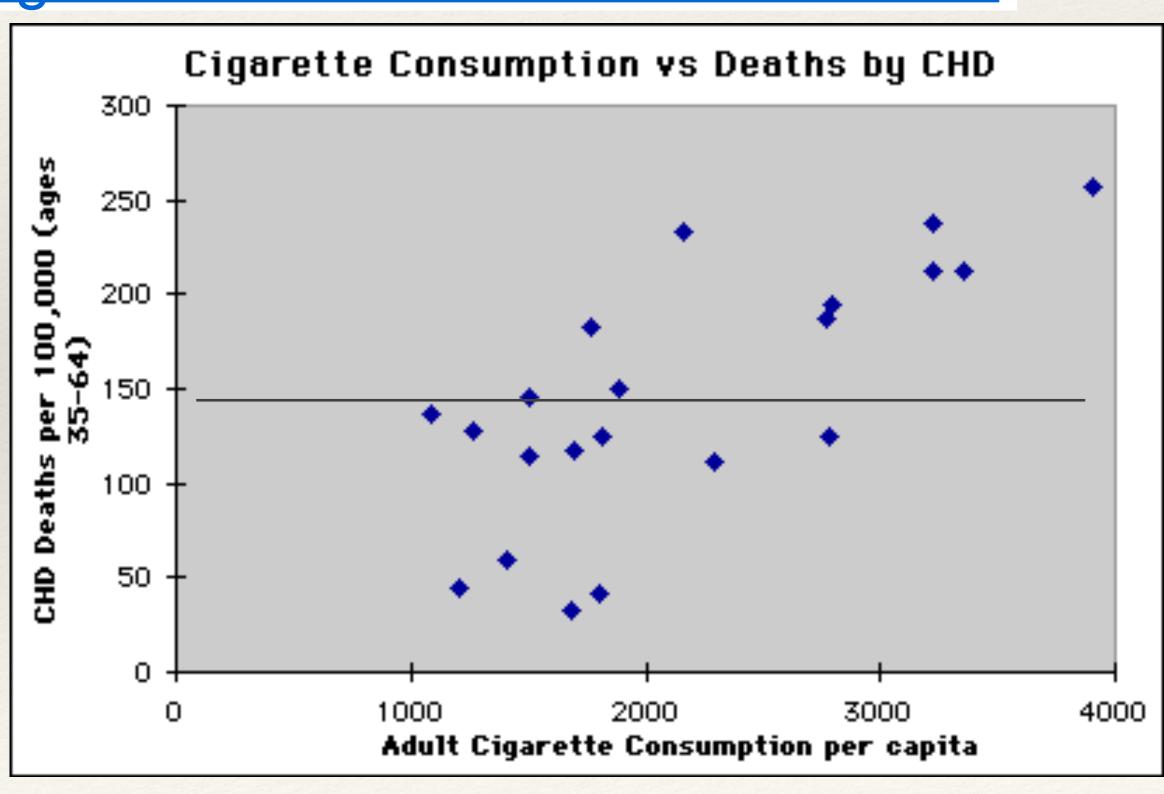
Evolving a model

- * We want a class that adjusts the model to fit the data
- * TF provides the GradientDescentOptimizer
- * Of course this isn't in the 20%:-(



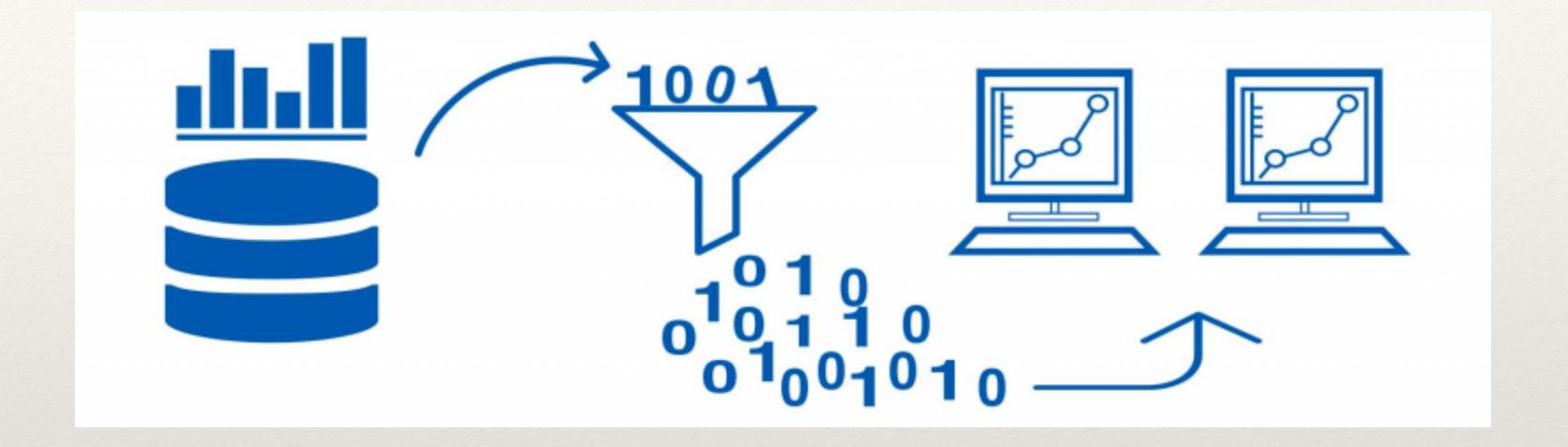
Getting the GradientDescentOptimizer

- * You can download a binary of this model online http://derekferguson-6c9ab059-eval-test.apigee.net/hello?w=3.0&b=2.0&lr=0.1
 - * w Provide the initial "slope"
 - * b Provide the initial "y intercept"
 - * lr Provide the "learning rate"



New TensorFlow Concepts

- * Graph importation
- * Place holders
- * Feeds





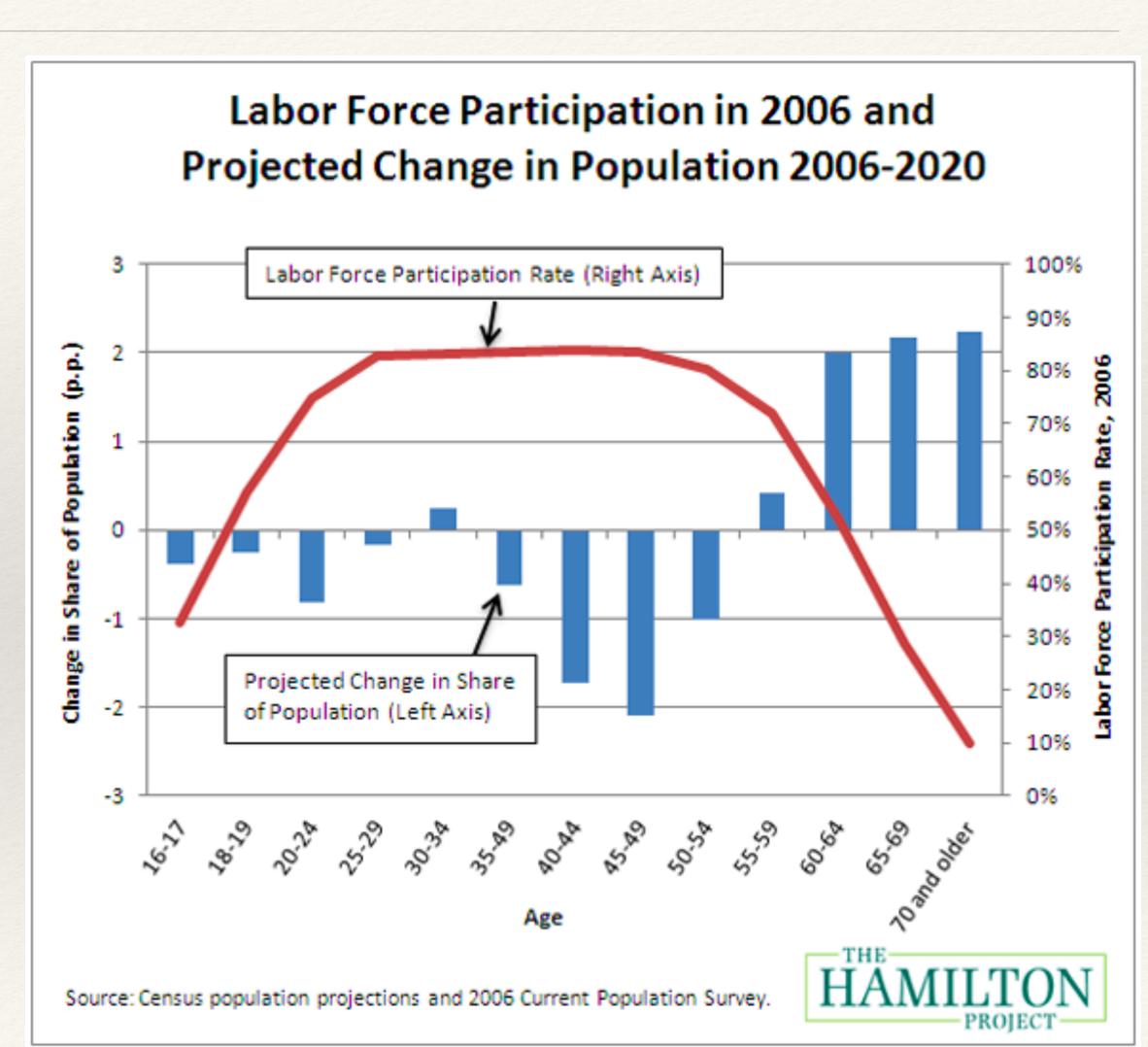
Hands On Lab

Demo 2

Training a model

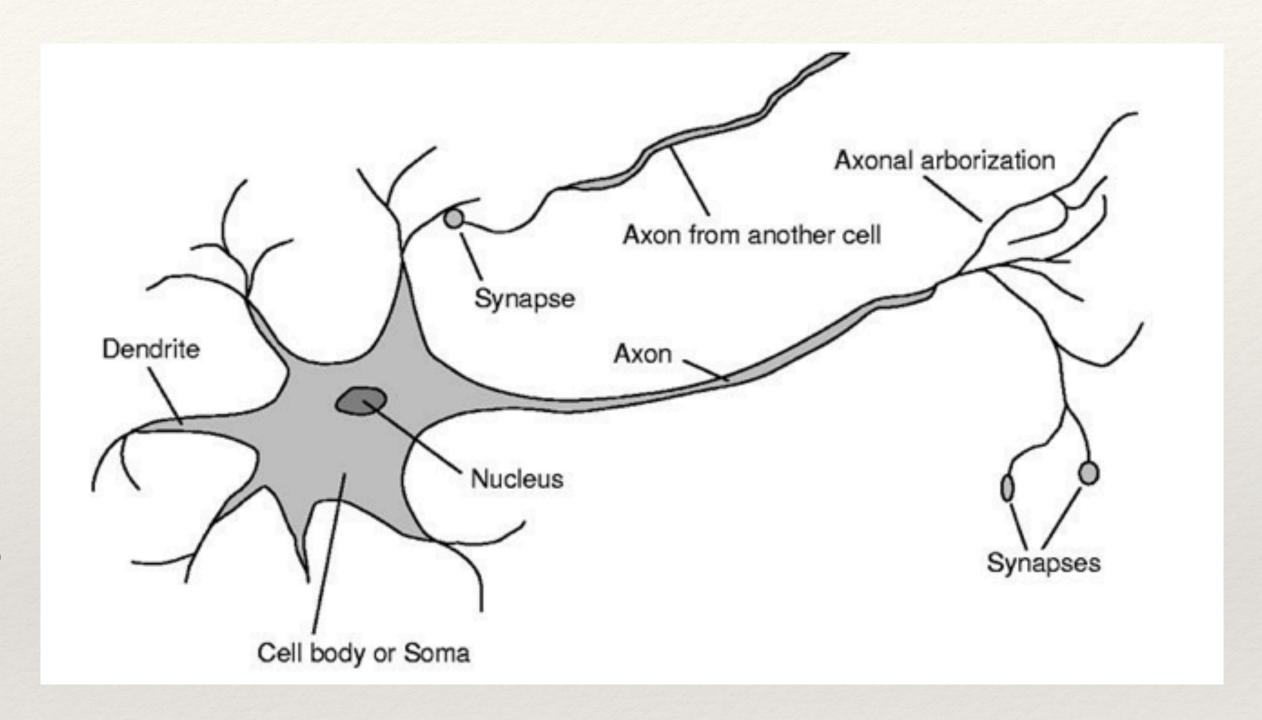
Is everything a line? (Clue: No)

- * At some point, smoking deaths level off
- * Participation in the work force peaks by age
- * Recognizing images of cars and bicycles definitely not a line



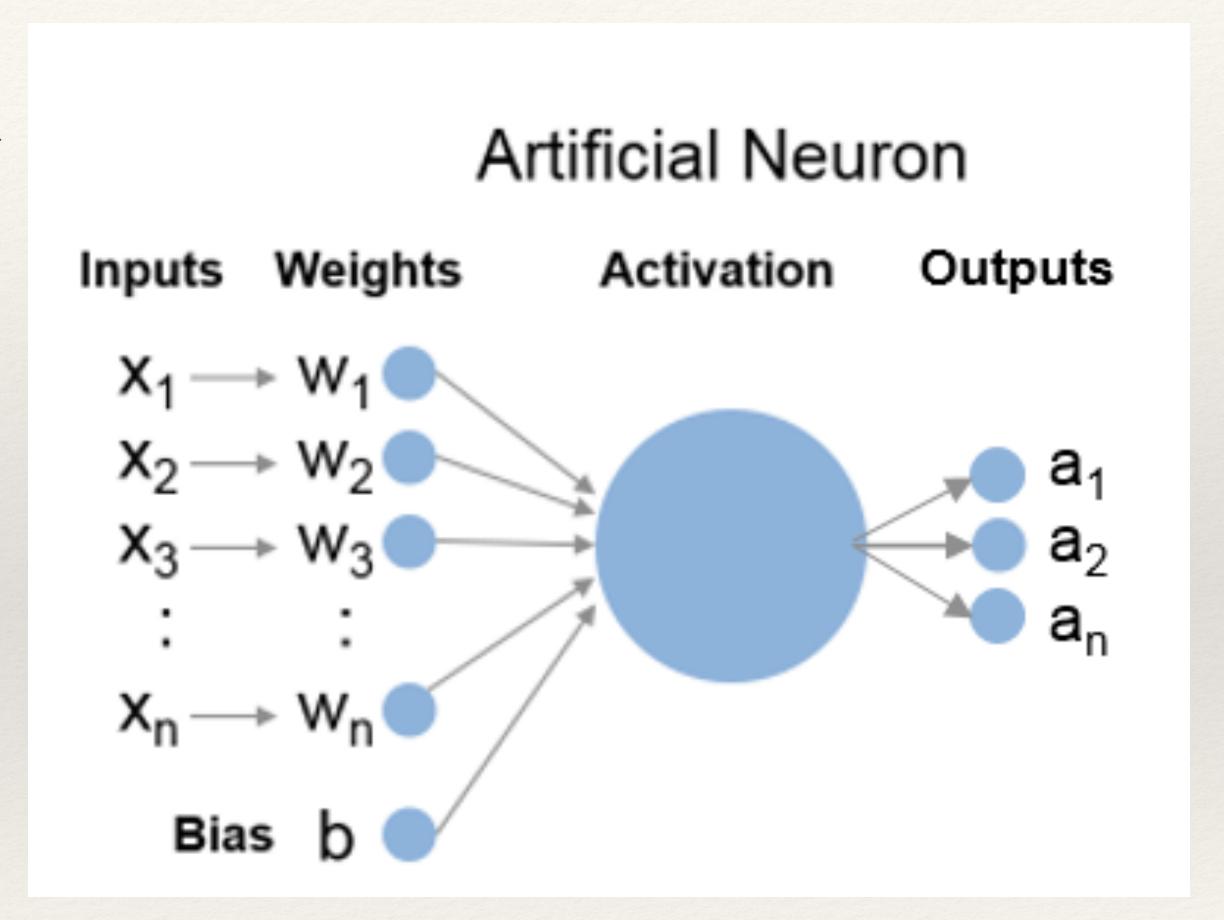
Neural Networks in Nature

- * Human nervous system has neurons that specialize
- * Specialized neurons work together on larger problems
- * Connections between neurons become complex



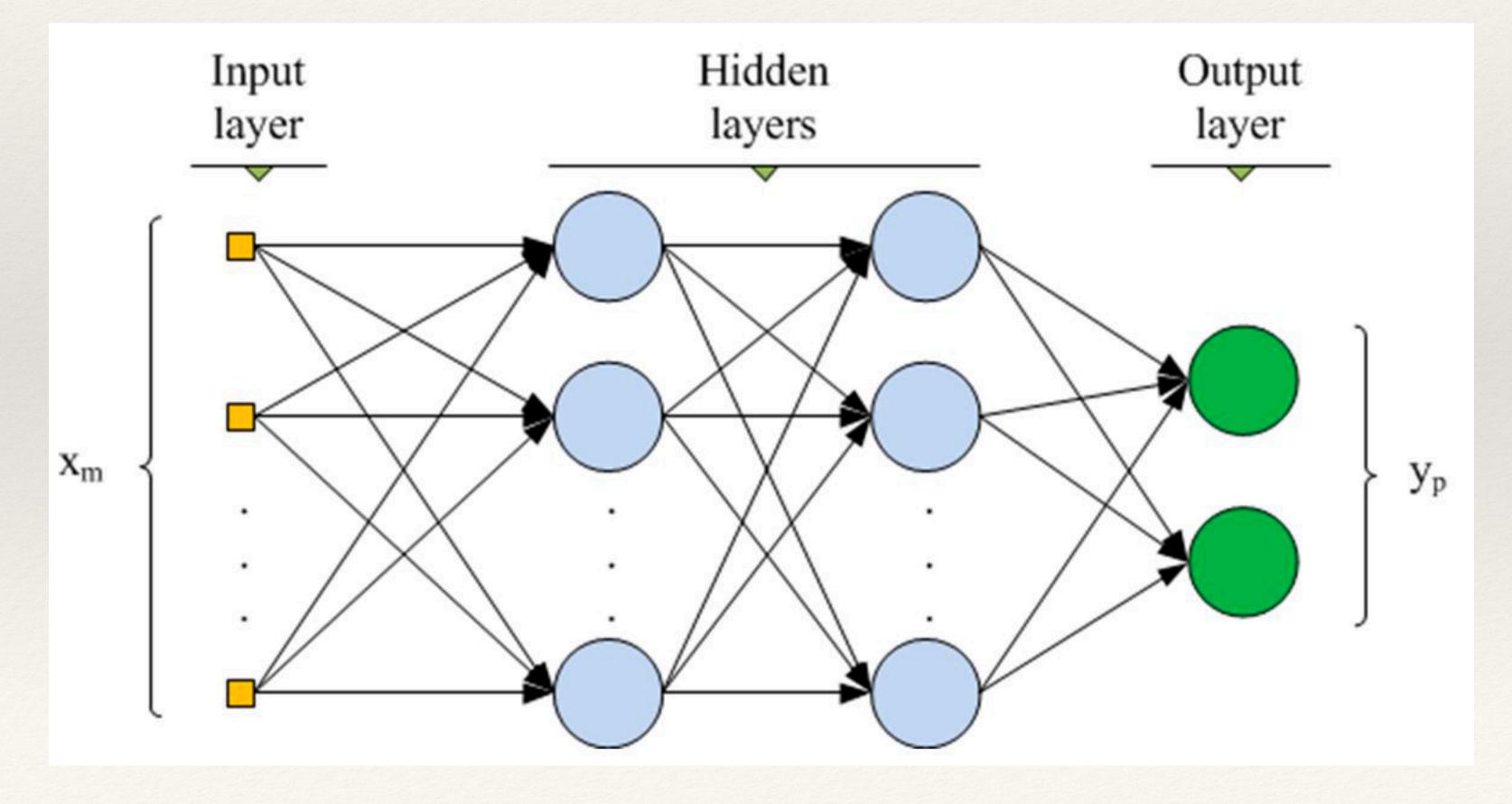
Neural Networks in TensorFlow - pt 1

- * In TensorFlow, this is modeled/trained as...
 - * Weights sensitivity to each input (feature or neuron)
 - * Biases constant added to output



Neural Networks in TensorFlow - pt 2

- * TensorFlow provides a DNNClassifier for neural networks
- * Of course, this isn't in the 20%, either...





Hands On Lab

Demo 3

Invoking a model inline

Deploying Trained Models

- * Before TF 1.8, Python users used Flask
- * Since TF 1.8, TensorFlow Serving is available

FROM tensorflow/serving

COPY exports/1530401406 /models/model

EXPOSE 8500

EXPOSE 8501

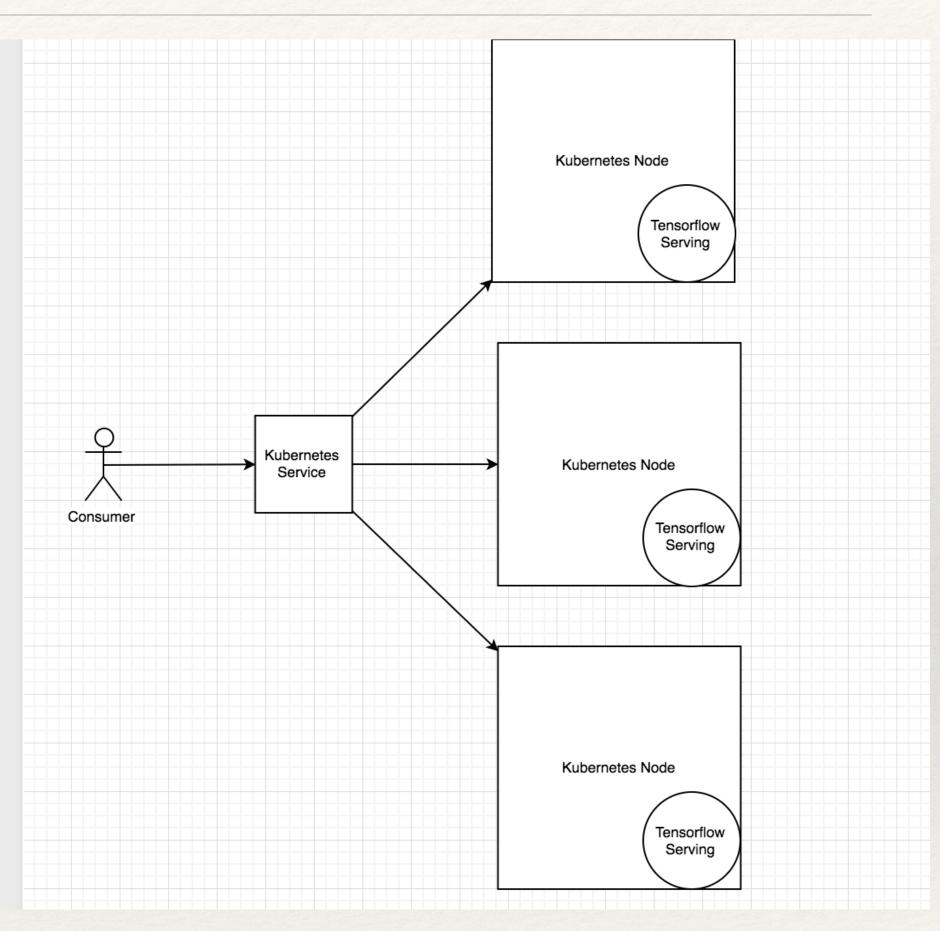
Executing your Docker predictor

```
docker build -t predictor.
```

```
docker run
```

```
-р 8500:8500 -р 8501:8501
```

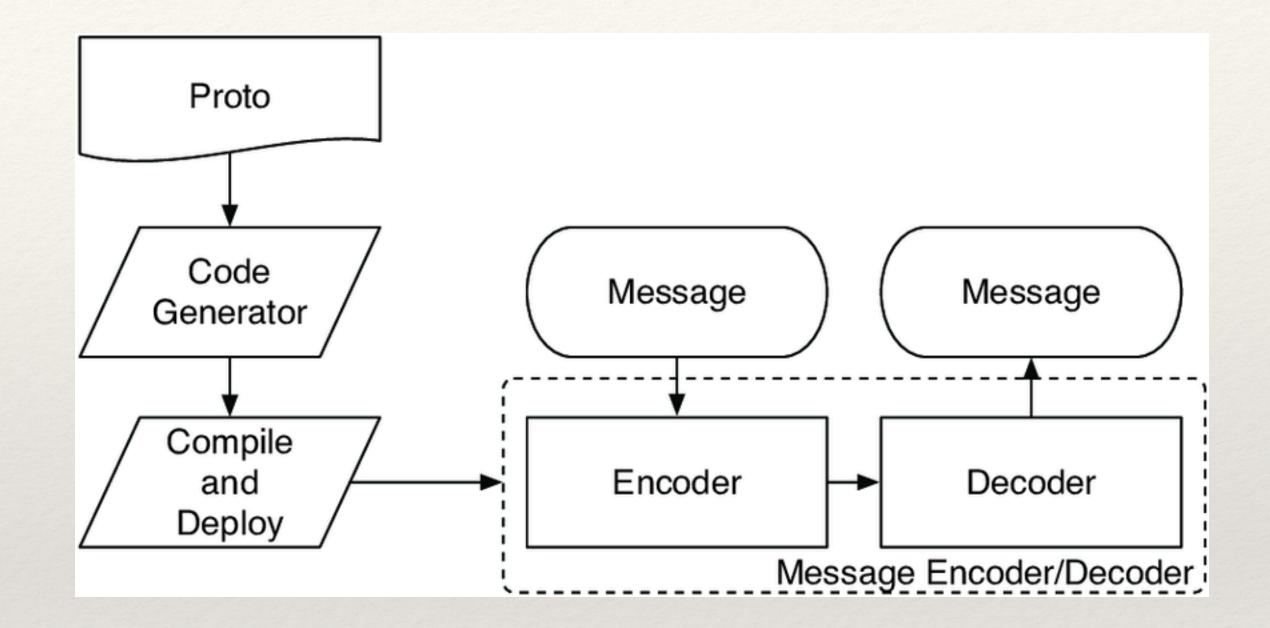
predictor



Kubernetes with TF Serving

Protocol Buffers

- * Developed by Google
- * Intended to be smaller and faster than XML
- * Define message formats in .proto files and generate transmission code



Protocol Buffers and TensorFlow

- * ManagedChannelBuilder communications engine
- * Tensor(Shape)Proto protobuf(s) for Tensors
- * ModelSpec protobuf for an entire model
- * PredictResponse value returned from execution

Better Yet - Let's Just Use REST

saved model cli.py show -dir modeler -all

```
+ ---
 GetPrediction
GetPrediction
                     http://localhost:8501/v1/models/model/versions/1:predict
    POST V
                                                Pre-request Script
                  Headers (1)
                                    Body 🔵
                                                                       Tests
Authorization
 form-data x-www-form-urlencoded raw binary JSON (application/json)
             "signature_name":"predict",
             "instances": [
                     "attscale":[2.0],
                     "bmiwho":[3.0],
"bothmapa2":[1.0],
                     "dage":[14.0],
"drink":[1.0],
                     "gafsex":[1.0],
"smokenow":[1.0],
   10
   11
                     "tottot":[5.0],
   12
   13
                     "zmainrea2":[6.0],
   14
15
16
17
                     "zrttime":[3.0]
```

10 Input Features (X)



Hands On Lab

Demo 4

TensorFlow Serving via REST

What we learned...

- * How machine learning works
- * What TensorFlow is
- * Where to get the required libraries
- * How to obtain, train and invoke TF models

Where to from here?

- * More of the missing 80%
- * Advanced model creation
- * Unsupervised learning
- * Remember: the whole thing is Open Source!

Q&A

Contact me! English или по-русски!

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