



# Introducing Machine Learning @Google with TensorFlow

Machine Learning for Marketeers

15th May 2018

Rokesh Jankie

[rokesh@google.com](mailto:rokesh@google.com)

Customer Engineer

Google Cloud

# Shift in Computing





#io17

2 Billion



#io17







# Google's Mission

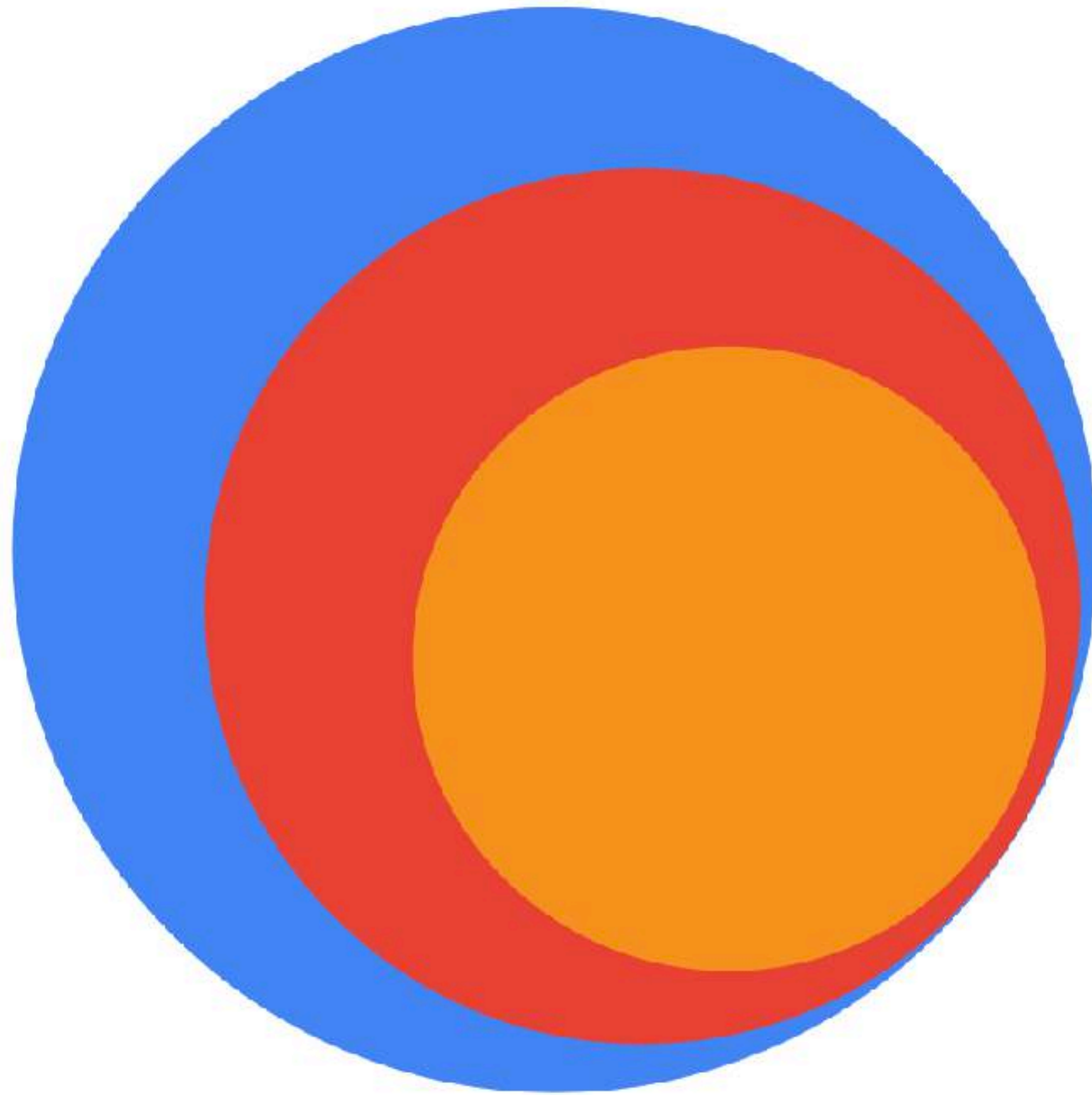
Organize the world's information  
and make it universally  
accessible and useful

---



Seven cloud products with  
**> ONE BILLION**  
Users

# Putting things in perspective



## Artificial Intelligence

The science to make things smart



## Machine Learning

Building machines that can learn



## Neural Network

A type of algorithms in machine learning





## Supervised Learning

Labelled data = data + output (target, class, etc) variable. Goals is to learn the relationship between data and output

## Semi-Supervised Learning

Partially labelled data

## Online/Active Learning

Real-Time incremental learning on data streams



## Unsupervised Learning

Unlabelled data.  
Goals is to learn associations, groups, etc

## Reinforcement Learning

Reinforcement Learning allows the machine or software agent to learn its behaviour based on feedback from the environment.



# How Can You Get Started with Machine Learning?

Ready to use Machine Learning models

- Three ways to get started
- Use a Cloud ML API
- Use an ML framework or library
- Develop new ML models



Cloud Vision API



Cloud Speech API



Cloud Job Discovery



Cloud Translation API



Cloud Natural Language API



Cloud Video Intelligence



More flexible, but more effort required



TensorFlow



# What is TensorFlow

- Open source Machine Learning library
- Especially useful for Deep Learning
- For research and production
- Apache 2.0 license



# TensorFlow History

- DistBelief  
Large Scale Distributed Deep Networks

Venue

NIPS (2012)

Publication Year

2012

Authors

Jeffrey Dean, Greg S. Corrado, Rajat Monga, Kai Chen, Matthieu Devin, Quoc V. Le, Mark Z. Mao, Marc'Aurelio Ranzato, Andrew Senior, Paul Tucker, Ke Yang, Andrew Y. Ng

Abstract



Recent work in unsupervised feature learning and deep learning has shown that being able to train large models can dramatically improve performance. In this paper, we consider the problem of training a deep network with billions of parameters using tens of thousands of CPU cores. We have developed a software framework called DistBelief that can utilize computing clusters with thousands of machines to train large models. Within this framework, we have developed two algorithms for large-scale distributed training: (i) Downpour SGD, an asynchronous stochastic gradient descent procedure supporting a large number of model replicas, and (ii) Sandblaster, a framework that supports a variety of distributed batch optimization procedures, including a distributed implementation of L-BFGS. Downpour SGD and Sandblaster L-BFGS both increase the



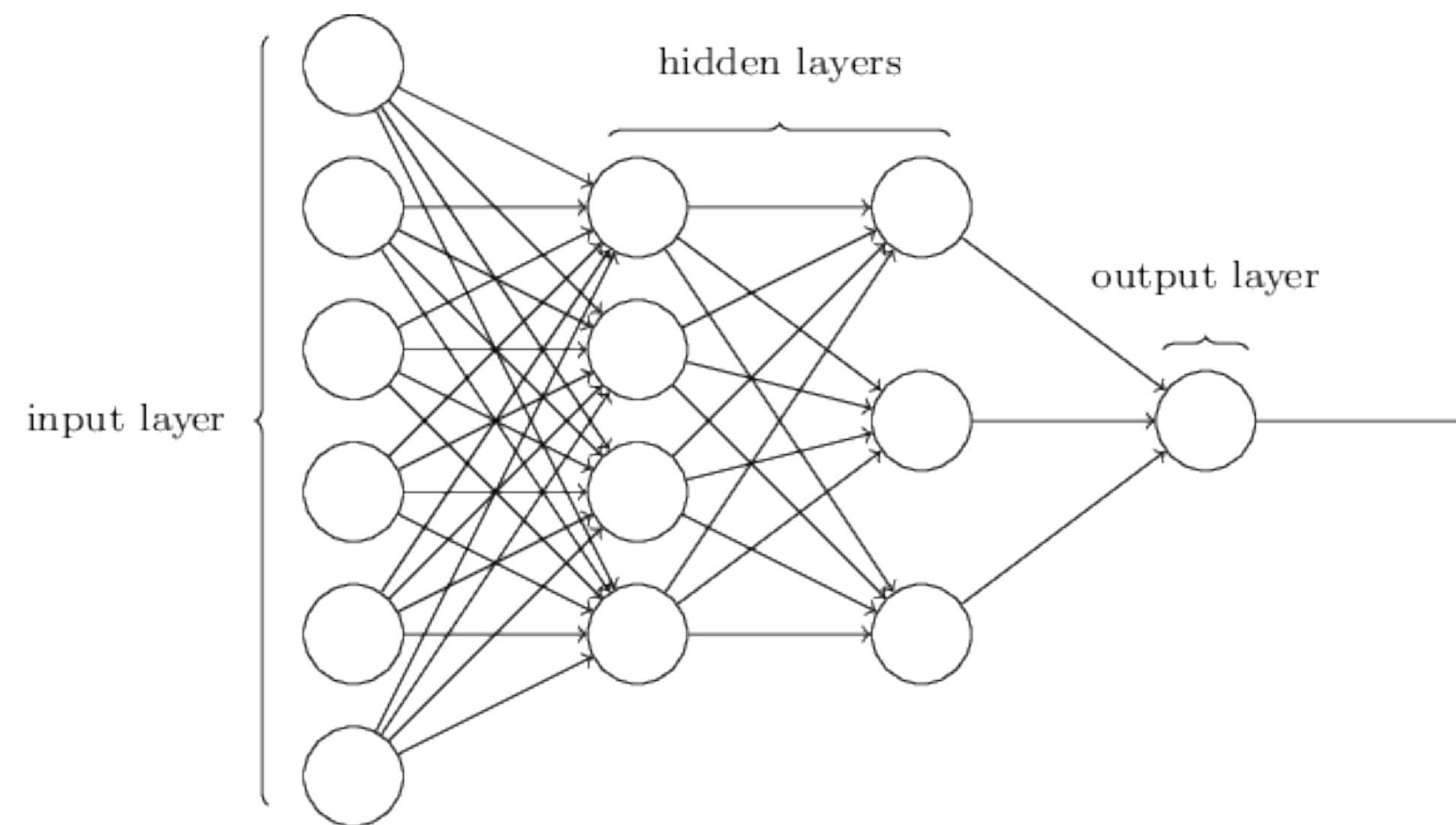
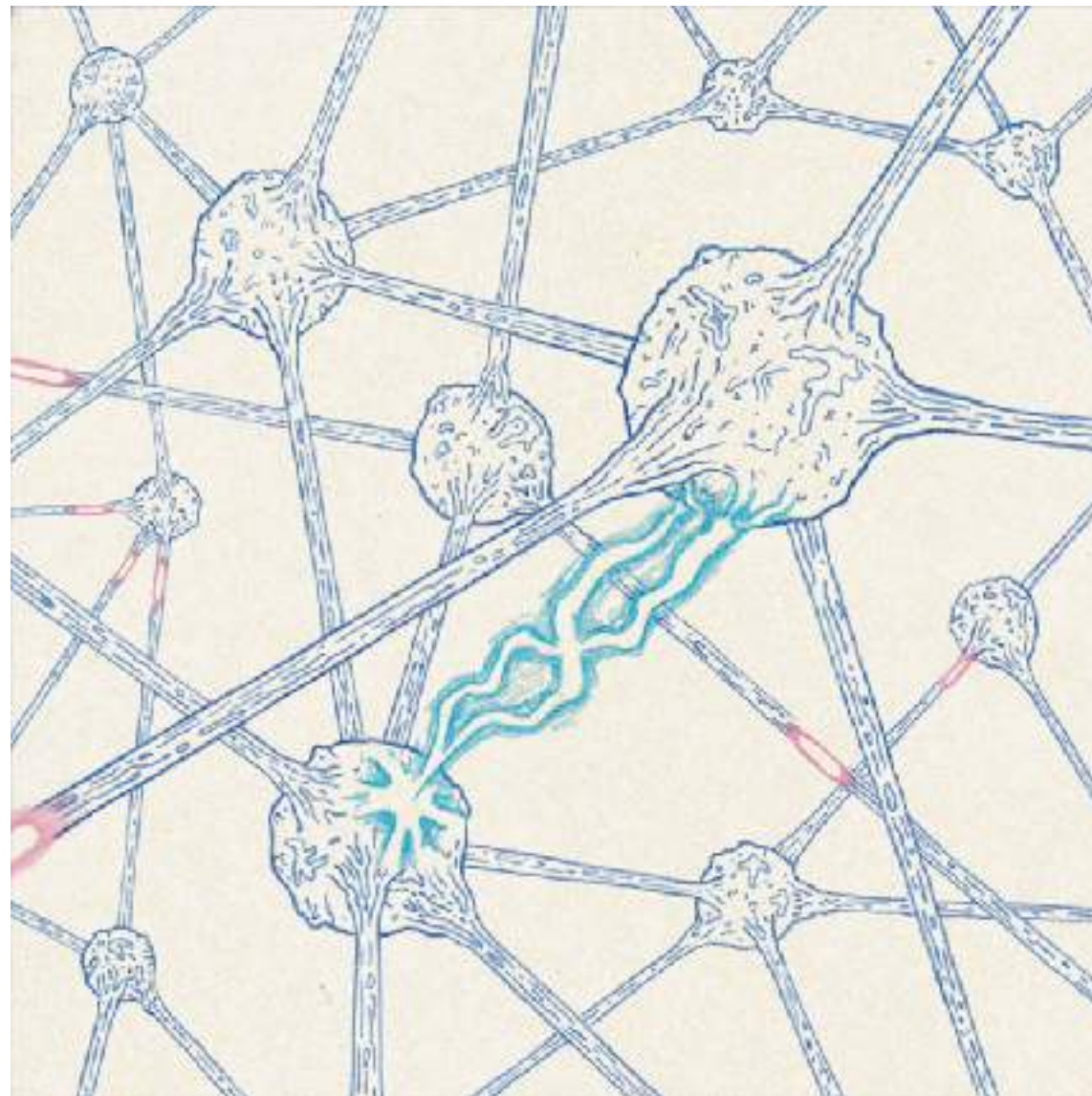
# Jeff Dean!





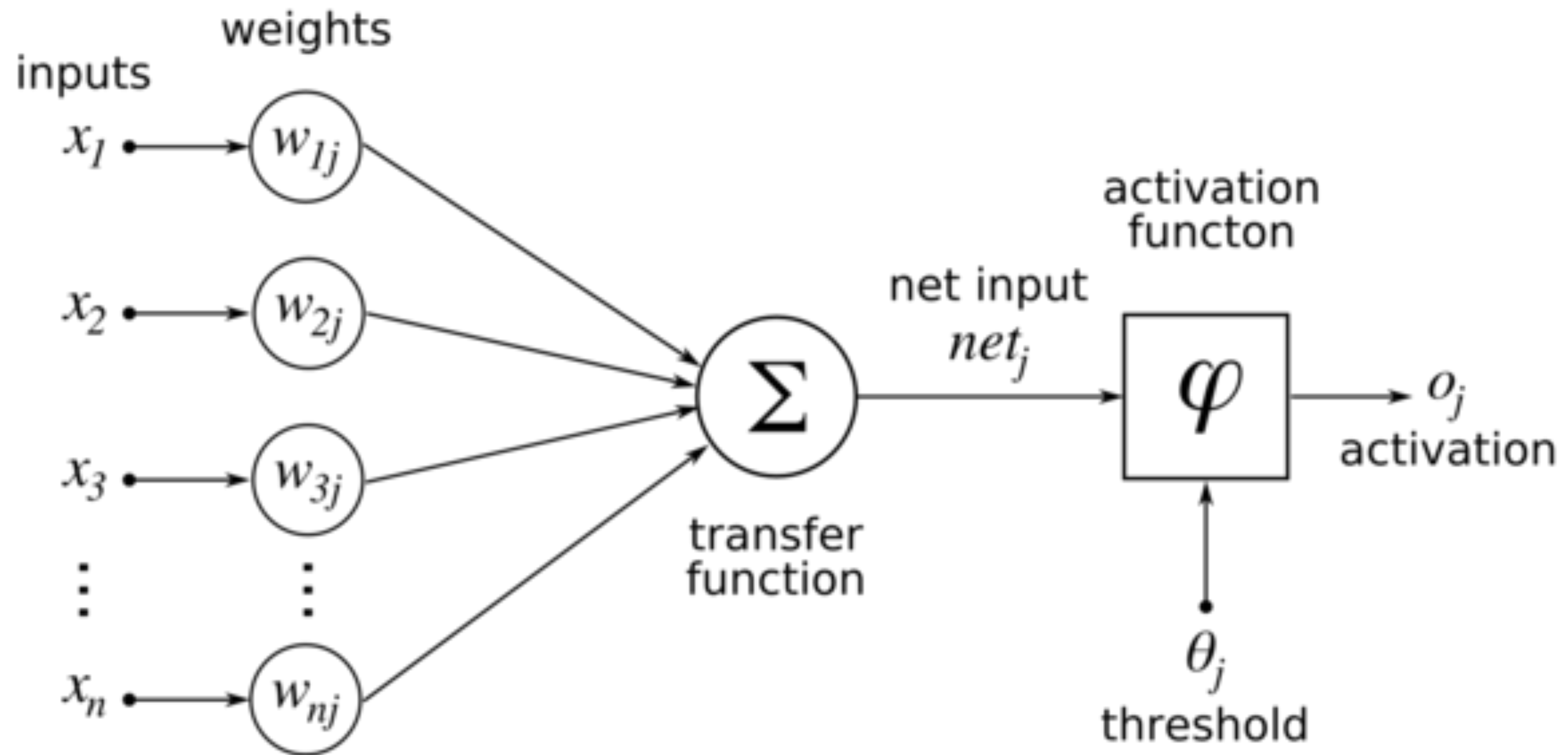
# You mentioned Deep Learning? Wazup?

- The first question to answer: What's a Neural Network ?
- Inspired by Biology:



- Two flavours: Supervised and Unsupervised





# Base Idea of Neural Networks

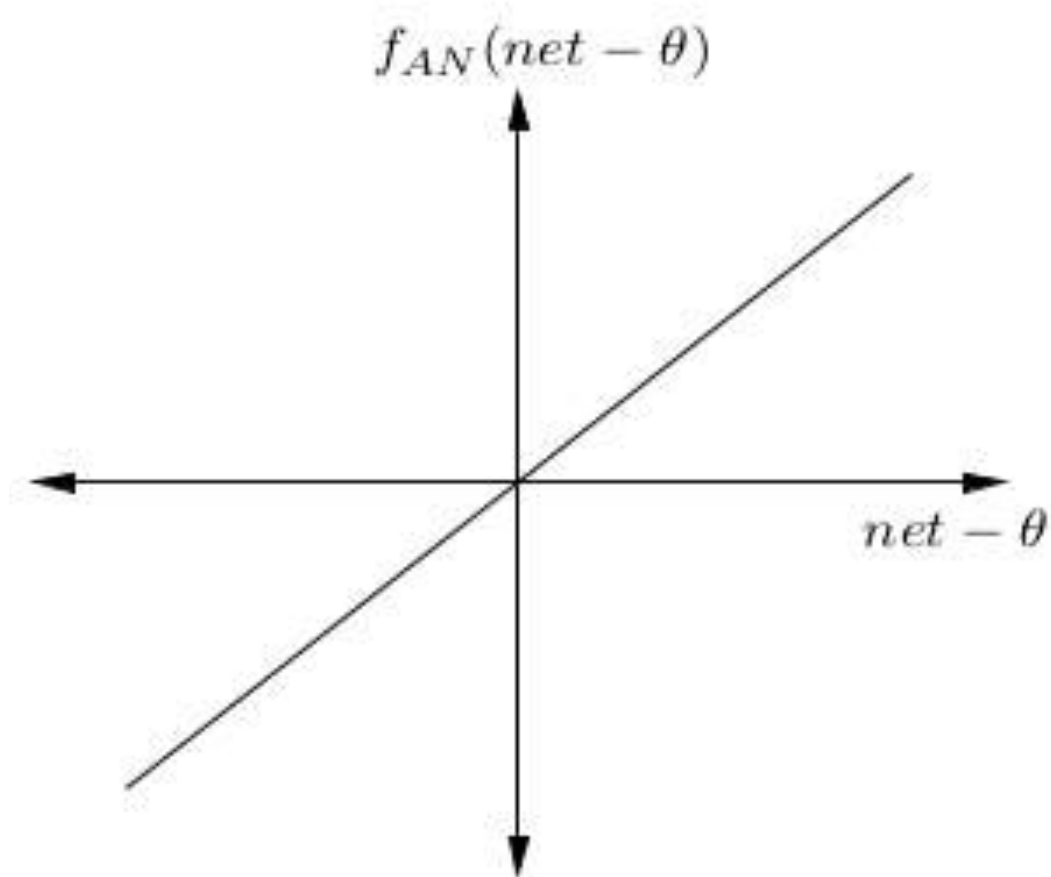


$$f(x) = y$$

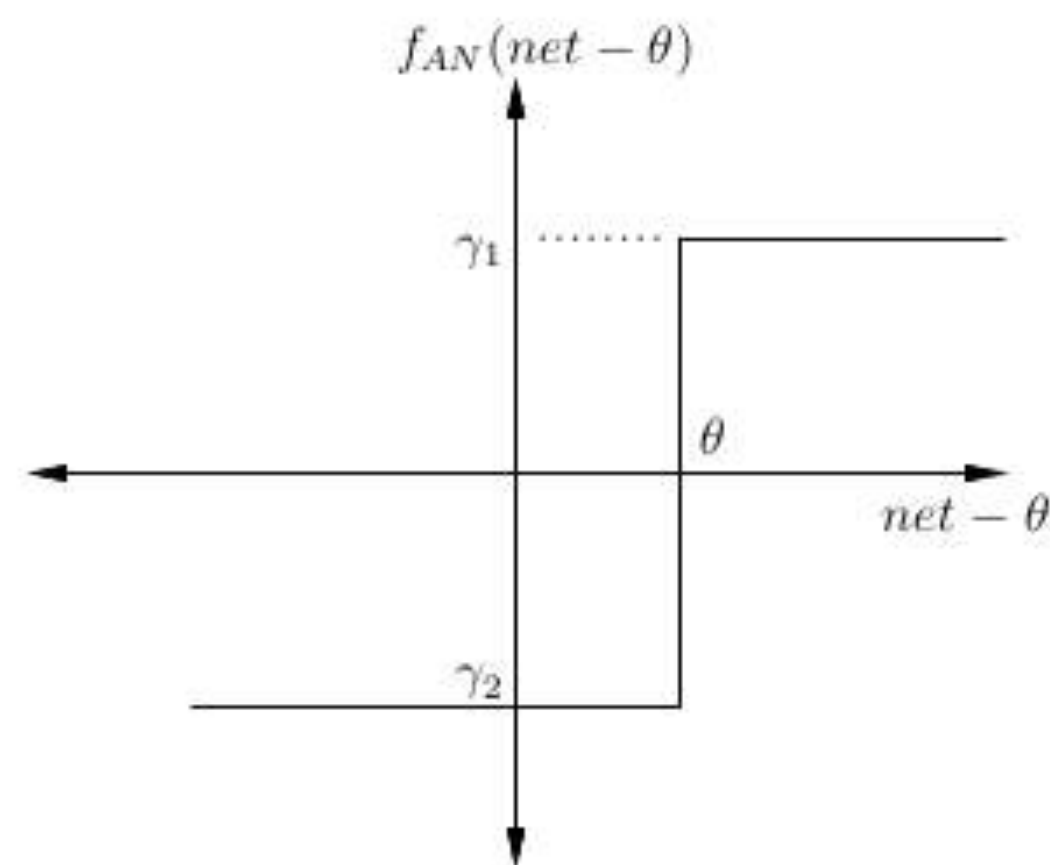
This is what we are trying to solve



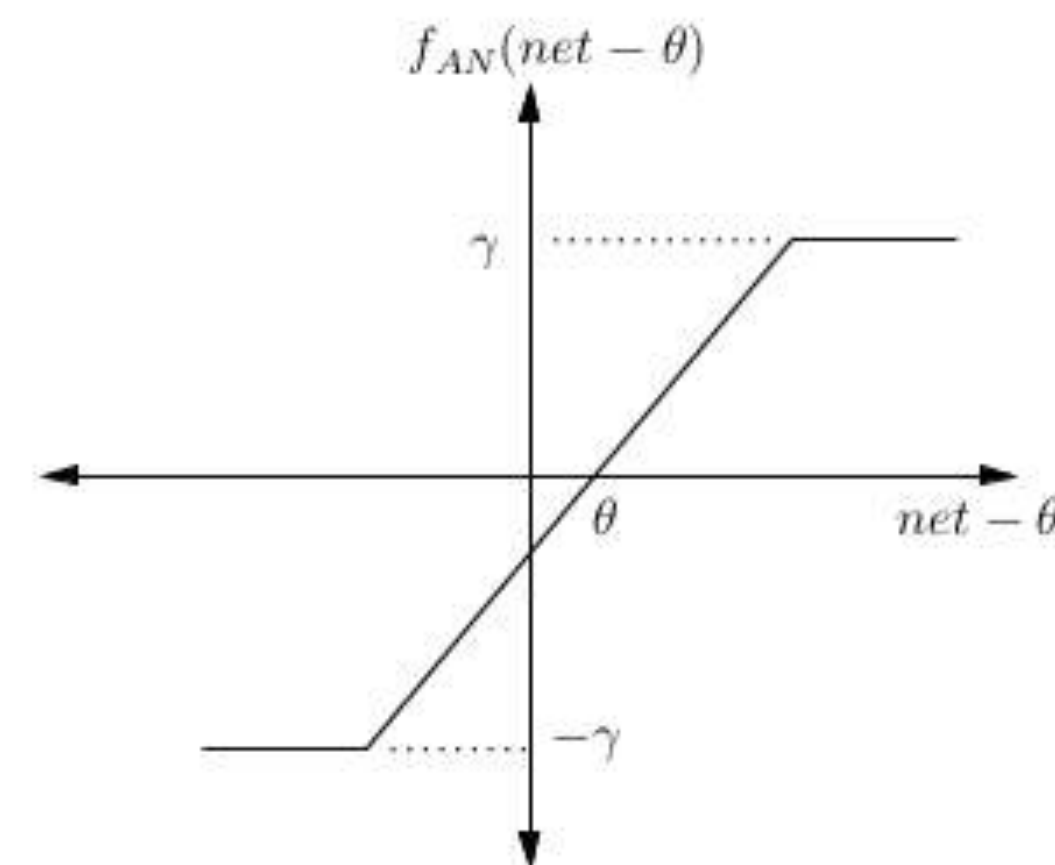
# Activation Functions



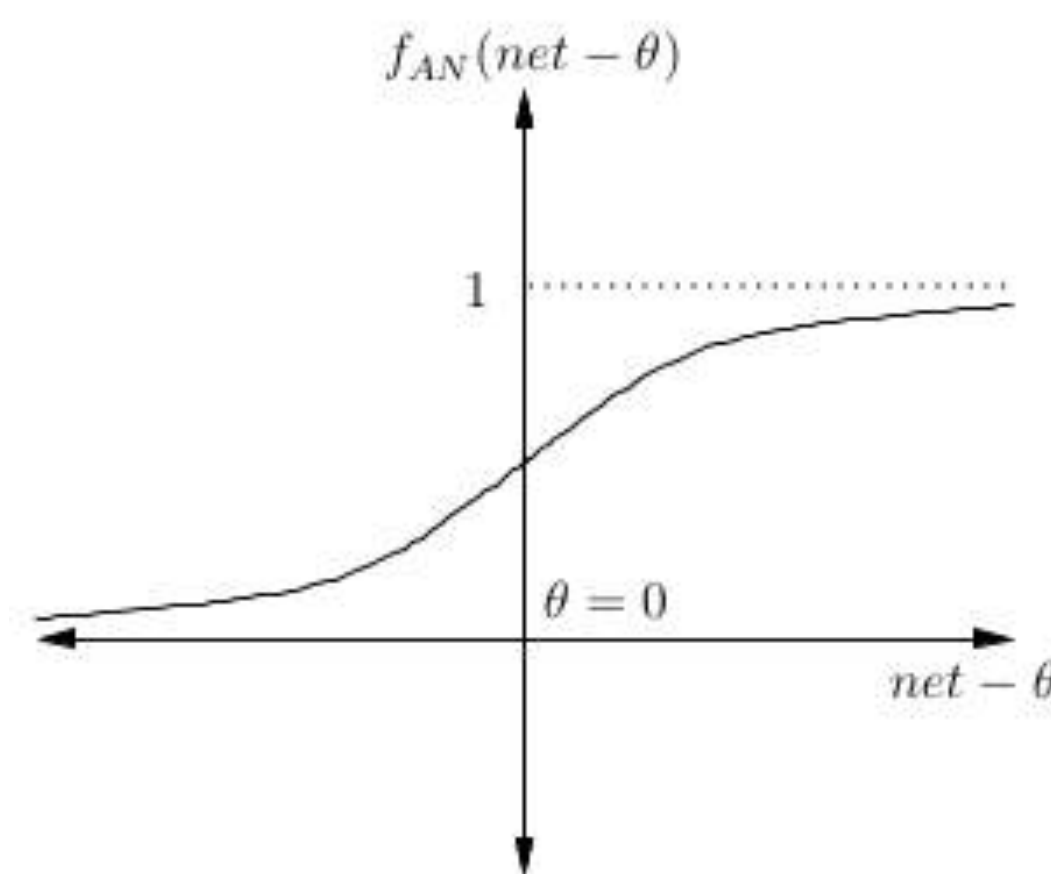
(a) Linear function



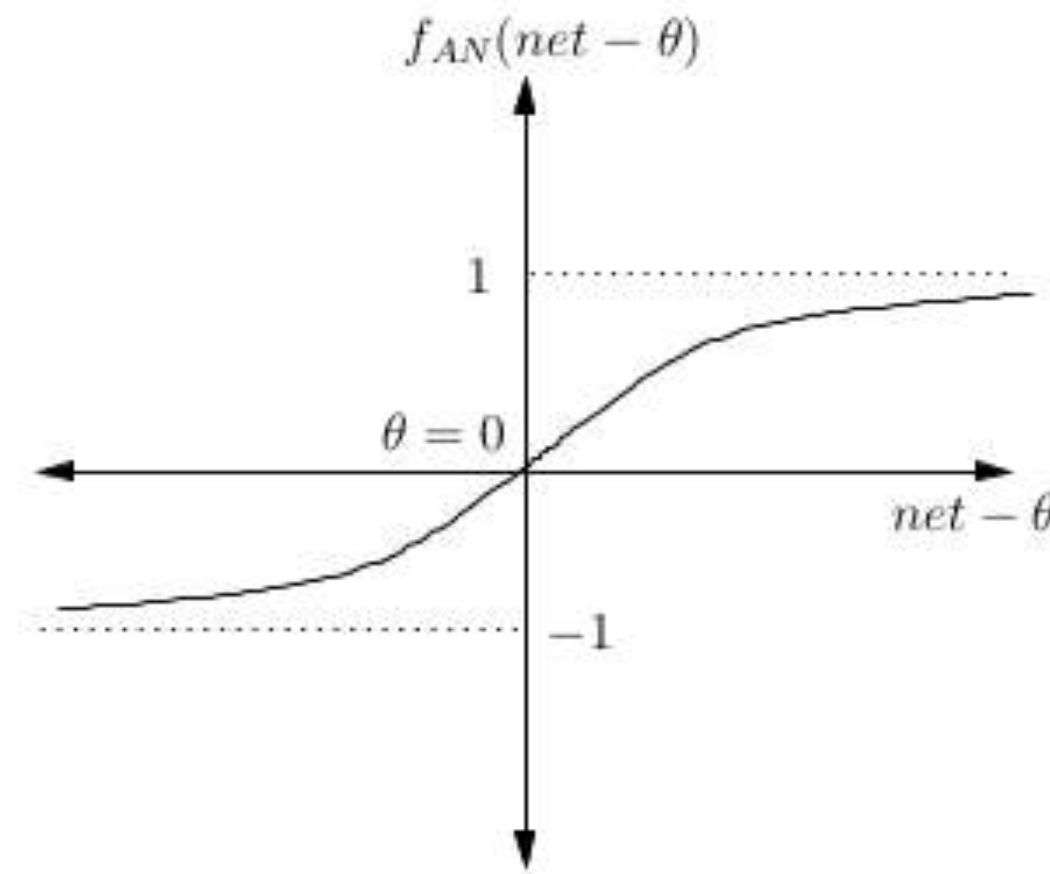
(b) Step function



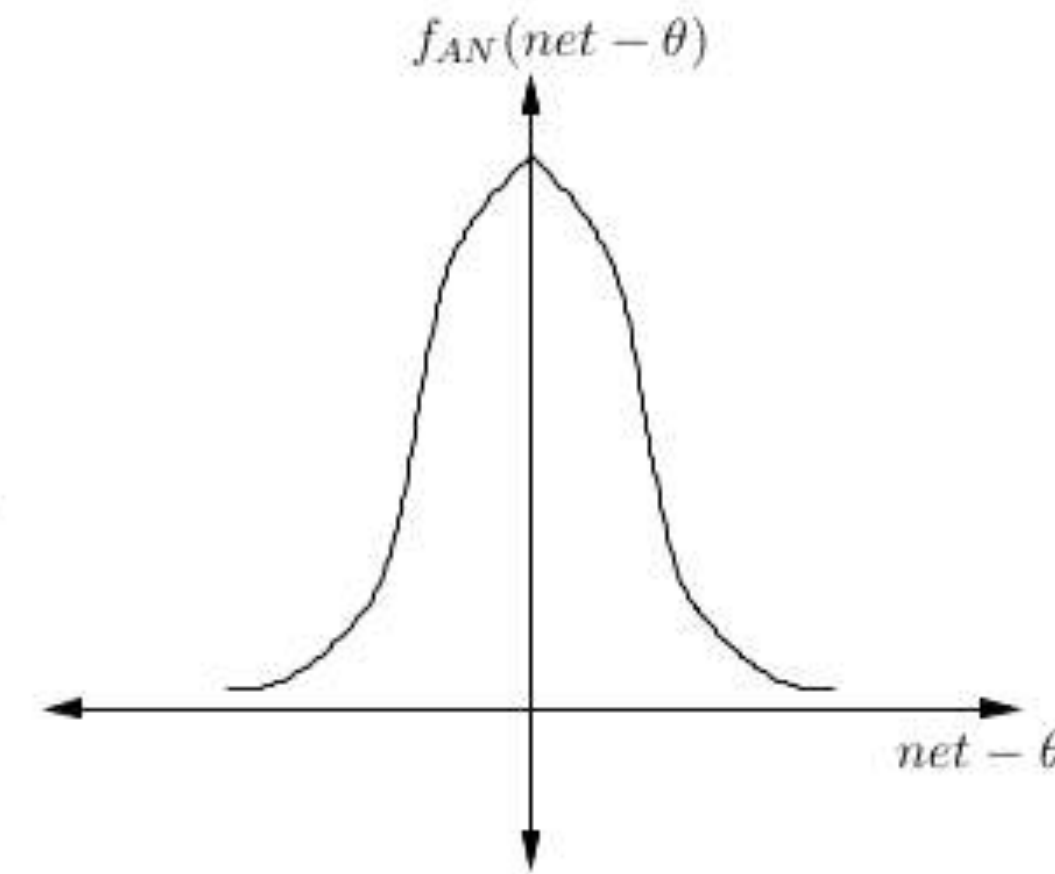
(c) Ramp function



(d) Sigmoid function



(e) Hyperbolic tangent function

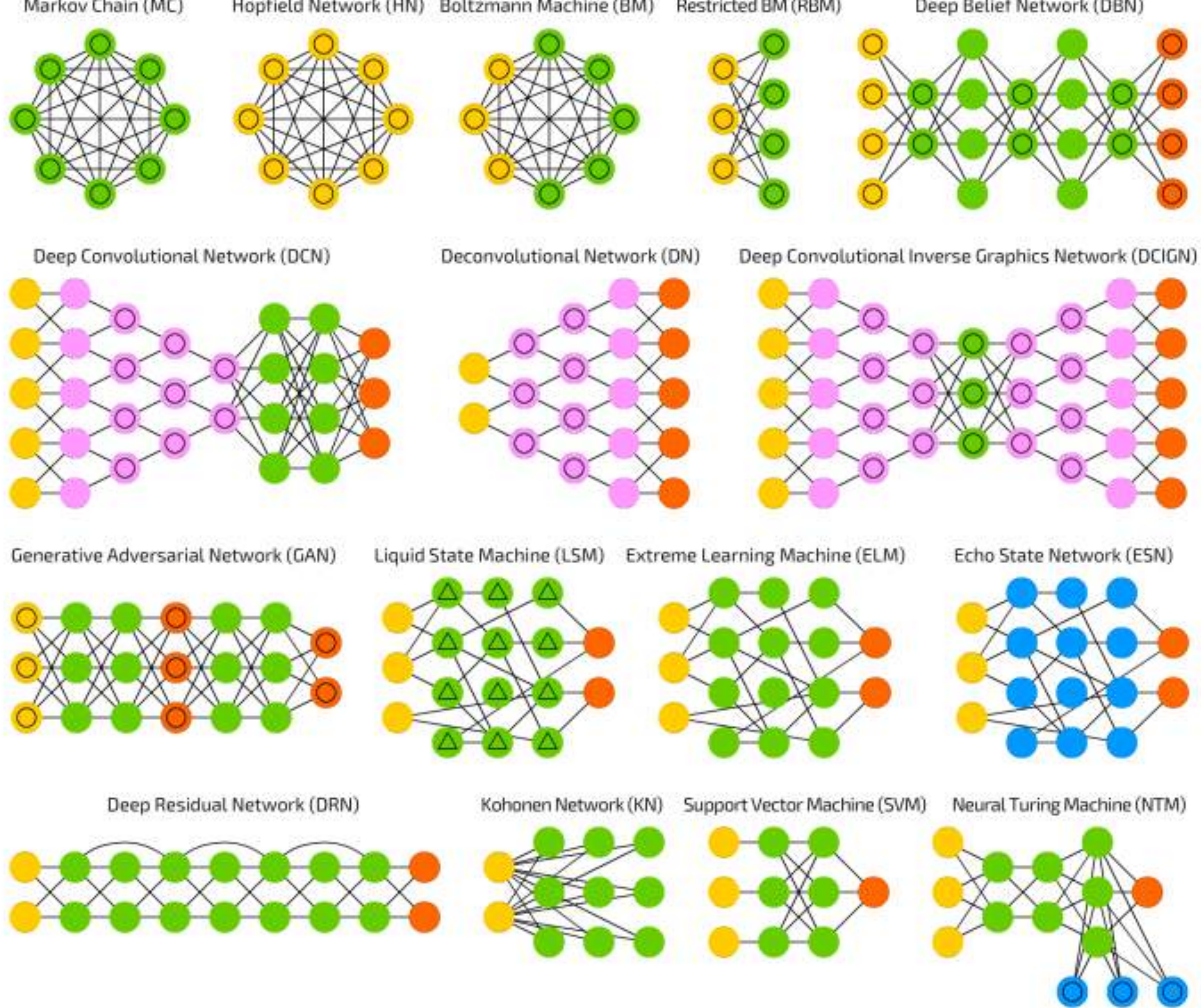


(f) Gaussian function



# Neura

- Many kinds





# What happened in the last decade...

- Algorithms: This area has seen some improvements, but most of the early wins came from fairly old ideas. Now that Deep Learning is showing success we are seeing some good advances as well.
- Datasets: Training large networks is hard without large enough datasets. MNIST can only go so far in pushing the limits. Having datasets like ImageNet has really helped pushed the state of the art in vision.
- Compute: the biggest game changer in recent years.



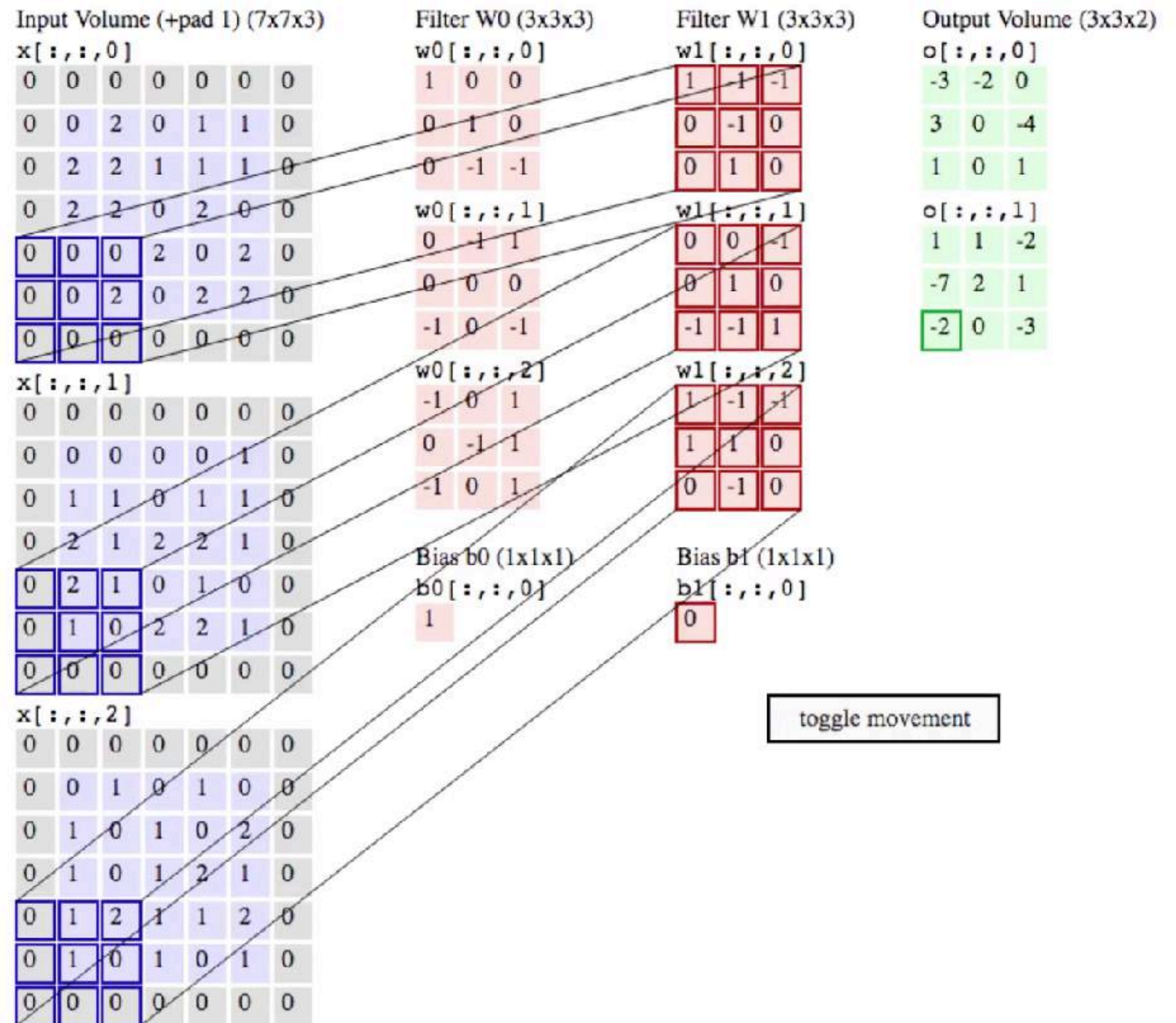


163 Zetabytes  
(that's 163 trillion gigabytes!)  
of data per year by 2025



# Convolution NNs

- Large Networks
- Lots of Input Nodes
- Lots of Layers
- Each Layer with a special task





# TensorFlow History: The Cat



It took 16,000 computers to identify a cat!

This was in 2012

# “TensorFlow”, the name...

Tensors: multidimensional data arrays

```
[ 0.3, 0.2, 0.4, 0.5, 0.9 ]
```

But also:

```
[  
  [ 0.3, 0.2, 0.4, 0.5, 0.9 ],  
  [ 0.6, 0.0, 0.8, 0.5, 0.4 ],  
  [ 0.2, 0.8, 0.8, 0.1, 0.3 ]  
]
```



# “TensorFlow”, the flow ?

- Math Operations
  - `tf.exp` `tf.pow` `tf.tan` `tf.sign`
- Control Flow
  - `tf.cond` `tf.do_while`
- Tensor Operations
  - `tf.matmul` `tf.add` `tf.reduce_sum` `tf.cumprod`

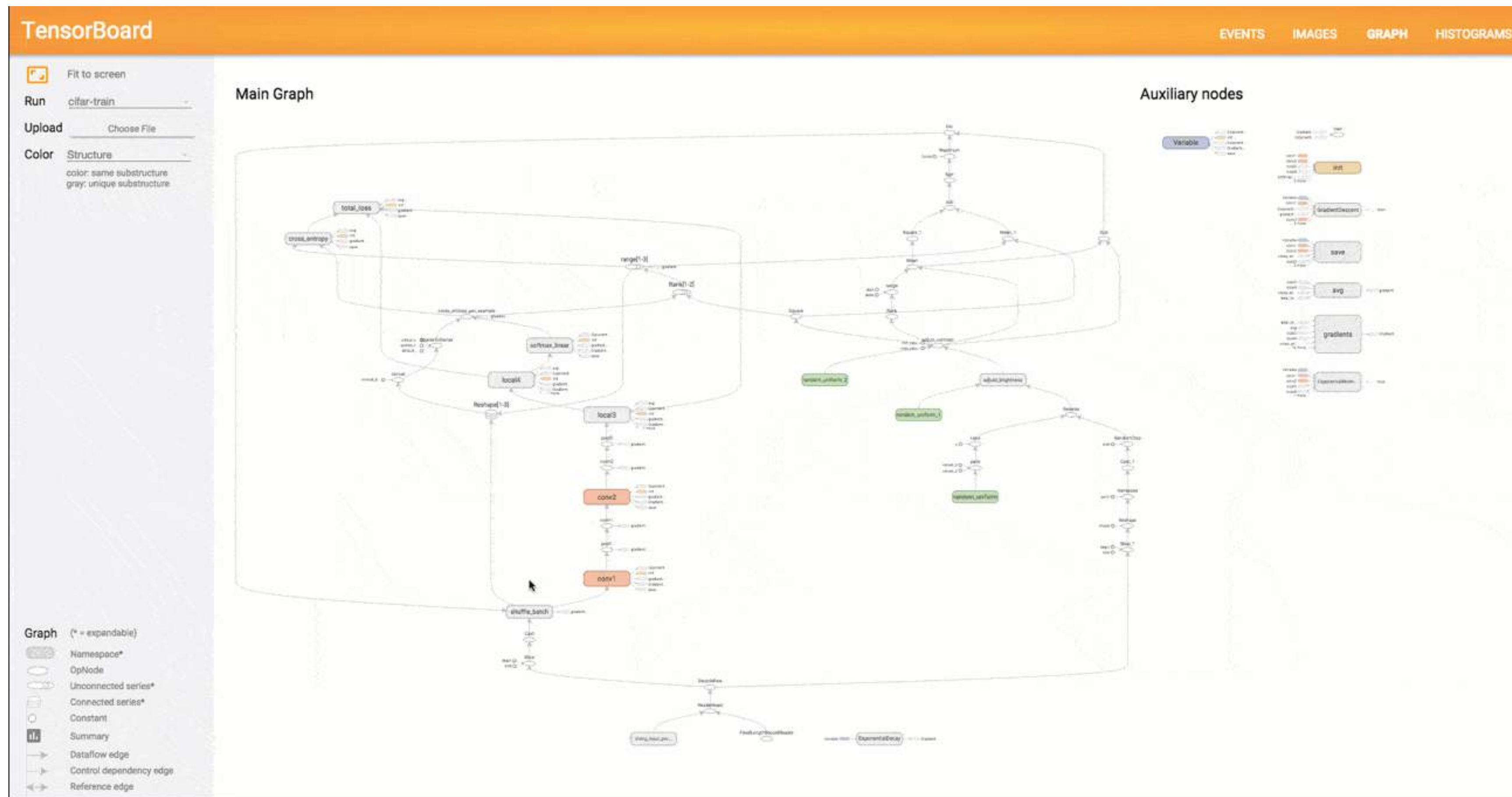
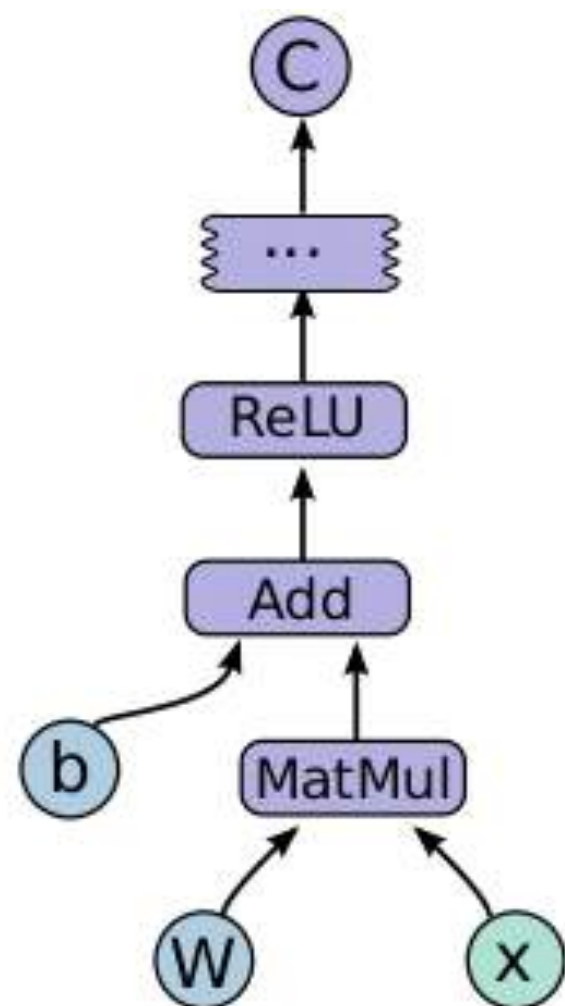
# “TensorFlow”, the Graph

Computation is defined as a directed acyclic graph (DAG) to optimize an objective function

- Graph is defined in high-level language (Python)
- Graph is compiled and optimized
- Graph is executed (in parts or fully) on available low level devices (CPU, GPU)
- Data (tensors) flow through the graph
- TensorFlow can compute gradients automatically



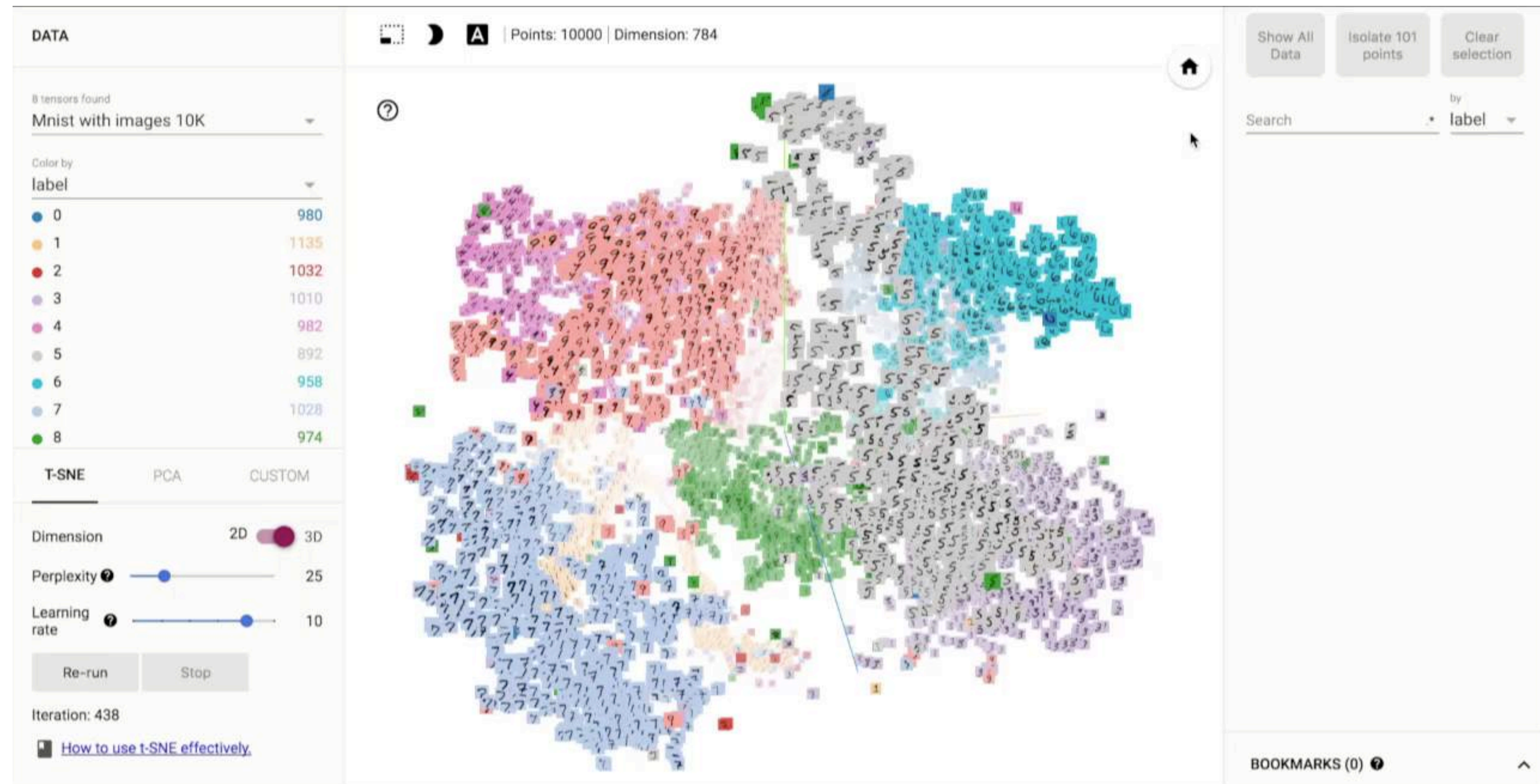
# “TensorFlow”, the Graph





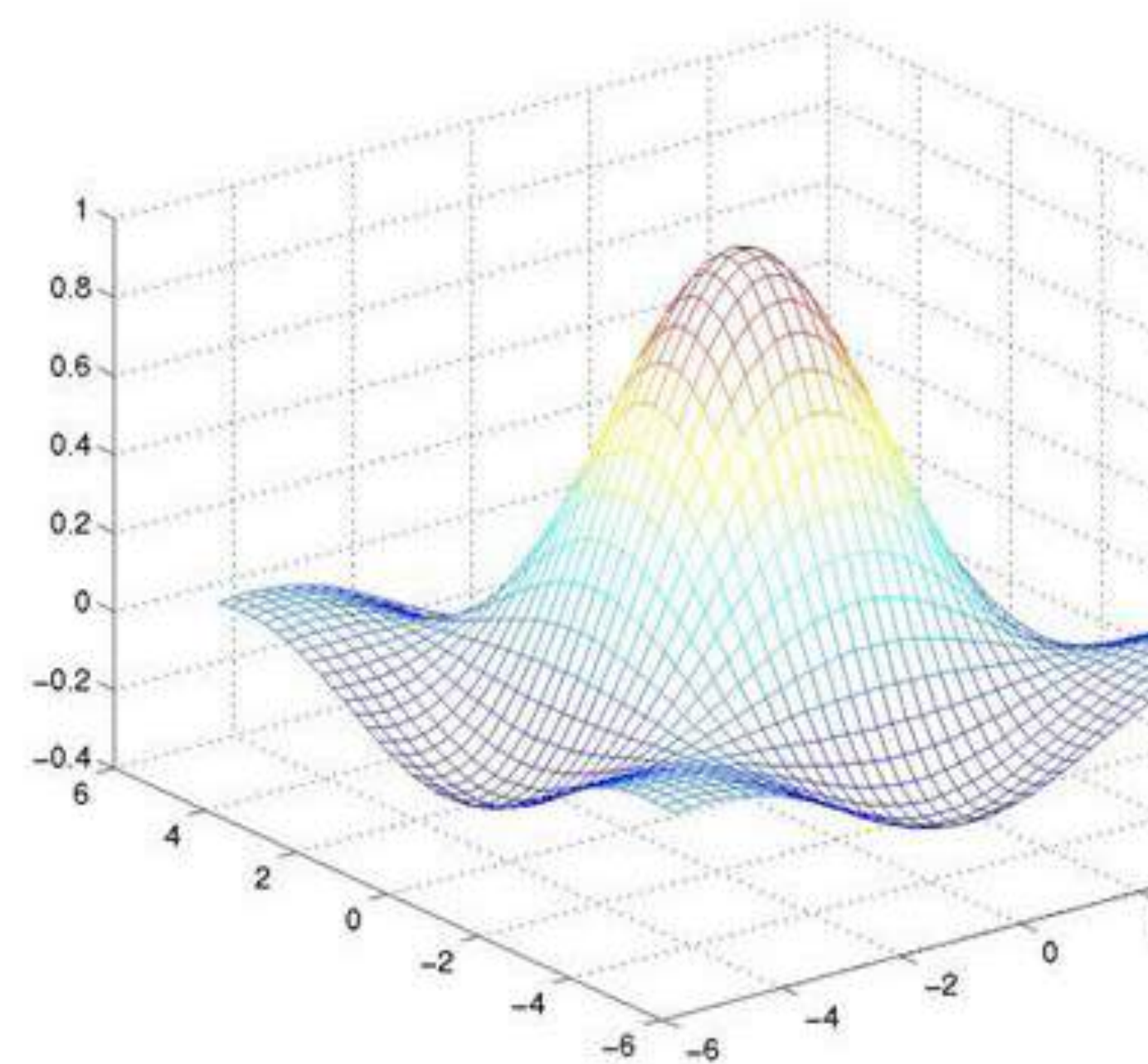
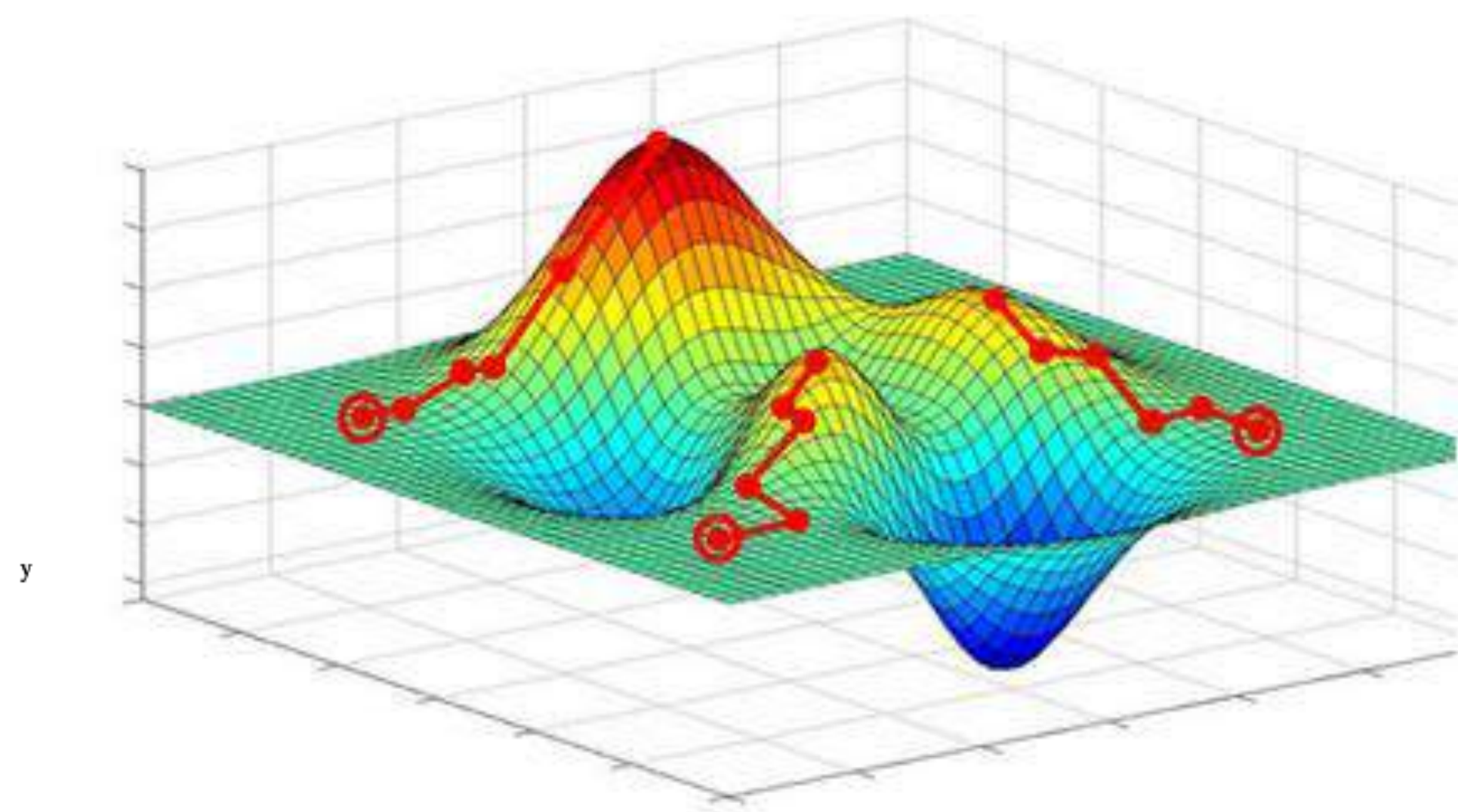
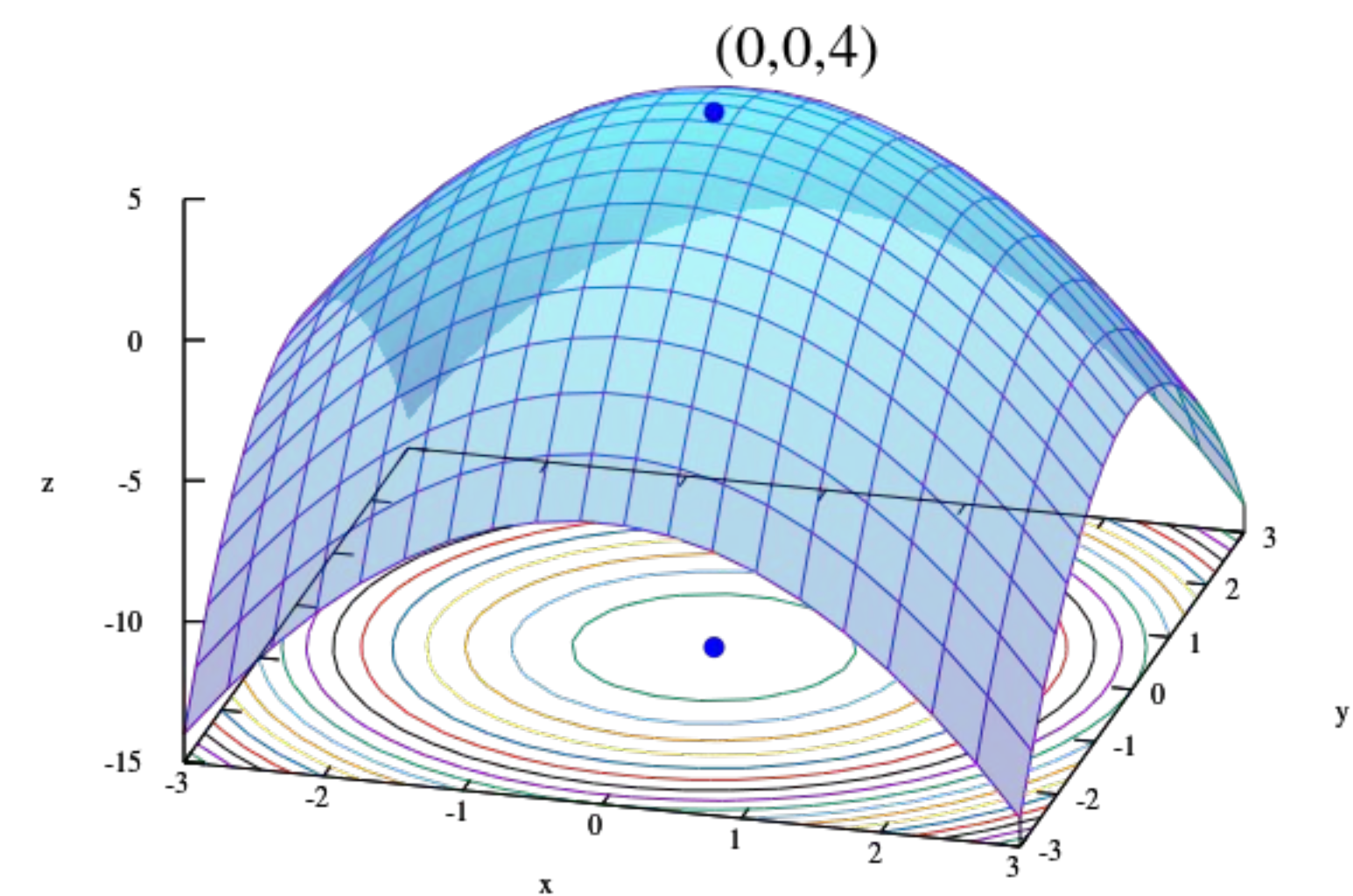
# TensorBoard Projector (new)

Recently launched  
Projector

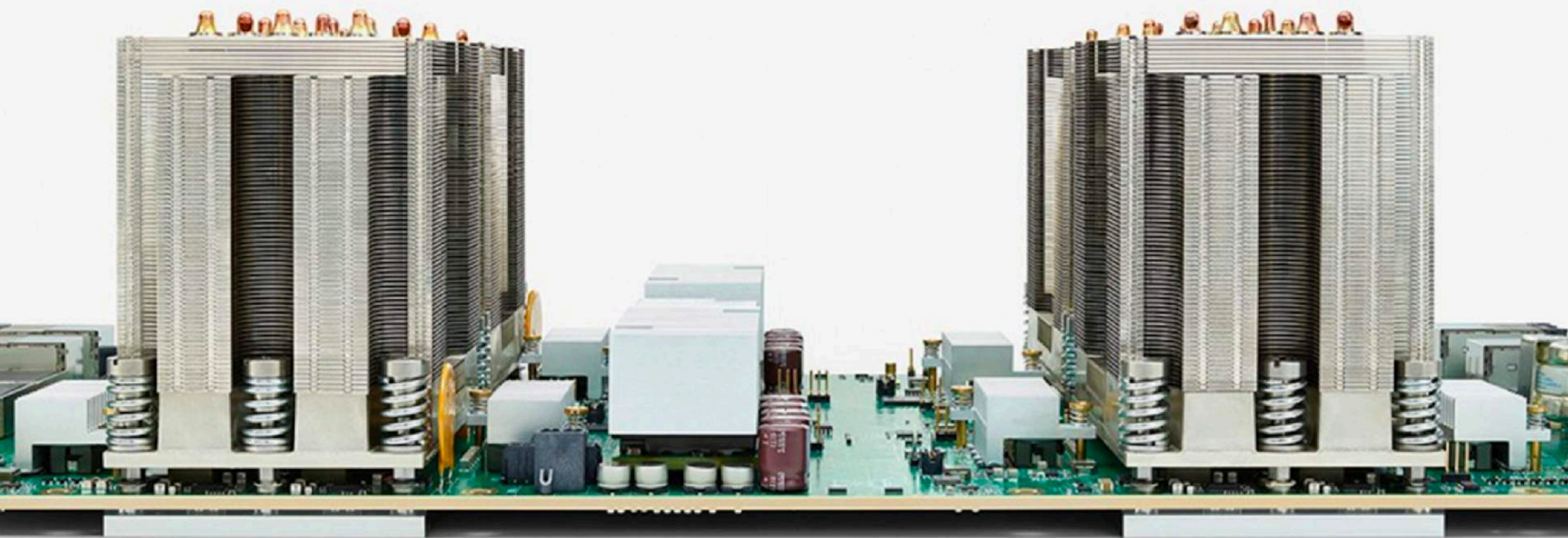




# “TensorFlow”, Gradients?











11.5 petaflops per pod





[Home](#) / [Google IO](#)

## NEWS

# Google's Tensor Processing Unit could advance Moore's Law 7 years into the future

Google unveils a custom chip, which it says advances computing performance by three generations.

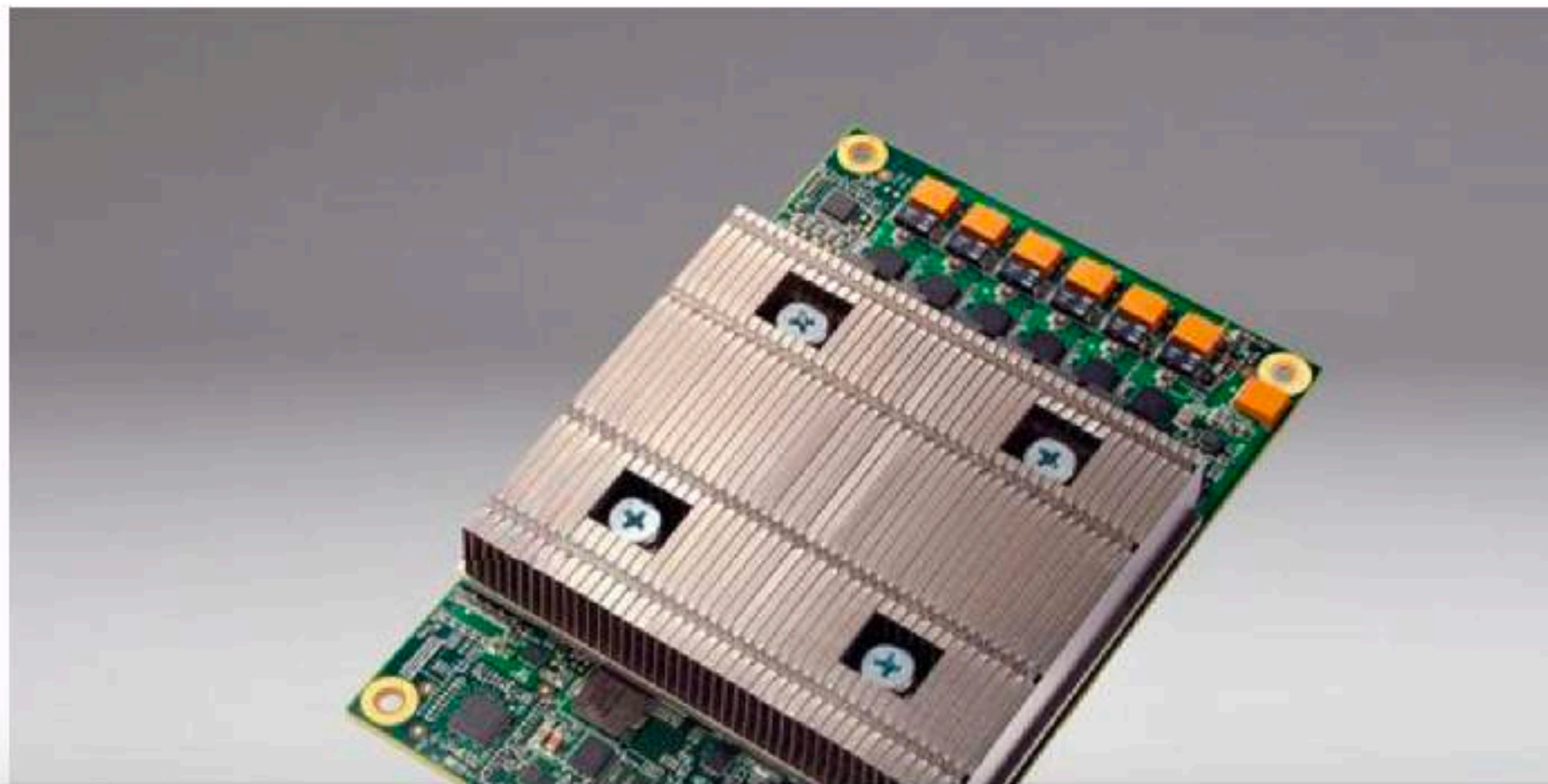
By [Gordon Mah Ung](#)

Executive Editor, PCWorld | MAY 18, 2016 2:08 PM PT



Where are you on the  
path to **cloud intelligence**?

[Find Out Now](#)



## MORE LIKE THIS



Nvidia chief downplays challenge from Google's hyper-specialized AI chip



May's mighty new PC hardware: Mammoth graphics champions and freakishly...

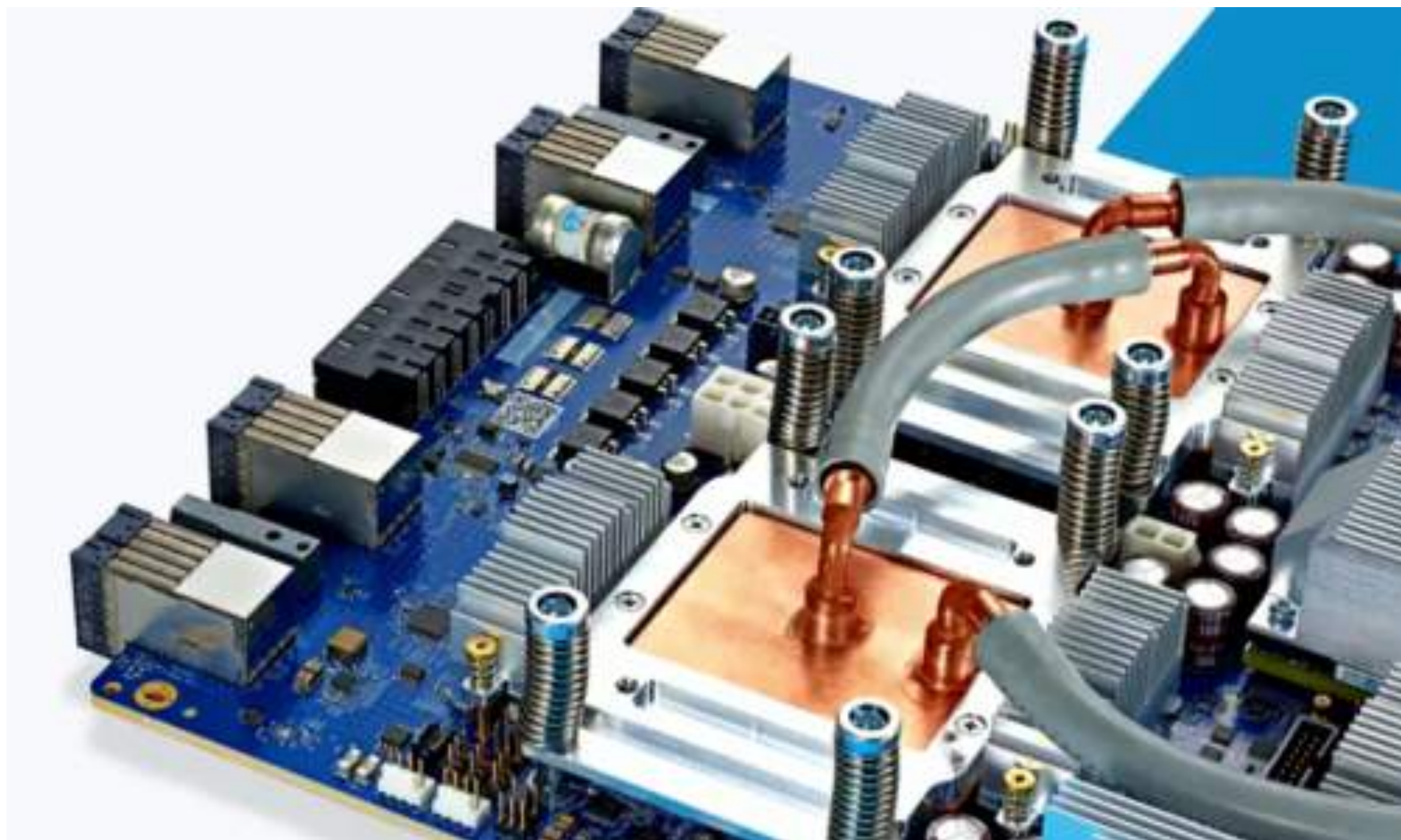


Google says its AI chips smoke CPUs, GPUs in performance tests

VIDEO



# TPU 3.0



8x faster



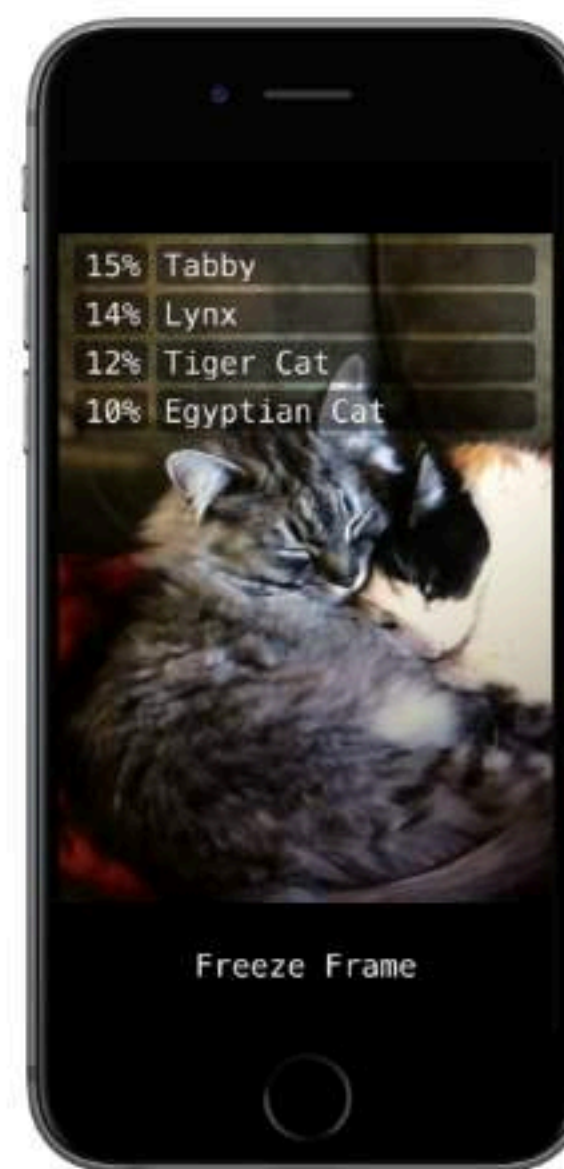
# Portable & Scalable



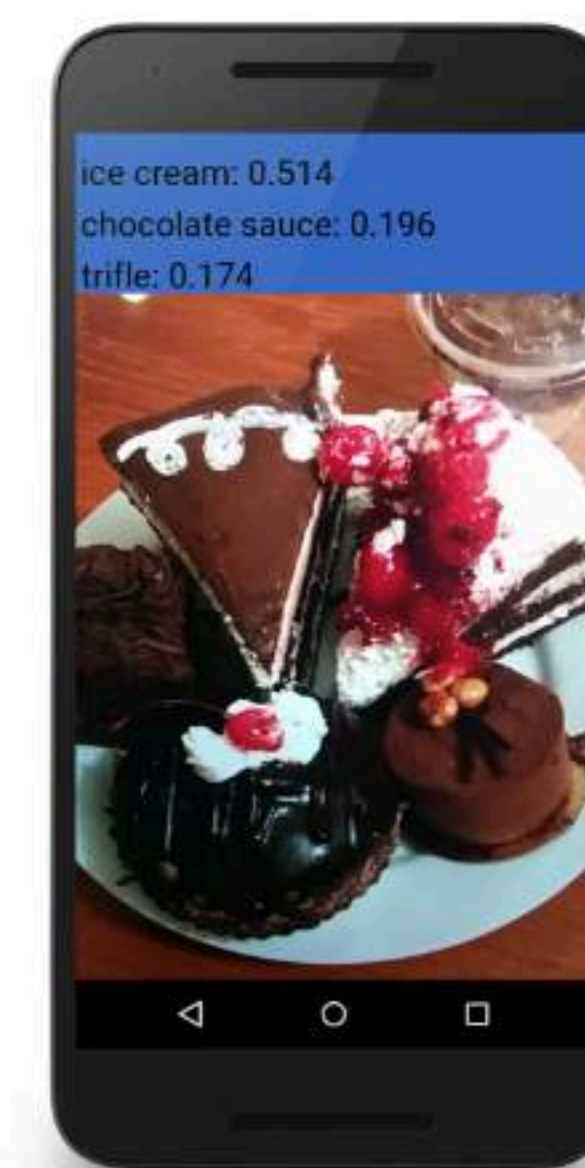
Your laptop



Datacenters



Android



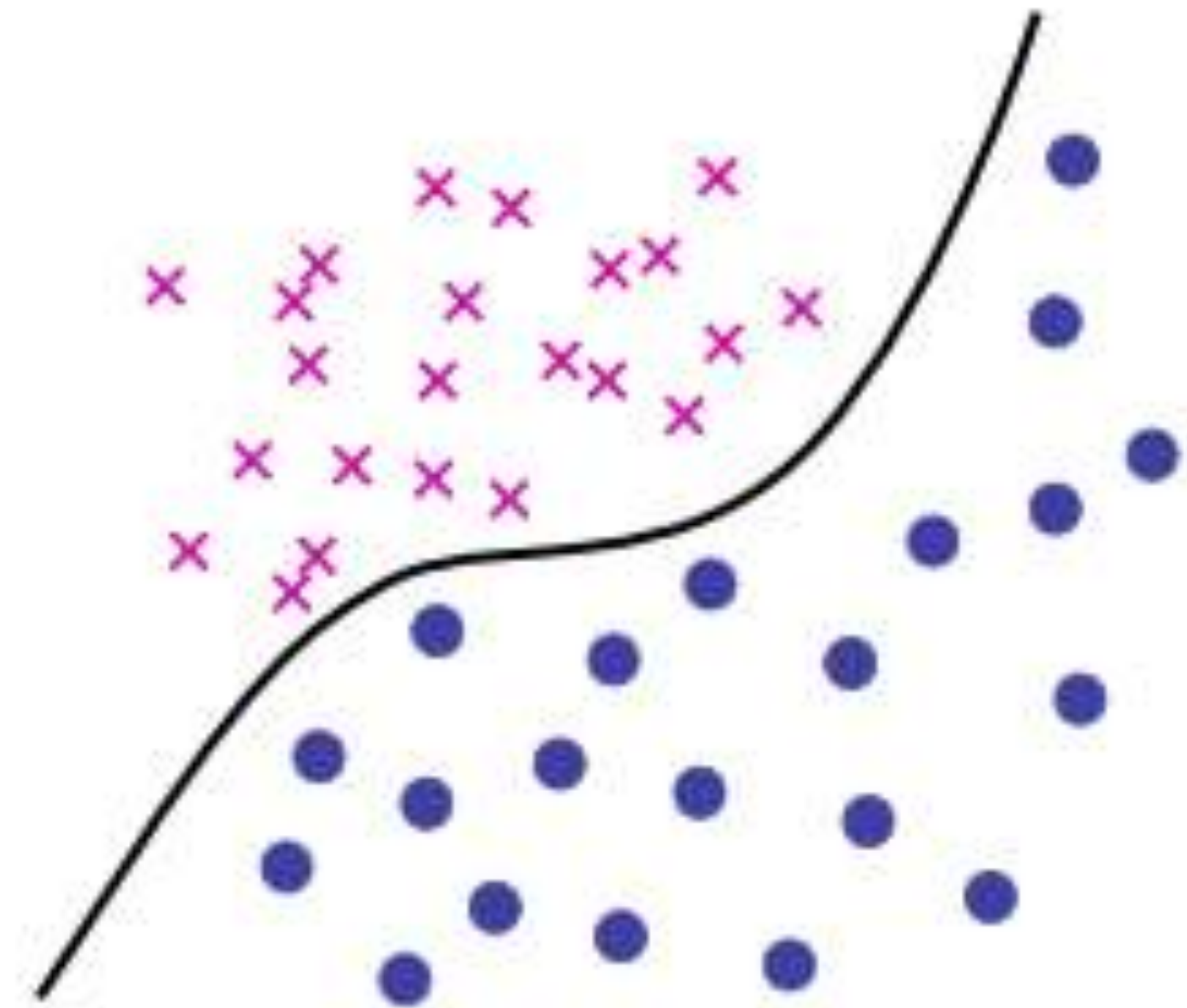
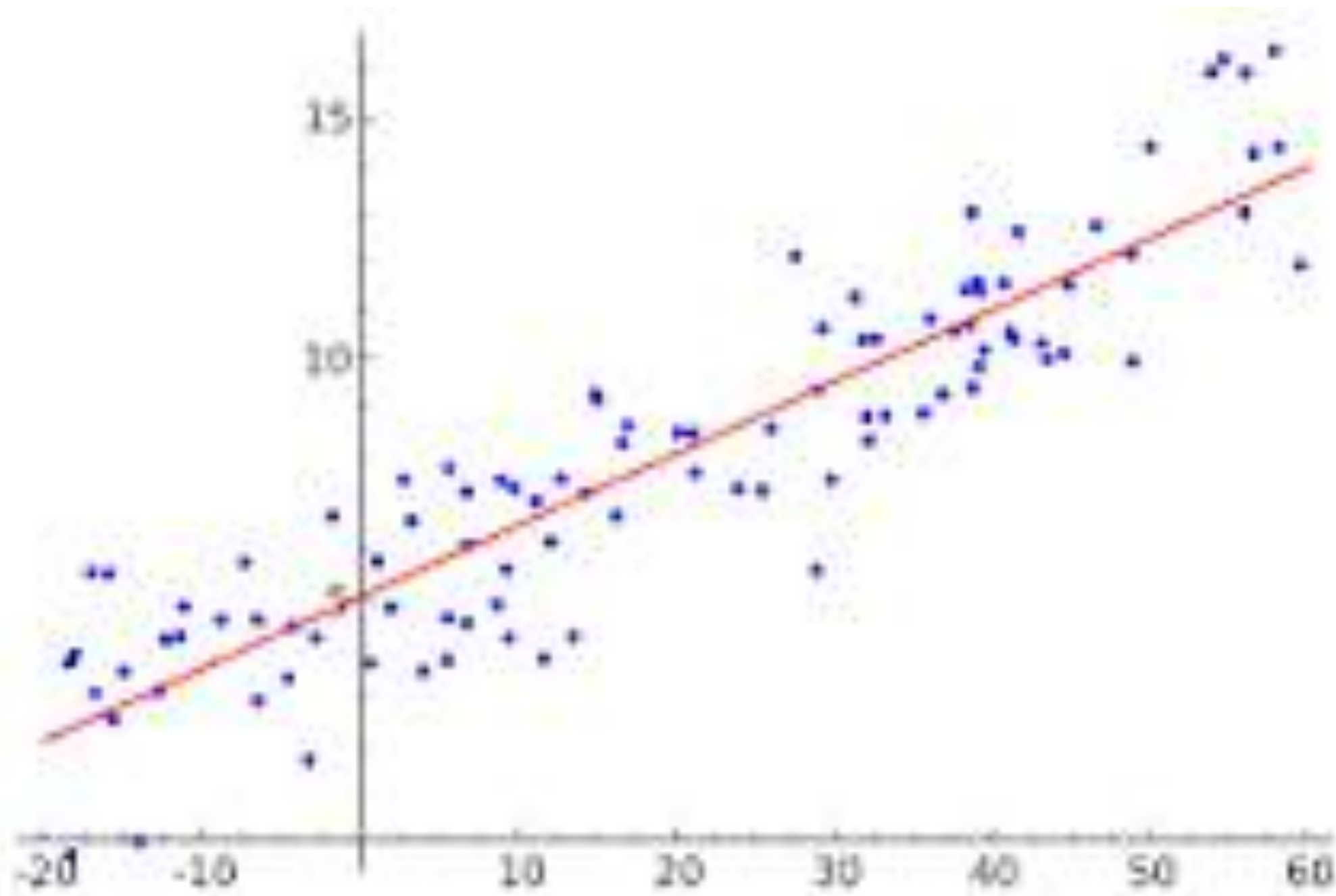
iOS



Raspberry  
Pi



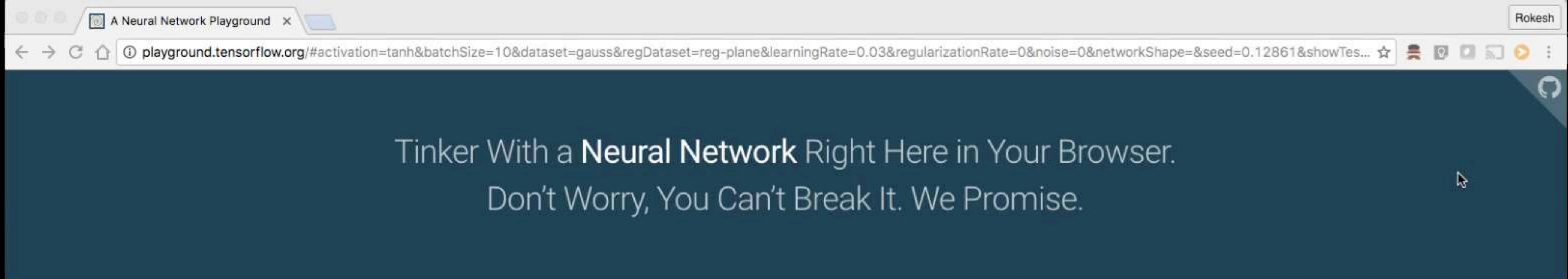
# Linear Regression & Classification





# TensorFlow Playground





Iterations  
000,000

Learning rate  
0.03

Activation  
Tanh

Regularization  
None

Regularization rate  
0

Problem type  
Classification

#### DATA

Which dataset do you want to use?



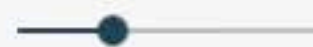
Ratio of training to test data: 50%



Noise: 0



Batch size: 10



REGENERATE

#### FEATURES

Which properties do you want to feed in?

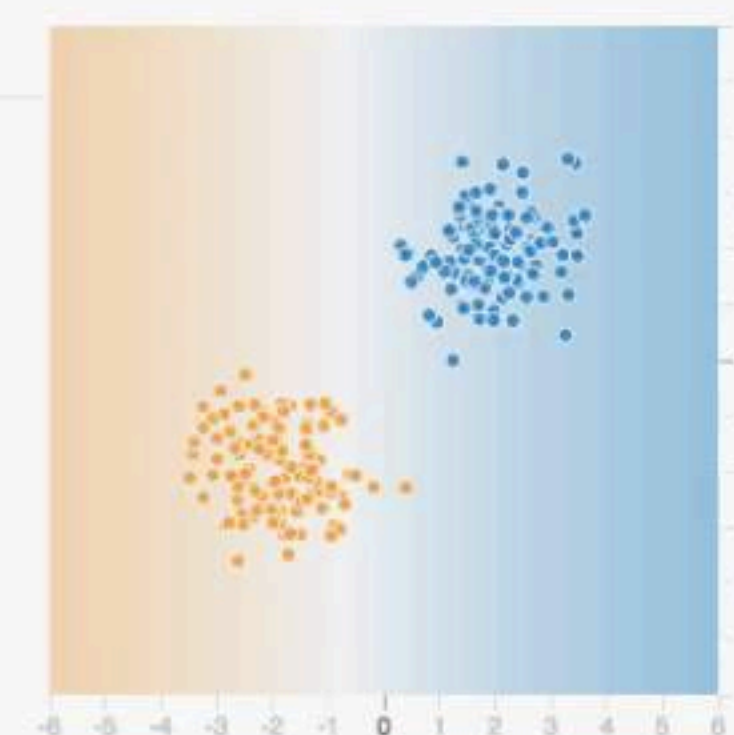


+ - 0 HIDDEN LAYERS

#### OUTPUT

Test loss 0.300

Training loss 0.291



Colors shows data, neuron and weight values.



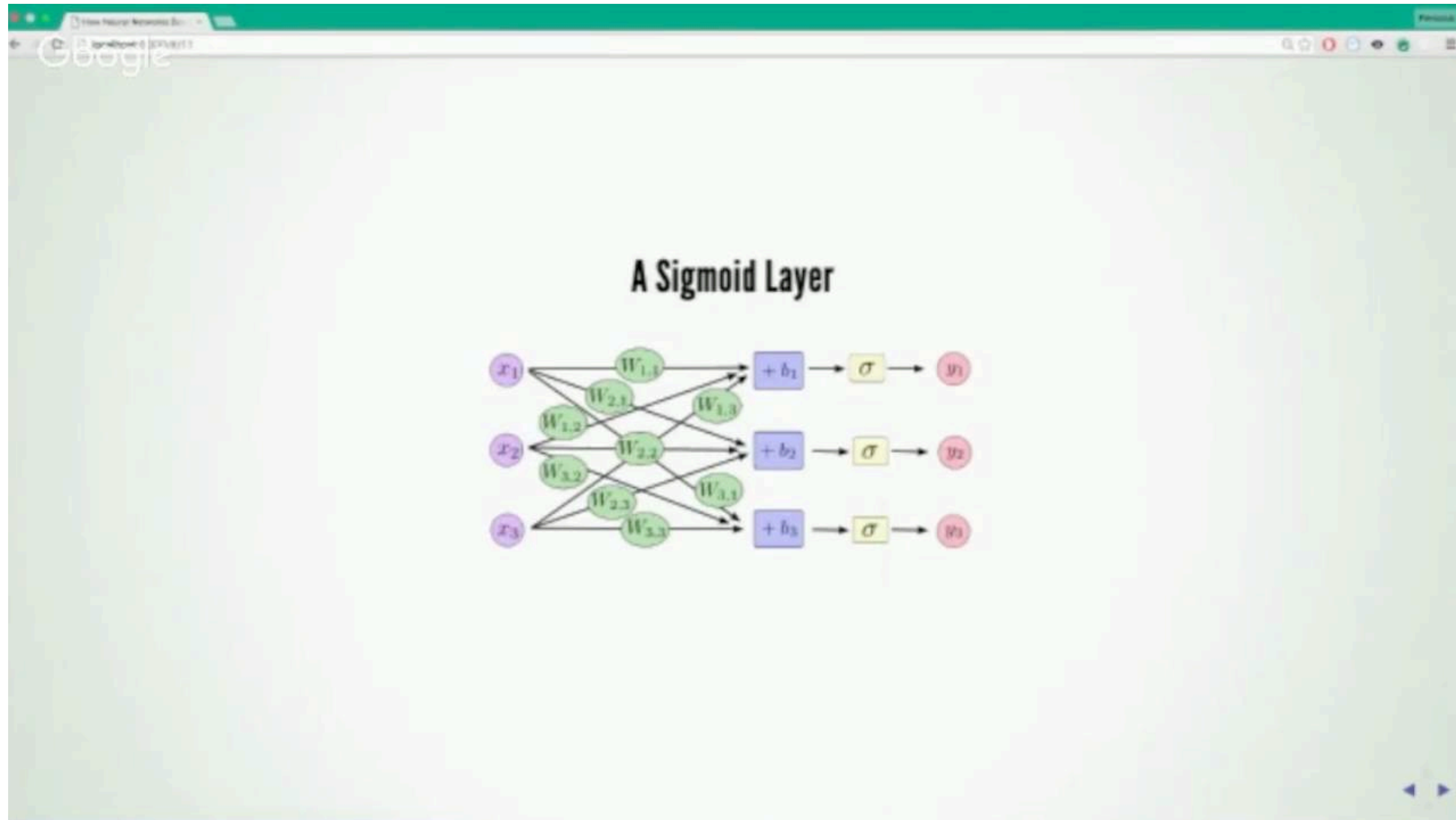
☐ Show test data

☐ Discretize output





# what just happened in Linear Regression ?





# Writing your Applications: Where to start ?

- The easiest way to go (nowadays) is Docker image
- When you start the docker image immediately a iPython Notebook is available
- Start playing around with notebooks! (we can do it together!)



# Code Sample 1: Simple

```
1  import tensorflow as tf
2  import numpy as np
3
4  x = tf.constant([[1, 2]])
5  neg_x = tf.neg(x)
6
7  print(neg_x)
8
9  with tf.Session() as sess:
10     result = sess.run(neg_x)
11  print(result)
```



# Code Sample 2: Which processing unit?

```
1  import tensorflow as tf
2
3  # Creates a graph.
4  a = tf.constant([1.0, 2.0, 3.0, 4.0, 5.0, 6.0], shape=[2, 3], name='a')
5  b = tf.constant([1.0, 2.0, 3.0, 4.0, 5.0, 6.0], shape=[3, 2], name='b')
6  c = tf.matmul(a, b)
7  # Creates a session with log_device_placement set to True.
8  sess = tf.Session(config=tf.ConfigProto(log_device_placement=True))
9  # Runs the op.
10 print sess.run(c)
11
```



# Code Sample 2 : output

```
root@f5662678fbad:~# python app1.py
Device mapping: no known devices.
I tensorflow/core/common_runtime/direct_session.cc:149] Device mapping:

b: /job:localhost/replica:0/task:0/cpu:0
I tensorflow/core/common_runtime/simple_placer.cc:388] b: /job:localhost/replica:0/task:0/cpu:0
a: /job:localhost/replica:0/task:0/cpu:0
I tensorflow/core/common_runtime/simple_placer.cc:388] a: /job:localhost/replica:0/task:0/cpu:0
MatMul: /job:localhost/replica:0/task:0/cpu:0
I tensorflow/core/common_runtime/simple_placer.cc:388] MatMul: /job:localhost/replica:0/task:0/cpu:0
[[ 22.  28.]
 [ 49.  64.]]
```



# Code Sample 2 : result

```
1  import tensorflow as tf
2
3  # Creates a graph.
4  with tf.device('/cpu:0'):
5      a = tf.constant([1.0, 2.0, 3.0, 4.0, 5.0, 6.0], shape=[2, 3], name='a')
6      b = tf.constant([1.0, 2.0, 3.0, 4.0, 5.0, 6.0], shape=[3, 2], name='b')
7      c = tf.matmul(a, b)
8  # Creates a session with log_device_placement set to True.
9  sess = tf.Session(config=tf.ConfigProto(log_device_placement=True))
10 # Runs the op.
11 print sess.run(c)
```





```
1  import tensorflow as tf
2  x = tf.placeholder(tf.float32, [None, 784])
3
4  #layer 1
5  W1 = tf.Variable(tf.zeros([784, 100]))
6  b1 = tf.Variable(tf.zeros([100]))
7  y1 = tf.matmul(x, W1) + b1 softmax
8
9  #layer 2
10 W2 = tf.Variable(tf.zeros([100, 10]))
11 b2 = tf.Variable(tf.zeros([10]))
12 y2 = tf.nn.softmax(tf.matmul(y1, W2) + b2)
13
14 #output
15 y = y2
16 y_ = tf.placeholder(tf.float32, [None, 10])
17
```



# What is the process ?

## TensorFlow Code for Researchers

```
import tensorflow as tf
```

initialization

```
X = tf.placeholder(tf.float32, [None, 28, 28])  
W = tf.Variable(tf.zeros([784, 10]))  
b = tf.Variable(tf.zeros([10]))  
init = tf.initialize_all_variables()
```

```
# model  
Y = tf.nn.softmax(tf.matmul(tf.reshape(X, [-1, 784]), W) + b)
```

```
# placeholder for correct answers  
Y_ = tf.placeholder(tf.float32, [None, 10])
```

```
# loss function  
cross_entropy = -tf.reduce_sum(Y_ * tf.log(Y))
```

```
# % of correct answers found in batch  
is_correct = tf.equal(tf.argmax(Y, 1), tf.argmax(Y_, 1))  
accuracy = tf.reduce_mean(tf.cast(is_correct, tf.float32))
```

loss

inference

training step

```
optimizer = tf.train.GradientDescentOptimizer(0.003)  
train_step = optimizer.minimize(cross_entropy)
```

```
sess = tf.Session()  
sess.run(init)
```

```
for i in range(1000):  
    # load batch of images and correct answers  
    batch_X, batch_Y = mnist.train.next_batch(100)  
    train_data = {X: batch_X, Y_: batch_Y}
```

```
    # train  
    sess.run(train_step, feed_dict=train_data)
```

```
    # success ? add code to print it  
    a, c = sess.run([accuracy, cross_entropy],  
                    feed_dict=train_data)
```

```
    # success on test data ?  
    test_data = {X: mnist.test.images, Y_: mnist.test.labels}  
    a, c = sess.run([accuracy, cross_entropy],  
                    feed_dict=test_data)
```

Run



# Examples

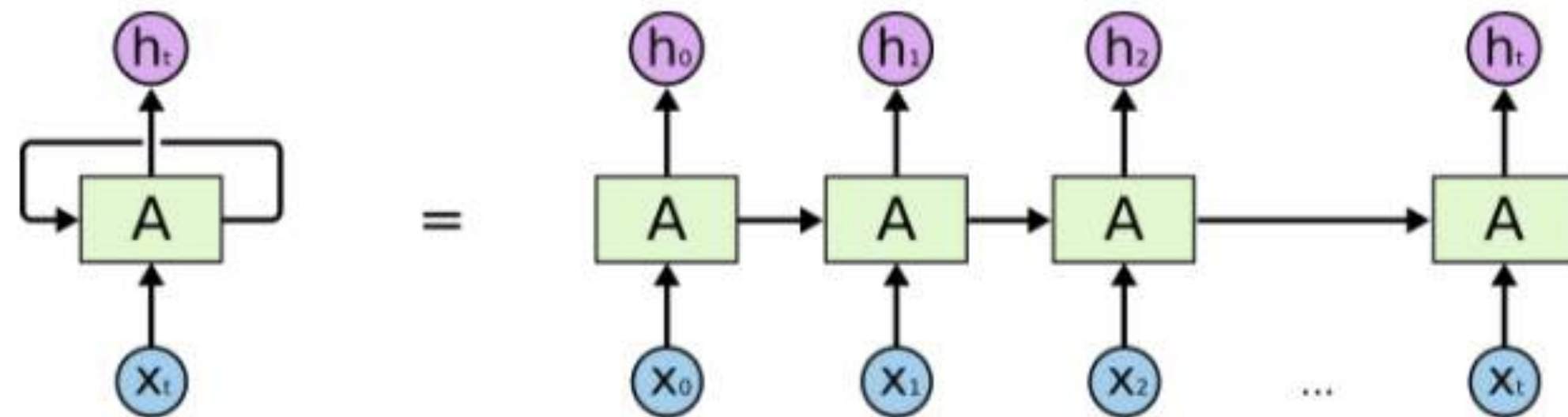
- <https://github.com/tensorflow/models/tree/master/research/im2txt>
- “Lip reading”
- Human-like sound: [WaveNet](#)
- [Skin cancer detection](#)
- [AI Experiments by Google](#)
- [Learn how to fly](#)





# LSTM Networks

- Special Kind of Recurrent Neural Network: Long Short Term Network



An unrolled recurrent neural network.

- “Humans don’t start their thinking from scratch every second. As you read this essay, you understand each word based on your understanding of previous words. You don’t throw everything away and start thinking from scratch again. Your thoughts have persistence”
- *Best example: Google Translate*



# Companies using TensorFlow





# Machine Learning Use Cases

## Manufacturing

- Predictive maintenance or condition monitoring
- Warranty reserve estimation
- Propensity to buy
- Demand forecasting
- Process optimization
- Telematics

## Travel and Hospitality

- Aircraft scheduling
- Dynamic pricing
- Social media – consumer feedback and interaction analysis
- Customer complaint resolution
- Traffic patterns and congestion management

## Retail

- Predictive inventory planning
- Recommendation engines
- Upsell and cross-channel marketing
- Market segmentation and targeting
- Customer ROI and lifetime value

## Financial Services

- Risk analytics and regulation
- Customer Segmentation
- Cross-selling and up-selling
- Sales and marketing campaign management
- Credit worthiness evaluation

## Healthcare and Life Sciences

- Alerts and diagnostics from real-time patient data
- Disease identification and risk satisfaction
- Patient triage optimization
- Proactive health management
- Healthcare provider sentiment analysis

## Energy, Feedstock and Utilities

- Power usage analytics
- Seismic data processing
- Carbon emissions and trading
- Customer-specific pricing
- Smart grid management
- Energy demand and supply optimization





# Google Cloud ML Engine

Announced in October 2016 at Google Horizon Event



ChromeFileEditViewHistoryBookmarksPeopleWindowHelp

Cloud Vision ExplorerKonpekiKonpekiGoogle Cloud DataLabStageDemo

localhost:8081/notebooks/StageDemo.ipynb


Google Cloud DatalabStageDemo (autosaved)

NotebookAdd CodeAdd MarkdownDeleteMove UpMove DownRunClearReset SessionNavigationHelp

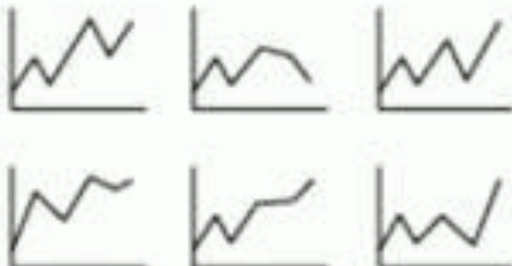
### Hypothesis

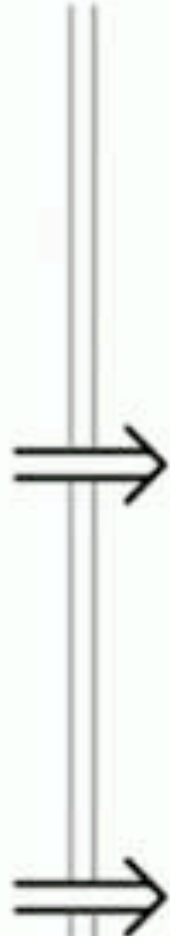
If we take 5+ decades of S&P data, and we take 5+ decades of data for 3,200 individual stocks, can we predict if the S&P will close up or down tomorrow?

S&P

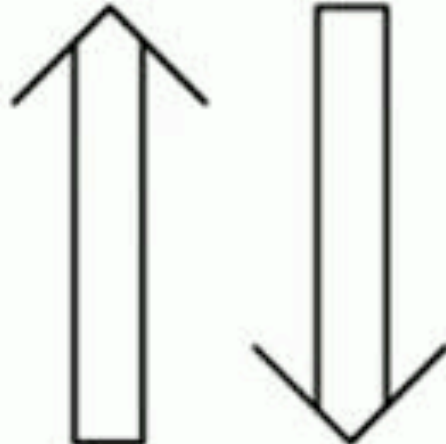


STOCKS





S&P





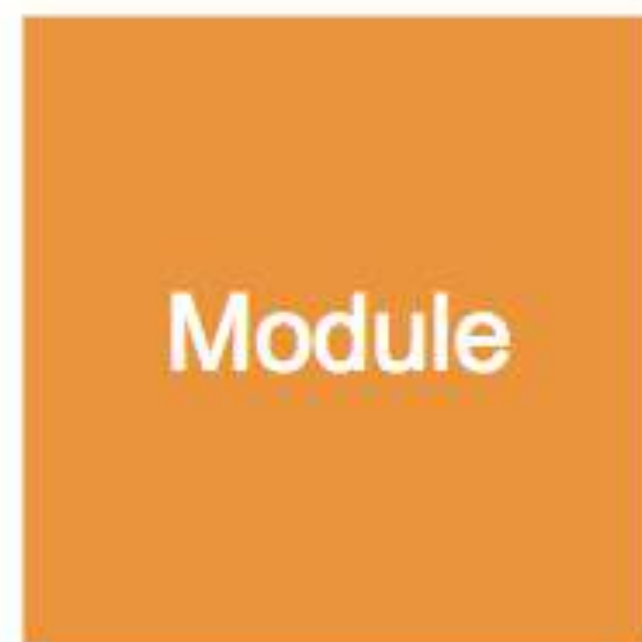
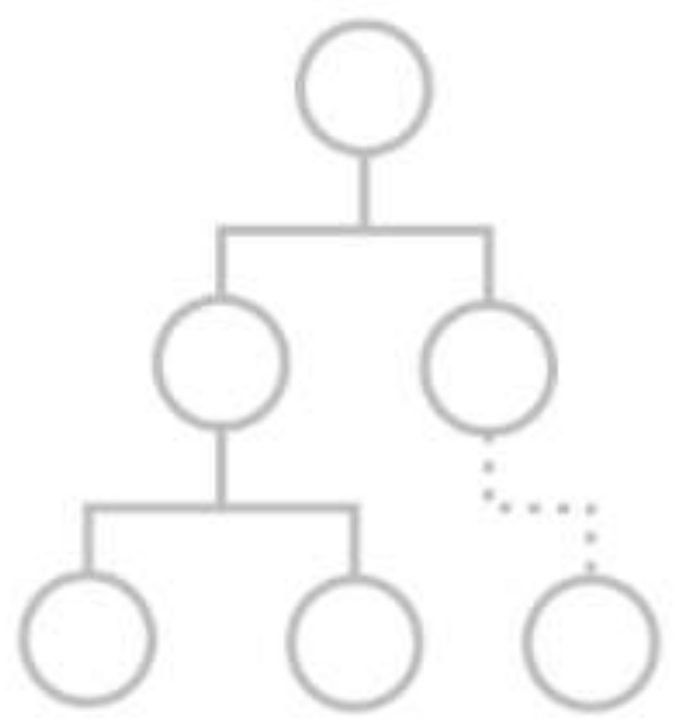
# Recent developments



# TensorFlow :Keras & Estimators

- A more abstract layer on top of TensorFlow.
- Makes code more readable.
- <https://blog.keras.io/keras-as-a-simplified-interface-to-tensorflow-tutorial.html>
- TensorFlow High-level APIs







Pre-made Estimators

Estimator

Keras  
Model

Layers

Datasets

Python Frontend

C++

Java

Go

...

TensorFlow Distributed Execution Engine

CPU

GPU

TPU

Android

XLA

iOS

...



Keras Model

TensorFlow  
SavedModel

Layers API

Ops API (Eager)

Browser

WebGL

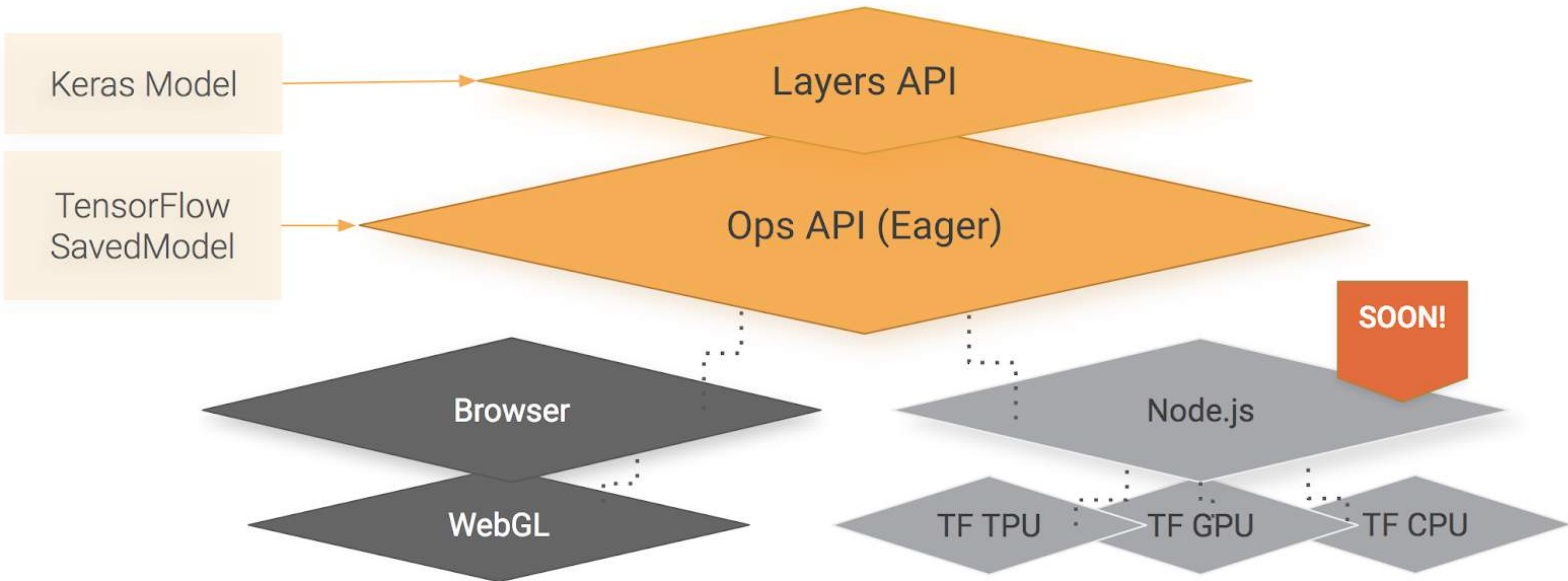
Node.js

SOON!

TF TPU

TF GPU

TF CPU





# Transfer Learning

- *Transfer learning is a machine learning method which utilizes a pre-trained neural network. For example, the image recognition model called Inception-v3 consists of two parts:*
  - *Feature extraction part with a convolutional neural network.*
  - *Classification part with fully-connected and softmax layers.*
- *In transfer learning, when you build a new model to classify your original dataset, you reuse the feature extraction part and re-train the classification part with your dataset. Since you don't have to train the feature extraction part (which is the most complex part of the model), you can train the model with less computational resources and training time.*



# GAN: Generative Adversarial Networks

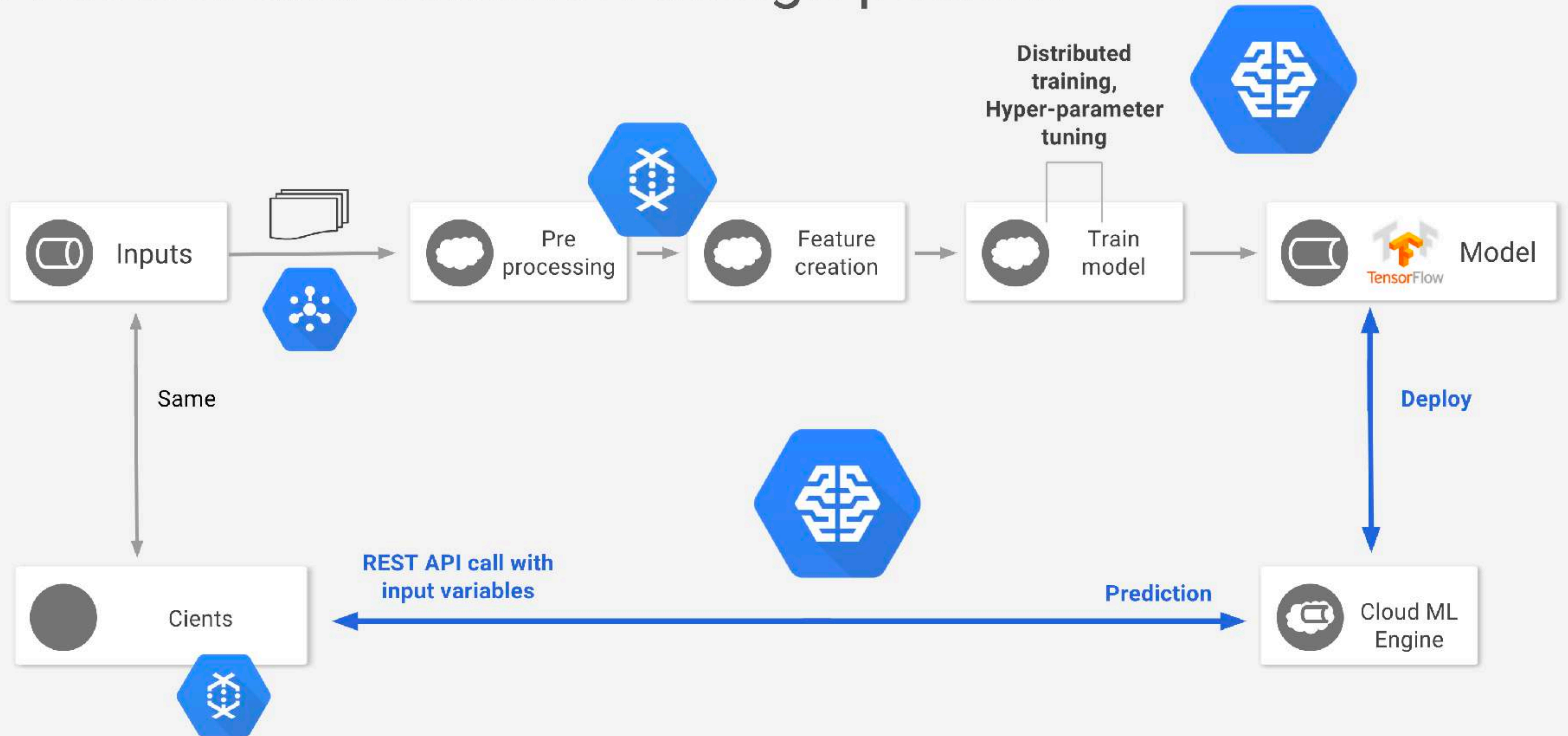
- <http://blog.aylien.com/introduction-generative-adversarial-networks-code-tensorflow/>
- <http://blog.evjang.com/2016/06/generative-adversarial-nets-in.html>
- <https://bamos.github.io/2016/08/09/deep-completion/>
- <https://arxiv.org/pdf/1701.00160.pdf>
- A new area of research where “generation of output” is done (here’s a [popular one](#))



Using in Production on ML Engine



# A useful and common design pattern





# Manage your data journey seamlessly



Collect



Transform



Analyze



Visualize



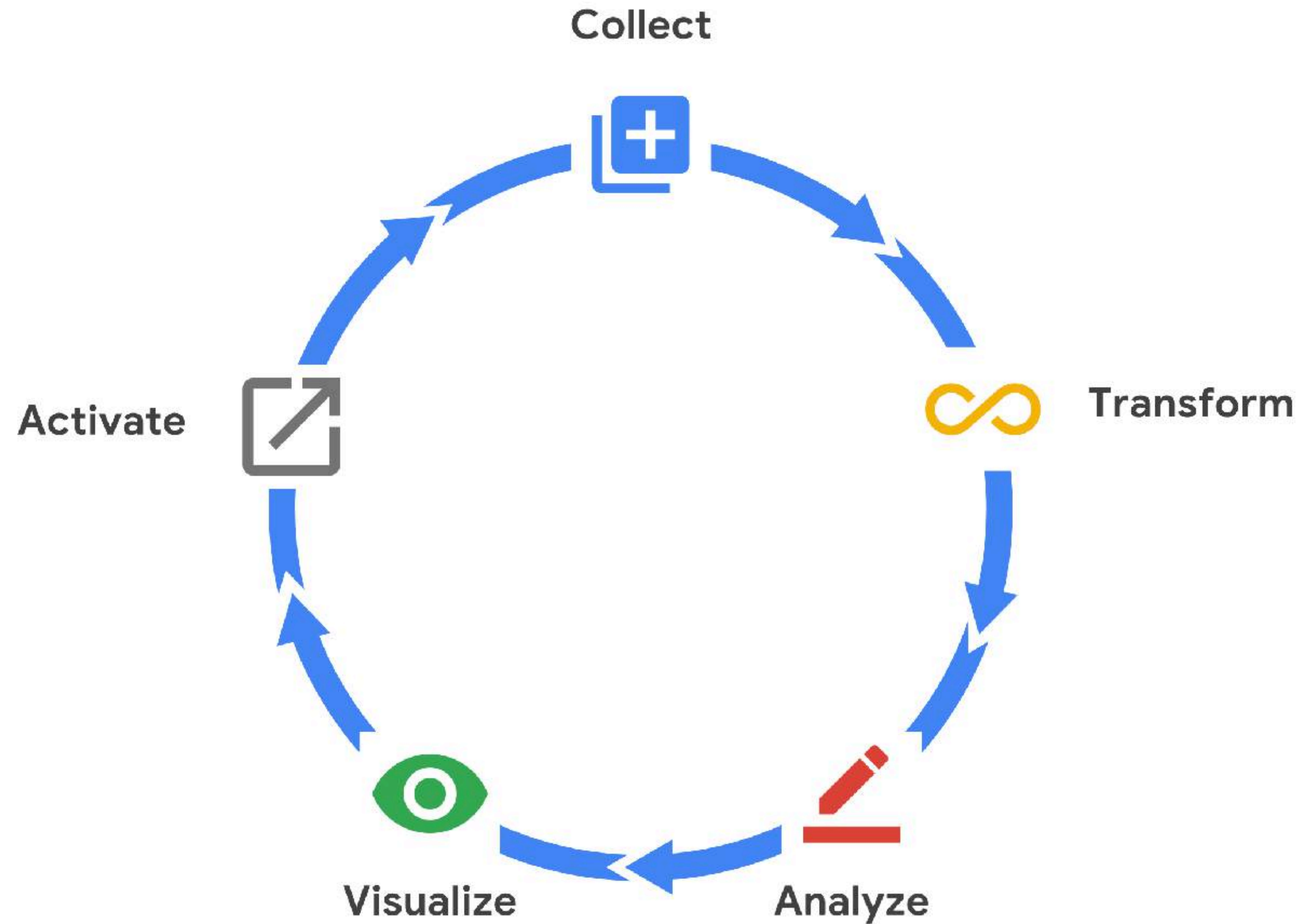
Activate





# Our vision

- **More** data sets
- **More** ML models
- **More** activation connectors





Possibilities...many, but I want to  
highlight one specifically





**BBC**

News Sport Weather iPlayer TV Radio More

Search

**NEWS** Find local news

Home UK World Business Politics Tech Science Health Education Entertainment & Arts More

Technology

## Google achieves AI 'breakthrough' at Go

An artificial intelligence program developed by Google beats Europe's top player at the ancient Chinese game of Go, about a decade earlier than expected.

🕒 27 January 2016 | Technology

- ▶ How did they do it?
- ▶ What is the game Go?

Facebook trains AI to beat humans at Go

A close-up photograph of a person's hand placing a black Go stone on a wooden board. The board is filled with many black and white stones in various positions. A clock is visible in the background.

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

By Mike Murphy | January 27, 2016

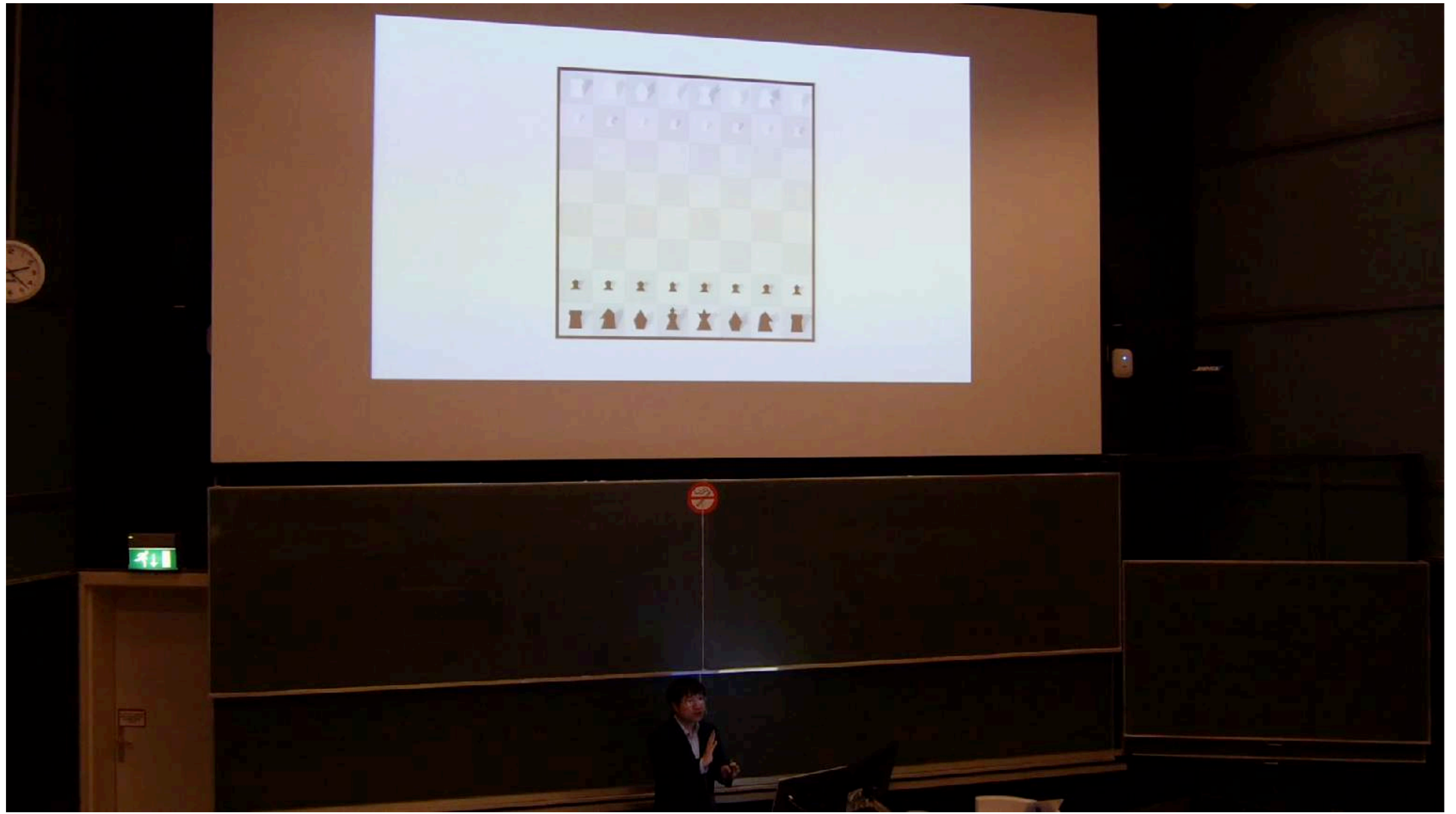
A photograph showing two people, a man and a woman, looking at a large Go board. The board is covered with many white and black stones. The man is pointing at a stone on the board.

Going up. (Reuters/Kiyoshi Ota)

Computers have slowly started to encroach on activities we previously believed only the brilliantly sophisticated human brain could handle. IBM's Deep Blue supercomputer beat Grand Master Garry Kasparov at chess in 1997, and in 2011 IBM's Watson beat former human winners at the quiz game *Jeopardy*. But the ancient board game Go has long been 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. Until now, that is.



# Solution space





“The possibilities and power are innumerable”

– Aja Huang



# After beating the world's elite Go players, Google's AlphaGo AI is retiring

Posted May 27, 2017 by [Jon Russell \(@jonrussell\)](#)



Next Story



Google's AlphaGo — the AI developed to tackle the world's most demanding strategy game — is stepping down from competitive matches after defeating the world's best talent. The latest to succumb is Go's top-ranked player, Ke Jie, who lost 3-0 in [a series hosted in China](#)



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# ALPHA GO

## AlphaGo

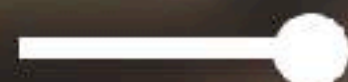
2017 |  undefined | 1h 30m

Seemingly simple but deceptively complex, the game of "Go" serves as the backdrop for this battle between artificial intelligence and man.

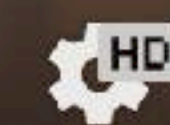
Genres: Documentaries, Science & Technology Documentaries  
Director: Greg Kohs



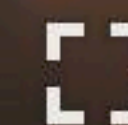
# AlphaGo Zero

The background of the video player is a close-up, shallow depth-of-field shot of a Go board. In the foreground, several black and white Go stones are visible, some in sharp focus and others blurred. Thin, white, semi-transparent lines radiate from a single point on the board, extending upwards and outwards to connect to the letters of the title 'AlphaGo Zero', which is centered in the upper half of the frame. The overall lighting is dim and atmospheric, with a warm, golden-brown hue from the wooden board.

0:02 / 2:13



HD YouTube





# What's next ?

- tensorflow.org
  - [github.com/tensorflow](https://github.com/tensorflow)
  - Want to learn more?
  - Udacity class on Deep Learning, [goo.gl/iHsslI](https://goo.gl/iHsslI)
  - <https://www.tensorflow.org/tutorials/>
- 
- Guides, codelabs, videos
  - MNIST for Beginners, [goo.gl/tx8R2b](https://goo.gl/tx8R2b)
  - TF Learn Quickstart, [goo.gl/uiefRn](https://goo.gl/uiefRn)
  - TensorFlow for Poets, [goo.gl/bVjFIL](https://goo.gl/bVjFIL)
  - ML Recipes, [goo.gl/KewA03](https://goo.gl/KewA03)
  - TensorFlow and Deep Learning without a PhD, [goo.gl/pHeXe7](https://goo.gl/pHeXe7)



# Google.ai

Research

Tools

Applied AI







# Final Note, by the man himself!

What will the Machine Learning  
landscape look like in five to ten years?



Thanks!

[www.tensorflow.org](http://www.tensorflow.org)

<https://cloud.google.com/free/>