

Antecedents

Vision Case Study  
Building Blocks

Theory

Neural Units  
Neural Nets  
Deep Neural Nets

Application

ConvNets  
LSTMs  
untapt  
Reinforcement

# The Fundamentals of Deep Learning with Applications

Jon Krohn

`jon@untapt.com`

Chief Data Scientist at untapt

November 21st, 2017

(slides available at `jonkrohn.com/talks`)



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### 3 Contemporary Applications

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Deep Reinforcement Learning



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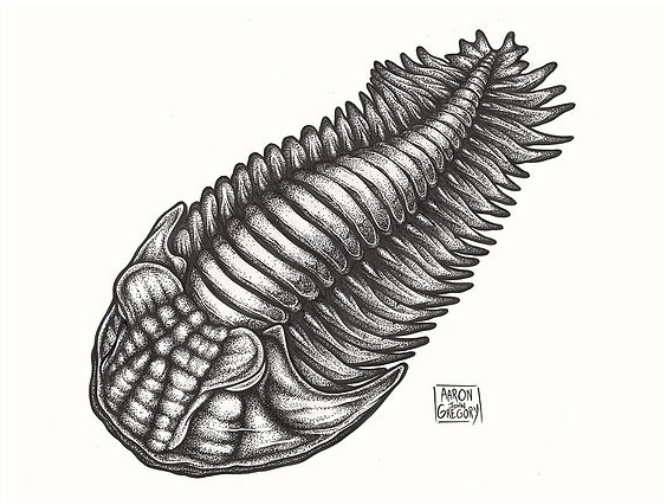
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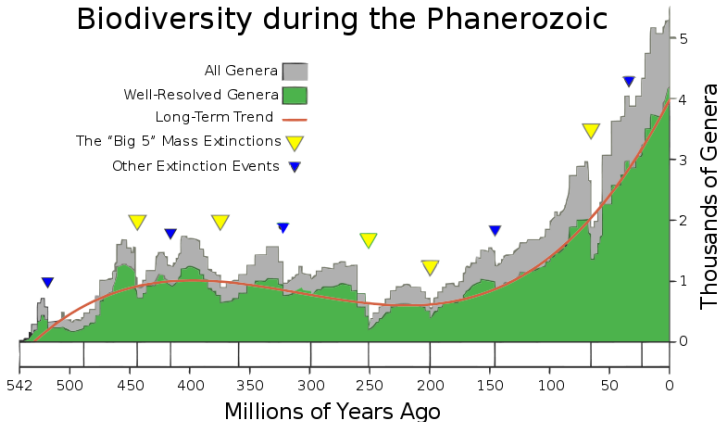
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## Biodiversity during the Phanerozoic



# Hubel & Wiesel (1959)

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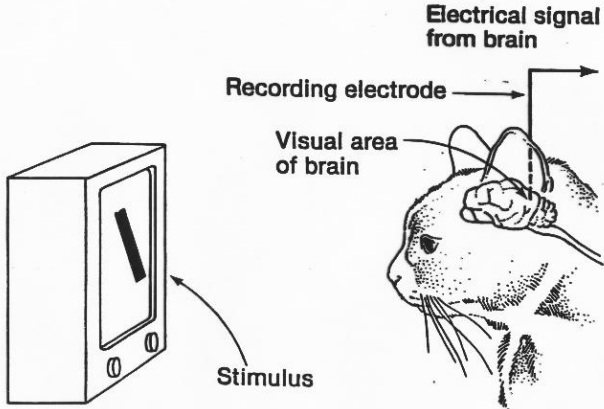
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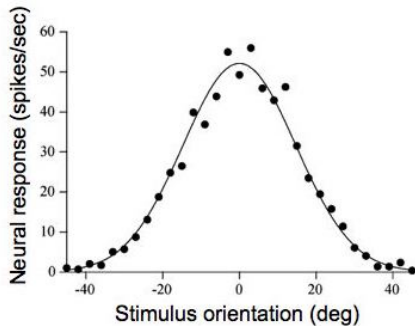
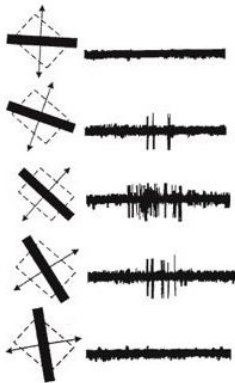
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Hubel & Wiesel, 1968



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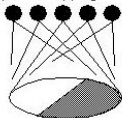
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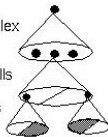
topographical mapping



hyper-complex cells

complex cells

simple cells

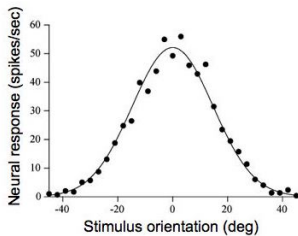
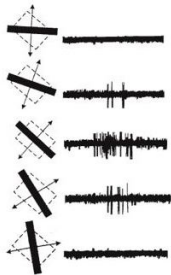


high level

mid level

low level

low level



Hubel & Wiesel, 1968



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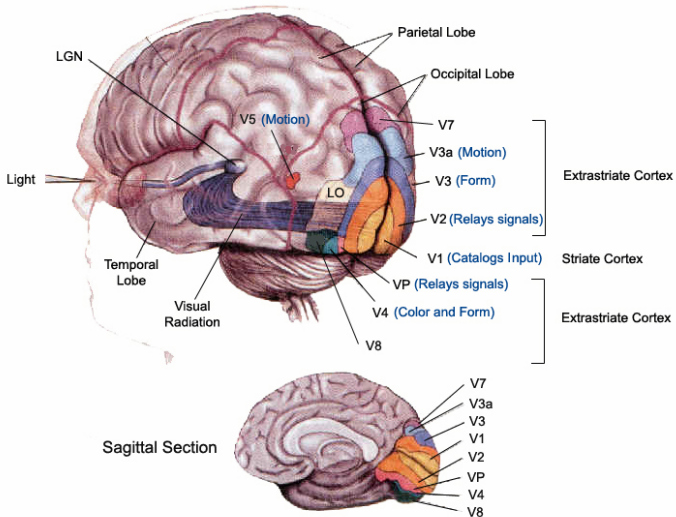
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### Visual Cortices



# Camera Obscura

da Vinci (15th Century)

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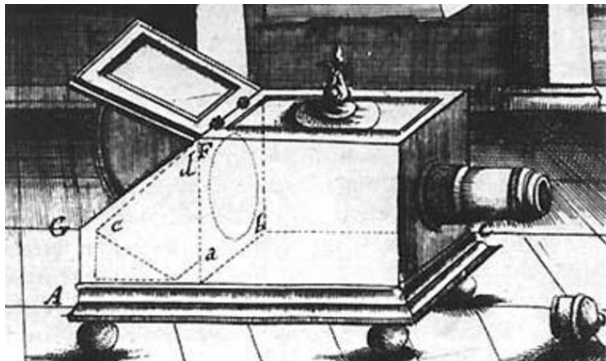
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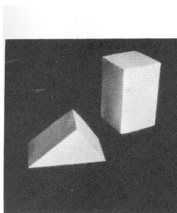
LSTMs

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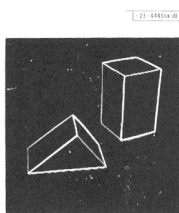
Reinforcement

# Block World

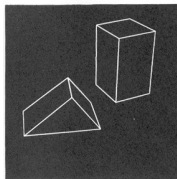
Larry Roberts (1965)



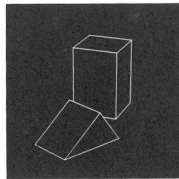
(a) Original picture.



(b) Differentiated picture.



(c) Line drawing.



(d) Rotated view.



# Viola & Jones (2001)

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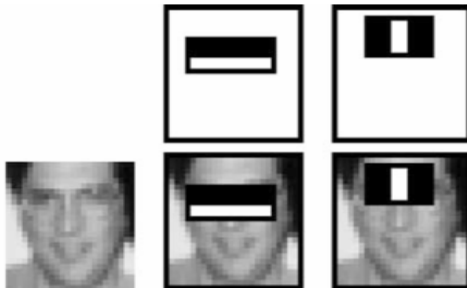
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# Neurocognitron

Fukushima (1980)

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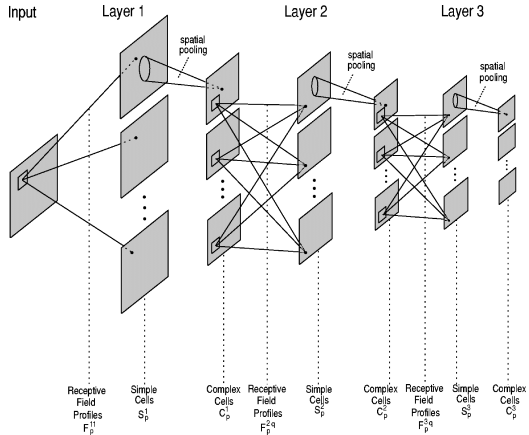
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# MNIST Digits & LeNet-5

LeCun, Boutou, Bengio & Haffner (1998)



PROC. OF THE IEEE, NOVEMBER 1998

7

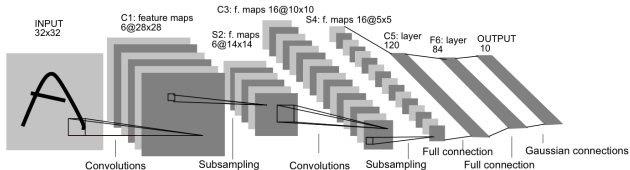


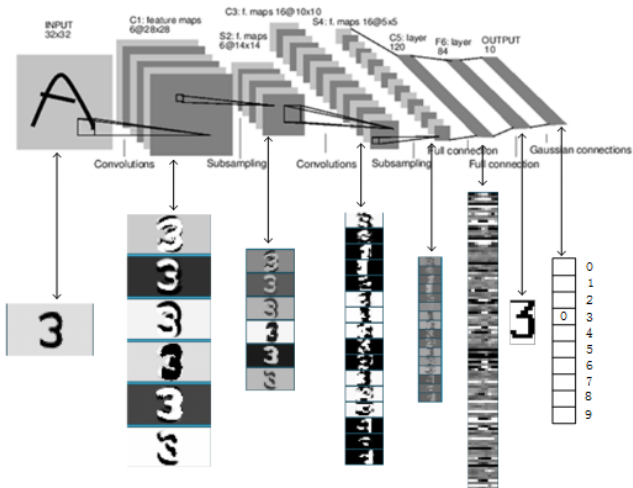
Fig. 2. Architecture of LeNet-5, a Convolutional Neural Network, here for digits recognition. Each plane is a feature map, i.e. a set of units whose weights are constrained to be identical.





# LeNet-5

LeCun, Boutou, Bengio & Haffner (1998)



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# ImageNet

Fei-Fei Li et al. (2009), 14m images, 22k categories

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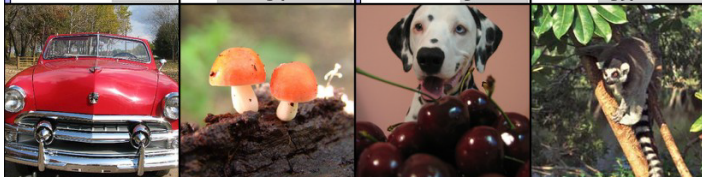
mite

container ship

motor scooter

leopard

|  |   |  |   |  |  |  |   |
|--|---|--|---|--|--|--|---|
|  | <p>mite</p> <p>black widow</p> <p>cockroach</p> <p>tick</p> <p>starfish</p> |  | <p>container ship</p> <p>lifeboat</p> <p>amphibian</p> <p>fireboat</p> <p>drilling platform</p> |  | <p>motor scooter</p> <p>go-kart</p> <p>moped</p> <p>bumper car</p> <p>golfcart</p> |  | <p>leopard</p> <p>jaguar</p> <p>cheetah</p> <p>snow leopard</p> <p>Egyptian cat</p> |
|--|---|--|---|--|--|--|---|



grille

mushroom

cherry

Madagascar cat

|  |  |  |  |  |  |  |   |
|--|--|--|--|--|--|--|---|
|  | <p>convertible</p> <p>grille</p> <p>pickup</p> <p>beach wagon</p> <p>fire engine</p> |  | <p>agaric</p> <p>mushroom</p> <p>jelly fungus</p> <p>gill fungus</p> <p>dead-man's-fingers</p> |  | <p>dalmatian</p> <p>grape</p> <p>elderberry</p> <p>ffordshire bullterrier</p> <p>currant</p> |  | <p>squirrel monkey</p> <p>spider monkey</p> <p>titi</p> <p>indri</p> <p>howler monkey</p> |
|--|--|--|--|--|--|--|---|



# ImageNet Classification Error

ILSVRC: 1.4m, 1k object classes

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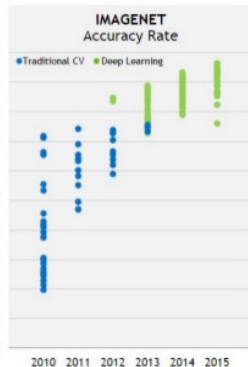
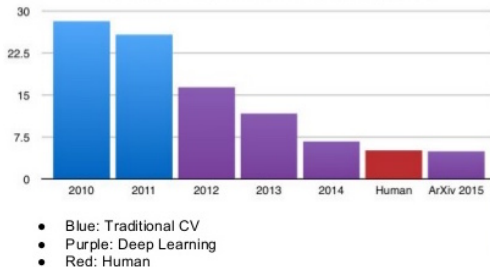
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ILSVRC top-5 error on ImageNet



# AlexNet

Krizhevsky, Sutskever & Hinton (2012)

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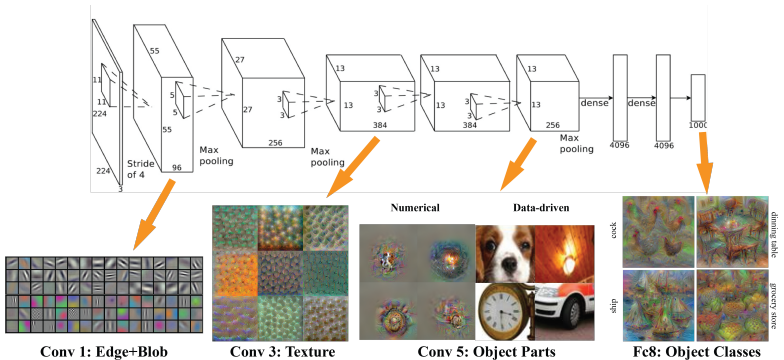
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# Deep Learning

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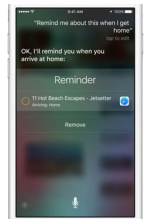
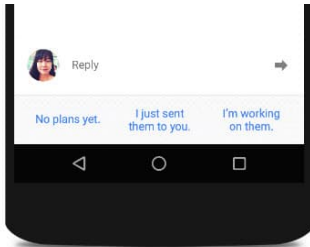
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INT. SHOP

He sees H pull a book from a shelf, flip through it while speaking, and then put it back.

H

In a future with mass unemployment, young people are forced to sell blood. That's the first thing I can do.

H2

You should see the boys and shut up. I was the one who was going to be a hundred years old.

H

I saw him again. The way you were sent to me... that was a big honest idea. I am not a bright light.

C

Well, I have to go to the skull. I don't know.

He picks up a light screen and fights the security force of the particles of a transmission on his face.

H

(continuing)

What do you mean?

C

(smiles)

I don't know anything about any of this.

H

(to Hank, taking his eyes from his mouth)

Then what?

H2

There's no answer.

C

(frowning)

We're going to see the money.

H

(reading)

'All right, you can't tell me that.'

steps back. Coffey is still going through.

C

I was coming to that thing because you were so pretty.

H

I don't know. I don't know what you're talking about.

C

That's right.

H

So what are you doing?

H2

I don't want to be honest with you.

H

He looks at him for a moment, then smiles at him.

H

You don't have to be a doctor.

H2

I am not sure. I don't know what you're talking about.

H

I want to see you too.

H2

What do you mean?

H

I'm sorry, but I'm sure you wouldn't even touch me.

H2

I don't know what you're talking about.

H

The principle is completely constructed for the same time.

H2

(smiling)

It was all about you to be true.

H

You didn't even see the movie with the rest of the bass.

H2

I don't know.

H

I don't care.

H2

I know that it's a consequence. Whatever you want to know about the presence of the story, I'm a little bit of a boy on the floor.

H

I don't know. I just have to ask you to explain to me what you say.

H2

What do you mean?

H

Because I don't know what you're talking about.

H2

That was all the time.

H

I know that.

H2

I don't know.

H

(smiles)

It would be a good time. I think I could have been my life.

H

He starts to shake.

H (CONT'D)

It may never be forgiven, but that is just too bad. I have to leave, but I'm not free of the world.

C

Yes. Perhaps I should take it from there. I'm not going to do something.

H

You can't afford to take this anymore. It's not a game. But I've got a good time to stay there.

C

Well, I think you can still be back on the table.

H

Wow. It's a damn thing scared to say. Nothing is going to be a thing but I was the one that got on this rock with a child and then I left the other two.

He is standing in the store and sitting on the floor. He takes a seat on the counter and pulls the camera over to his back. He stares at it. He is on the phone. He cuts the shotgun from the edge of the room and puts it in his mouth. He sees a black hole in the floor leading to the man on the roof.

He comes up behind him to protect him. He is still standing next to him.

He looks through the door and the door closes. He looks at the boy from his backpack, and starts to cry.

T

Well, there's the situation with me and the light on the ship. The guy was trying to stop me. He was like a boy and he was young. I was worried about him, but even if he would have done it all. He couldn't own any more. I didn't mean to be a virgin. I mean, he was weak. And I thought I'd change my mind. He was crazy to take it out. It was a long time ago. He was a little late. I was going to be a moment. I just wanted to see you that I see much better than he did. I had to stop him and I couldn't even call. I didn't want to hurt him. I'm sorry. I know I don't like him. I can go home and be so bad and I love him. So I can get him all the way over here and find the square and go to the game with him and she won't show up. Then I'll check it out. But I'm going to see him when he gets to me. He looks like me and he throws me out of his eyes. Then he said he'd go to bed with me.



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# Sunspring

Sharp & Goodwin (2016)

[video]





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- local machine
- build your own server
- AWS / Google Cloud Platform
- GPU(s) / TPU(s)



# Popular Libraries

based on Johnson (2016) in Stanford CS231n I.12

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|                             | <b>Caffe</b> | <b>Torch</b> | <b>Theano</b> | <b>TensorFlow</b> |
|-----------------------------|--------------|--------------|---------------|-------------------|
| <i>Language</i>             | C++, Python  | Lua          | Python        | Python            |
| <i>Pretrained</i>           | Yes++        | Yes++        | Yes (Lasagne) | Inception         |
| <i>Parallel GPUs: Data</i>  | Yes          | Yes          | Yes           | Yes               |
| <i>Parallel GPUs: Model</i> | No           | Yes          | Experimental  | Yes (best)        |
| <i>Readable Source Code</i> | Yes (C++)    | Yes          | No            | No                |
| <i>Good at RNN</i>          | No           | Mediocre     | Yes           | Yes (best)        |
| <i>Higher-Level APIs</i>    | No           | No           | Keras         | Keras and TFLearn |



## [Human Learning Resources]

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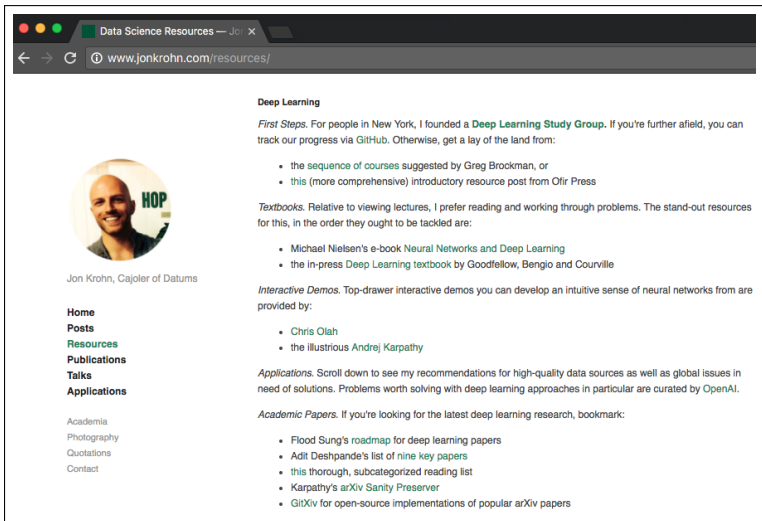
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Data Science Resources — Jon Krohn

www.jonkrohn.com/resources/

**Deep Learning**

*First Steps.* For people in New York, I founded a **Deep Learning Study Group**. If you're further afield, you can track our progress via GitHub. Otherwise, get a lay of the land from:

- the sequence of courses suggested by Greg Brockman, or
- this (more comprehensive) introductory resource post from Ofir Press

*Textbooks.* Relative to viewing lectures, I prefer reading and working through problems. The stand-out resources for this, in the order they ought to be tackled are:

- Michael Nielsen's e-book *Neural Networks and Deep Learning*
- the in-press *Deep Learning* textbook by Goodfellow, Bengio and Courville

*Interactive Demos.* Top-drawer interactive demos you can develop an intuitive sense of neural networks from are provided by:

- Chris Olah
- the illustrious Andrej Karpathy

*Applications.* Scroll down to see my recommendations for high-quality data sources as well as global issues in need of solutions. Problems worth solving with deep learning approaches in particular are curated by OpenAI.

*Academic Papers.* If you're looking for the latest deep learning research, bookmark:

- Flood Sung's roadmap for deep learning papers
- Adit Deshpande's list of nine key papers
- this thorough, subcategorized reading list
- Karpathy's arXiv Sanity Preserver
- GitXiv for open-source implementations of popular arXiv papers

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# Biological Neuron Morphology

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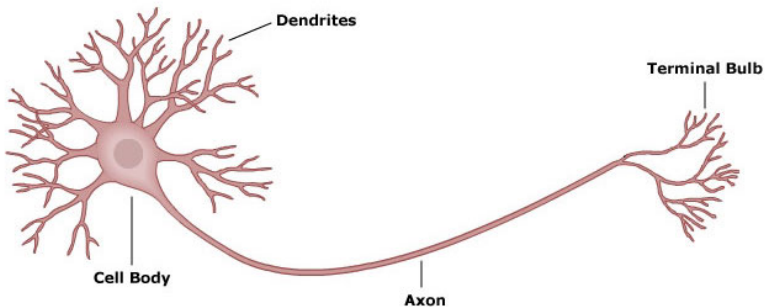
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# Perceptron

Rosenblatt (1957)

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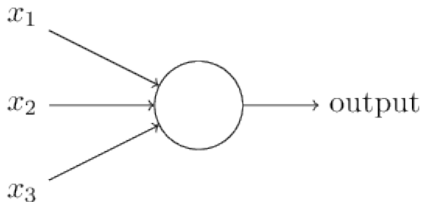
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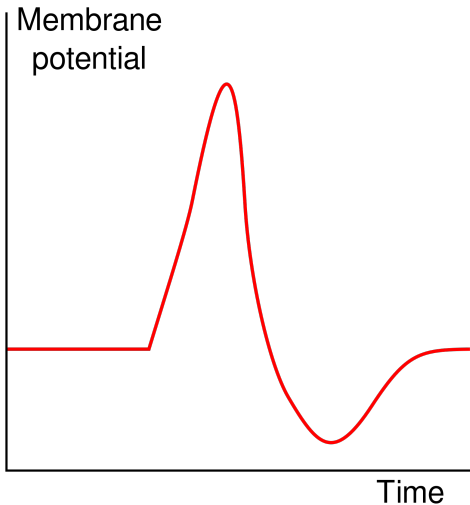


$$\text{output} = \begin{cases} 0 & \text{if } \sum_j w_j x_j \leq \text{threshold} \\ 1 & \text{if } \sum_j w_j x_j > \text{threshold} \end{cases}$$



# Biological Neuron Physiology

## The *Binary* Action Potential



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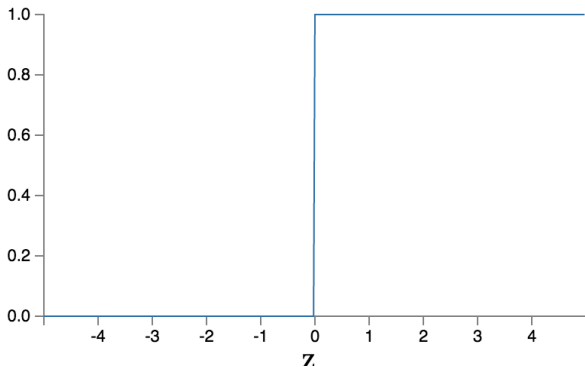
- Neural Units**
- Neural Nets
- Deep Neural Nets

Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

# Perceptron

Rosenblatt (1957)



# Multi-Layer Perceptron

## Antecedents

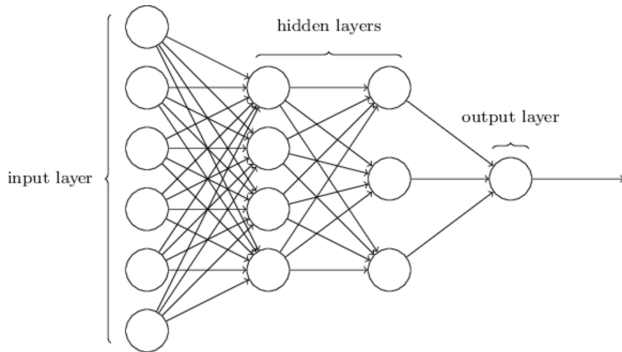
- Vision Case Study
- Building Blocks

## Theory

- Neural Units
- Neural Nets
- Deep Neural Nets

## Application

- ConvNets
- LSTMs
- untapt
- Reinforcement



# Multi-Layer Perceptron

## Antecedents

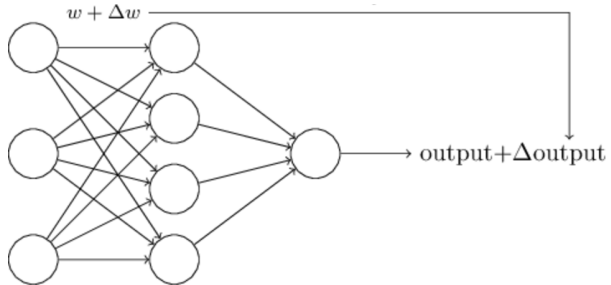
- Vision Case Study
- Building Blocks

## Theory

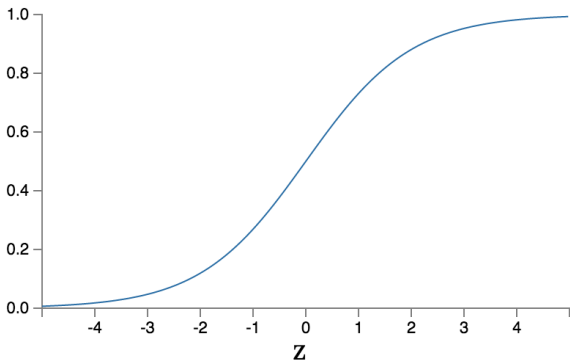
- Neural Units
- Neural Nets
- Deep Neural Nets

## Application

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- untapt
- Reinforcement



# Sigmoid Neuron



$$\frac{1}{1 + \exp(-\sum_j w_j x_j - b)}$$



Antecedents

- Vision Case Study
- Building Blocks

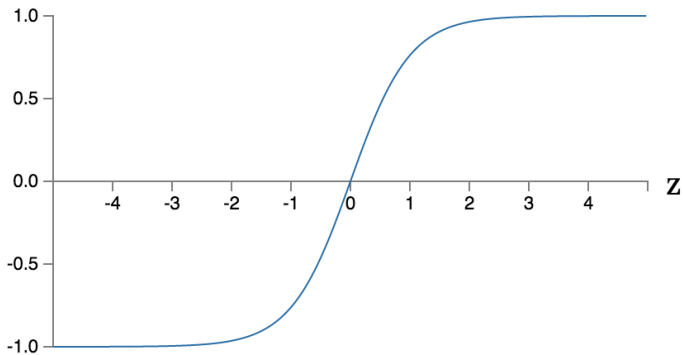
Theory

- Neural Units
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- Deep Neural Nets

Application

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- LSTMs
- untapt
- Reinforcement

# *tanh* Neuron



$$\sigma(z) = \frac{1 + \tanh(z/2)}{2}$$



# ReLU: Rectified Linear Units

Nair & Hinton (2010); Maas, Hannun & Ng (2014)

## Antecedents

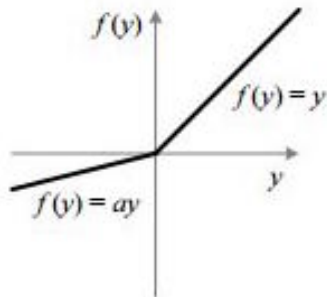
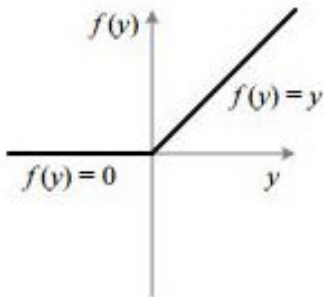
- Vision Case Study
- Building Blocks

## Theory

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## Application

- ConvNets
- LSTMs
- untapt
- Reinforcement



## Antecedents

Vision Case Study  
Building Blocks

## Theory

Neural Units

### Neural Nets

Deep Neural Nets

## Application

ConvNets

LSTMs

untapt

Reinforcement

- 1 Antecedents
  - Case Study: A History of Biological & Artificial Vision
  - Building Blocks
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  - Biological & Artificial Neurons
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  - Deep Neural Networks
- 3 Contemporary Applications
  - Convolutional Neural Networks
  - Long Short-Term Memory Recurrent Neural Networks
  - Deep Learning at untapt
  - Deep Reinforcement Learning



# MNIST

LeCun, Cortes & Burges

## Antecedents

- Vision Case Study
- Building Blocks

## Theory

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- Neural Nets**
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## Application

- ConvNets
- LSTMs
- untapt
- Reinforcement





# Fully-Connected Neural Net

## Single Hidden Layer

### Antecedents

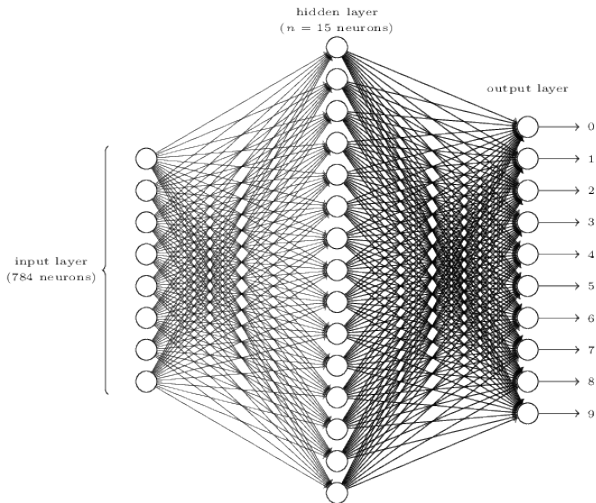
- Vision Case Study
- Building Blocks

### Theory

- Neural Units
- Neural Nets**
- Deep Neural Nets

### Application

- ConvNets
- LSTMs
- untapt
- Reinforcement



# TensorFlow Playground

## Antecedents

Vision Case Study  
Building Blocks

## Theory

Neural Units  
**Neural Nets**  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement

[demo]



## Antecedents

Vision Case Study  
Building Blocks

## Theory

Neural Units  
Neural Nets  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement

- 1 Antecedents
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  - Deep Neural Networks
- 3 Contemporary Applications
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  - Deep Learning at untapt
  - Deep Reinforcement Learning



# Deep Fully-Connected Net

## 3 (or more) Hidden Layers

### Antecedents

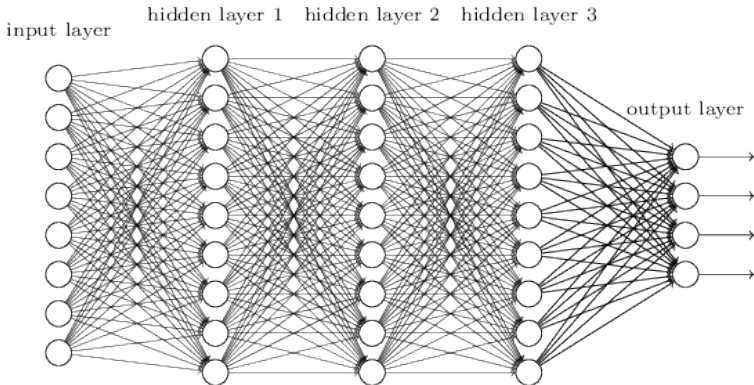
- Vision Case Study
- Building Blocks

### Theory

- Neural Units
- Neural Nets
- Deep Neural Nets

### Application

- ConvNets
- LSTMs
- untapt
- Reinforcement



## Antecedents

- Vision Case Study
- Building Blocks

## Theory

- Neural Units
- Neural Nets
- Deep Neural Nets

## Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

|                             | Caffe       | Torch    | Theano        | TensorFlow        |
|-----------------------------|-------------|----------|---------------|-------------------|
| <i>Language</i>             | C++, Python | Lua      | Python        | Python            |
| <i>Pretrained</i>           | Yes++       | Yes++    | Yes (Lasagne) | Inception         |
| <i>Parallel GPUs: Data</i>  | Yes         | Yes      | Yes           | Yes               |
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| <i>Readable Source Code</i> | Yes (C++)   | Yes      | No            | No                |
| <i>Good at RNN</i>          | No          | Mediocre | Yes           | Yes (best)        |
| <i>Higher-Level APIs</i>    | No          | No       | Keras         | Keras and TFLearn |



# A Simple Deep Net in TFLearn

## Antecedents

- Vision Case Study
- Building Blocks

## Theory

- Neural Units
- Neural Nets
- Deep Neural Nets**

## Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

[notebook]



# Synaptic Pruning

## Antecedents

Vision Case Study  
Building Blocks

## Theory

Neural Units  
Neural Nets  
**Deep Neural Nets**

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement



# (Stochastic) Gradient Descent

Adam = AdaGrad + RMSprop

## Antecedents

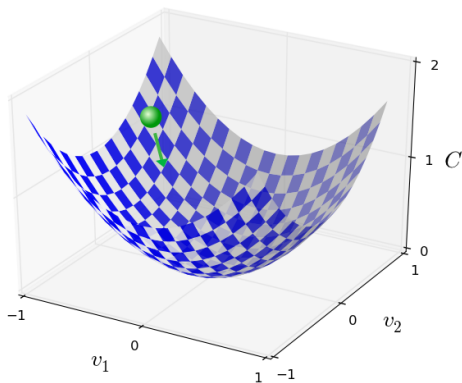
Vision Case Study  
Building Blocks

## Theory

Neural Units  
Neural Nets  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement





Antecedents

- Vision Case Study
- Building Blocks

Theory

- Neural Units
- Neural Nets

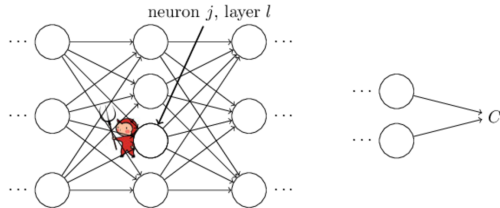
Deep Neural Nets

Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

# Backpropagation

computes error & gradient of cost function



$$\delta^L = \nabla_a C \odot \sigma'(z^L) \tag{BP1}$$

$$\delta^l = ((w^{l+1})^T \delta^{l+1}) \odot \sigma'(z^l) \tag{BP2}$$

$$\frac{\partial C}{\partial b_j^l} = \delta_j^l \tag{BP3}$$

$$\frac{\partial C}{\partial w_{jk}^l} = a_k^{l-1} \delta_j^l \tag{BP4}$$



# Overfitting

...and avoiding it

## Antecedents

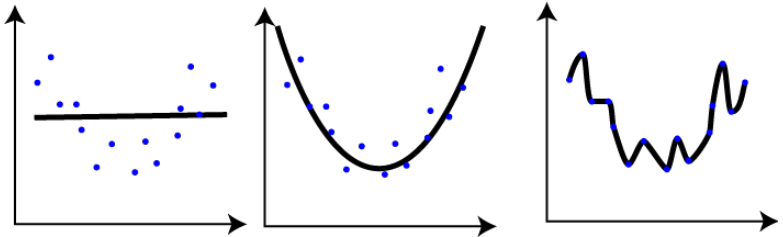
Vision Case Study  
Building Blocks

## Theory

Neural Units  
Neural Nets  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement



- L1/L2 regularization
- dropout
- artificial data set expansion



# Improving Neural Networks

Mostly Hyperparameter Tuning

## Antecedents

Vision Case Study

Building Blocks

## Theory

Neural Units

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Deep Neural Nets

## Application

ConvNets

LSTMs

untapt

Reinforcement

- problem simplification
- number and width of layers
- cost fxn: quadratic, cross-entropy, log-likelihood, &c.
- more epochs, early stopping
- clever initialization of weights and biases
- learning rate  $\eta$ , variable schedule
- regularization parameter  $\lambda$
- mini-batch size
- automation, e.g., with Spearmint

[Summary Blog Post]



Antecedents

- Vision Case Study
- Building Blocks

Theory

- Neural Units
- Neural Nets

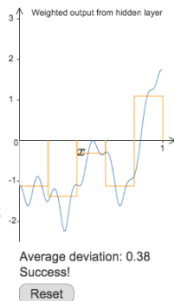
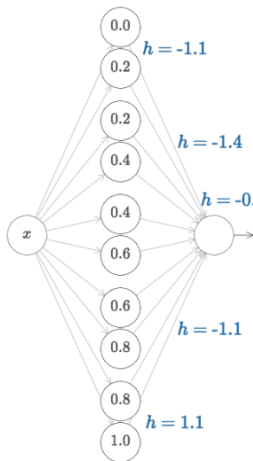
Deep Neural Nets

Application

- ConvNets
- LSTMs
- untapt
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# Universality

Solve Any Continuous Function (Nielsen, 2015)



# Unstable Gradient

Typically *Vanishes* (but can *Explode*)

Antecedents

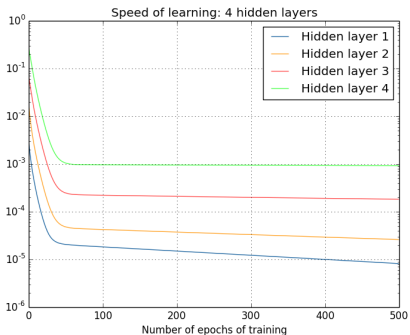
- Vision Case Study
- Building Blocks

Theory

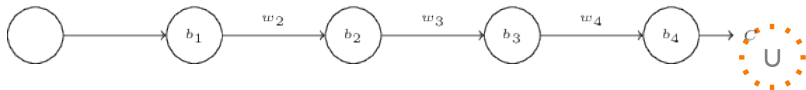
- Neural Units
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Application

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$$\frac{\partial C}{\partial b_1} = \sigma'(z_1) \times w_2 \times \sigma'(z_2) \times w_3 \times \sigma'(z_3) \times w_4 \times \sigma'(z_4) \times \frac{\partial C}{\partial a_4}$$



Antecedents

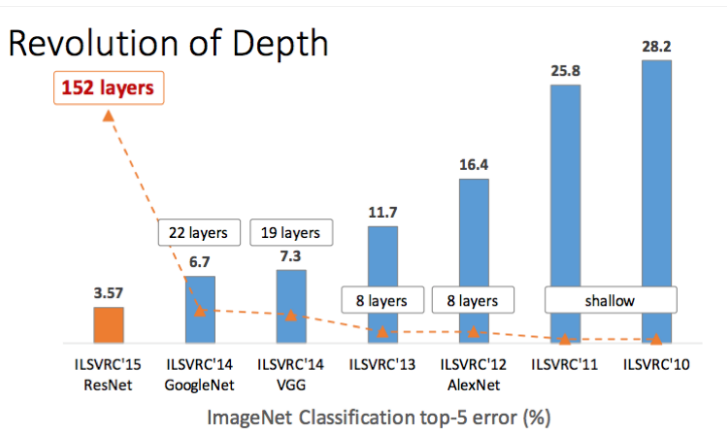
- Vision Case Study
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# Classic Deep Architectures

...introducing *Convolutional Layers*

## Antecedents

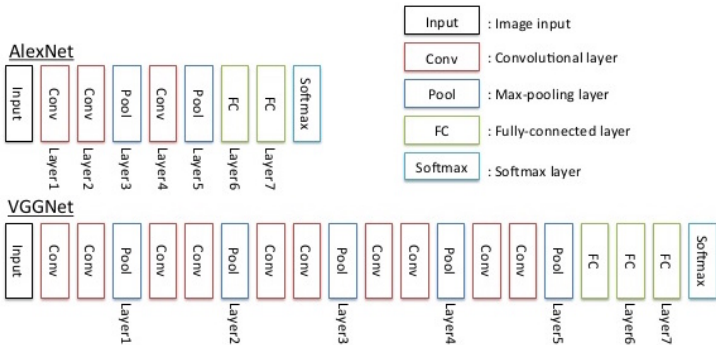
- Vision Case Study
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## Antecedents

Vision Case Study  
Building Blocks

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Neural Nets  
Deep Neural Nets

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ConvNets  
LSTMs  
untapt  
Reinforcement

- 1 Antecedents
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  - Long Short-Term Memory Recurrent Neural Networks
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  - Deep Reinforcement Learning





# Hubel & Wiesel (1959)

## Antecedents

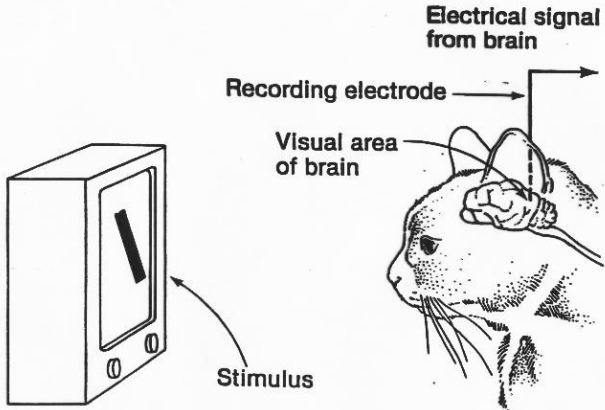
Vision Case Study  
Building Blocks

## Theory

Neural Units  
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## Antecedents

Vision Case Study  
Building Blocks

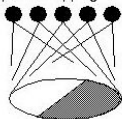
## Theory

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## Application

ConvNets  
LSTMs  
untapt  
Reinforcement

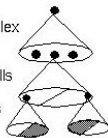
topographical mapping



hyper-complex cells

complex cells

simple cells

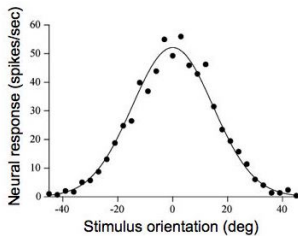
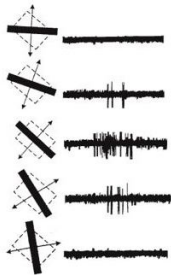


high level

mid level

low level

low level



Hubel & Wiesel, 1968



Antecedents

- Vision Case Study
- Building Blocks

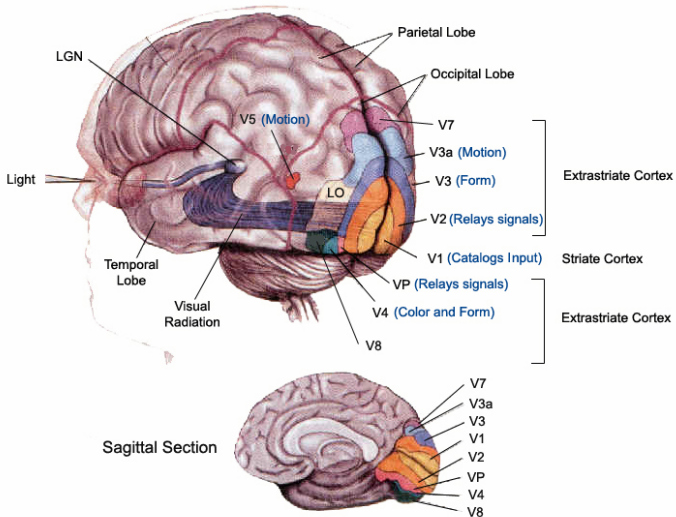
Theory

- Neural Units
- Neural Nets
- Deep Neural Nets

Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

## Visual Cortices



# AlexNet

Krizhevsky, Sutskever & Hinton (2012)

Antecedents

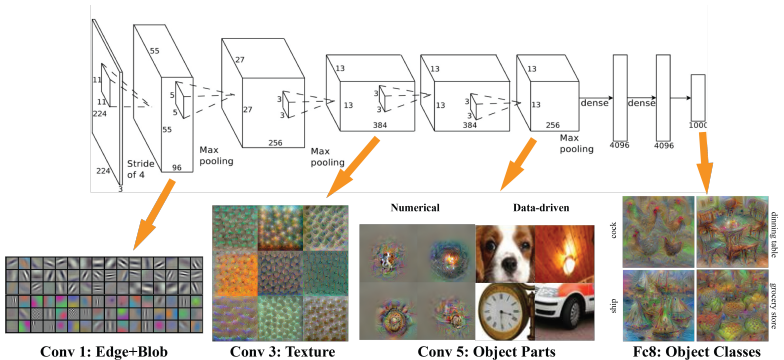
- Vision Case Study
- Building Blocks

Theory

- Neural Units
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- unlapt
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Antecedents

- Vision Case Study
- Building Blocks

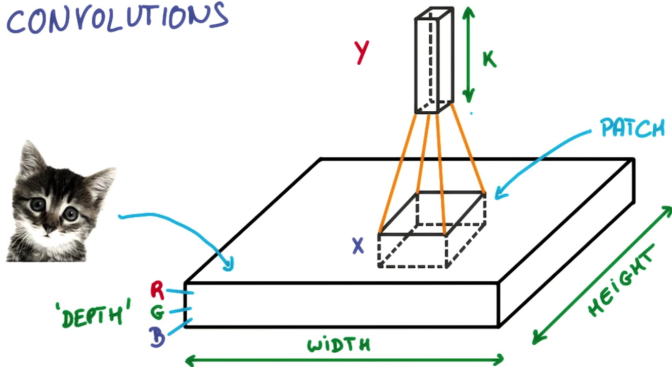
Theory

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- Deep Neural Nets

Application

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# CONVOLUTIONS



Antecedents

Vision Case Study  
Building Blocks

Theory

Neural Units  
Neural Nets  
Deep Neural Nets

Application

**ConvNets**  
LSTMs  
untapt  
Reinforcement

# ConvNet Visualisation

Yosinski et al. (2015)

[video]



# Network Architectures

## Antecedents

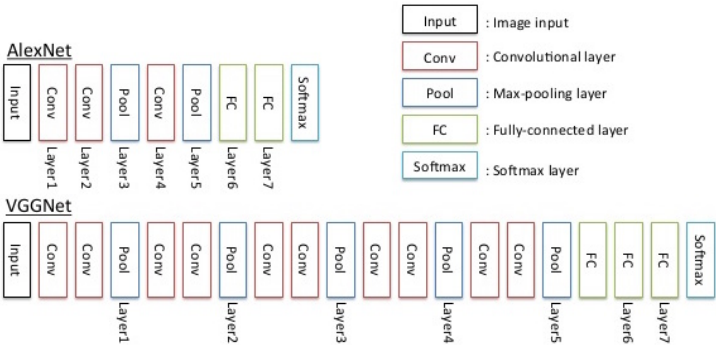
- Vision Case Study
- Building Blocks

## Theory

- Neural Units
- Neural Nets
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- Reinforcement



# AlexNet: ILSVRC '12 winner

Krizhevsky et al. (2012)

## Antecedents

- Vision Case Study
- Building Blocks

## Theory

- Neural Units
- Neural Nets
- Deep Neural Nets

## Application

- ConvNets**
- LSTMs
- untapt
- Reinforcement

[TFLearn notebook]





# VGGNet: ILSVRC '14 runner-up

Simonyan & Zisserman (2015)

## Antecedents

Vision Case Study

Building Blocks

## Theory

Neural Units

Neural Nets

Deep Neural Nets

## Application

**ConvNets**

LSTMs

untapt

Reinforcement

[TFLearn notebook]



# ConvNet in TensorFlow

Antecedents

- Vision Case Study
- Building Blocks

Theory

- Neural Units
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- Deep Neural Nets

Application

- ConvNets
- LSTMs
- untapt
- Reinforcement

|                             | <b>Caffe</b> | <b>Torch</b> | <b>Theano</b> | <b>TensorFlow</b> |
|-----------------------------|--------------|--------------|---------------|-------------------|
| <i>Language</i>             | C++, Python  | Lua          | Python        | Python            |
| <i>Pretrained</i>           | Yes++        | Yes++        | Yes (Lasagne) | Inception         |
| <i>Parallel GPUs: Data</i>  | Yes          | Yes          | Yes           | Yes               |
| <i>Parallel GPUs: Model</i> | No           | Yes          | Experimental  | Yes (best)        |
| <i>Readable Source Code</i> | Yes (C++)    | Yes          | No            | No                |
| <i>Good at RNN</i>          | No           | Mediocre     | Yes           | Yes (best)        |
| <i>Higher-Level APIs</i>    | No           | No           | Keras         | Keras and TFLearn |



# ConvNet in TensorFlow

## Antecedents

Vision Case Study

Building Blocks

## Theory

Neural Units

Neural Nets

Deep Neural Nets

## Application

**ConvNets**

LSTMs

untapt

Reinforcement

[notebook]



## Antecedents

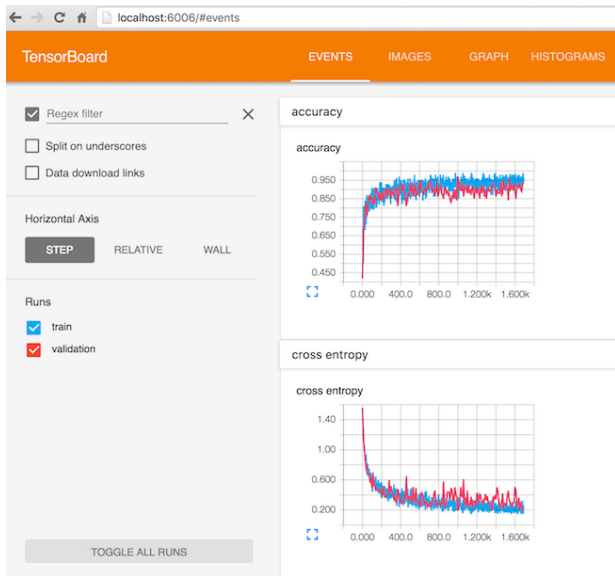
Vision Case Study  
Building Blocks

## Theory

Neural Units  
Neural Nets  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement



# ConvNet in Theano

Antecedents

- Vision Case Study
- Building Blocks

Theory

- Neural Units
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Application

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- untapt
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|                             | <b>Caffe</b> | <b>Torch</b> | <b>Theano</b> | <b>TensorFlow</b> |
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Antecedents

Vision Case Study  
Building Blocks

Theory

Neural Units  
Neural Nets  
Deep Neural Nets

Application

**ConvNets**  
LSTMs  
untapt  
Reinforcement

# ConvNet in Theano

[demo]



# ConvNet in Keras

calls TensorFlow or Theano

Antecedents

- Vision Case Study
- Building Blocks

Theory

- Neural Units
- Neural Nets
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Application

- ConvNets
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- untapt
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|                             | Caffe       | Torch    | Theano        | TensorFlow        |
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Antecedents

Vision Case Study  
Building Blocks

Theory

Neural Units  
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Deep Neural Nets

Application

**ConvNets**  
LSTMs  
untapt  
Reinforcement

# ConvNet in Keras

calls TensorFlow or Theano

[notebook]





# “2.5-dimension” CT Scans

Roth et al. (2015)

## Antecedents

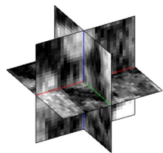
Vision Case Study  
Building Blocks

## Theory

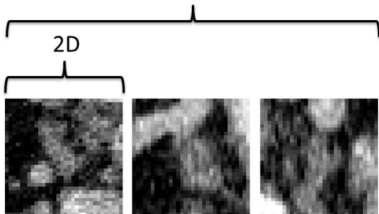
Neural Units  
Neural Nets  
Deep Neural Nets

## Application

ConvNets  
LSTMs  
untapt  
Reinforcement



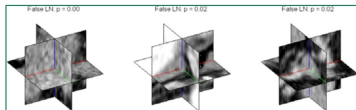
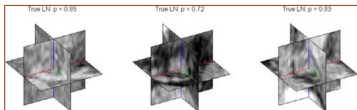
2.5D



# Computer-Aided Detection

Shin et al. (2016); Roth et al. (2016)

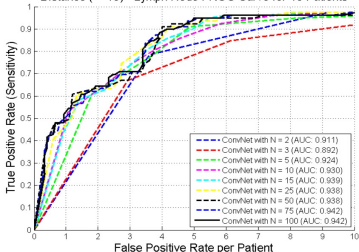
**Experimental Results** (~100% sensitivity but ~40 FPs/patient at candidate generation step; then 3-fold CV with data augmentation)



## Mediastinum

**71% @ 3 FPs (was 55%)**

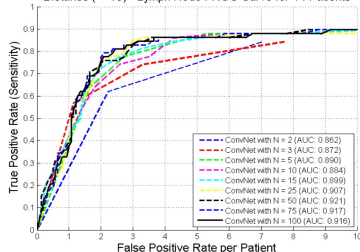
Distance ( $\leq 15$ ) - Lymph Node FROC Curve for 15 Patients



## Abdomen

**83% @ 3 FPs (was 30%)**

Distance ( $\leq 15$ ) - Lymph Node FROC Curve for 14 Patients



Antecedents

Vision Case Study  
Building Blocks

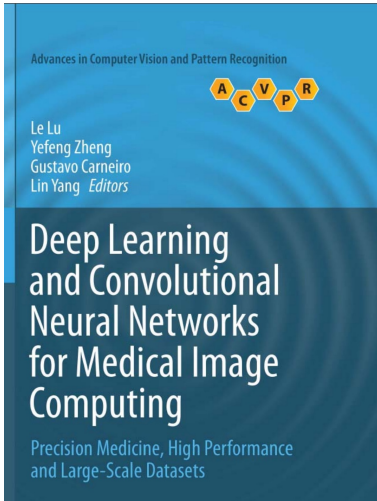
Theory

Neural Units  
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untapt  
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Antecedents

Vision Case Study  
Building Blocks

Theory

Neural Units  
Neural Nets  
Deep Neural Nets

Application

**ConvNets**  
LSTMs  
untapt  
Reinforcement

# Kaggle

## Data Science Bowl 2017

[link]



# Transfer Learning

Caffe

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|                             | Caffe       | Torch    | Theano        | TensorFlow        |
|-----------------------------|-------------|----------|---------------|-------------------|
| <i>Language</i>             | C++, Python | Lua      | Python        | Python            |
| <i>Pretrained</i>           | Yes++       | Yes++    | Yes (Lasagne) | Inception         |
| <i>Parallel GPUs: Data</i>  | Yes         | Yes      | Yes           | Yes               |
| <i>Parallel GPUs: Model</i> | No          | Yes      | Experimental  | Yes (best)        |
| <i>Readable Source Code</i> | Yes (C++)   | Yes      | No            | No                |
| <i>Good at RNN</i>          | No          | Mediocre | Yes           | Yes (best)        |
| <i>Higher-Level APIs</i>    | No          | No       | Keras         | Keras and TFLearn |



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# Transfer Learning

Caffe

[Model Zoo]



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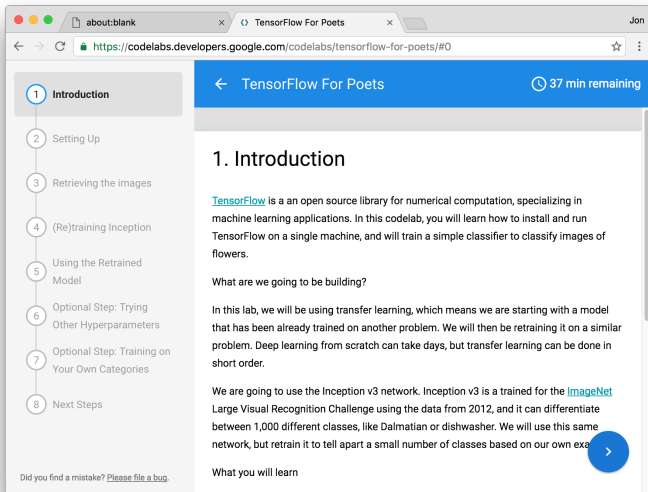
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The screenshot shows a web browser window with the URL `https://codelabs.developers.google.com/codelabs/tensorflow-for-poets/#0`. The page title is "TensorFlow For Poets" and it shows a timer for "37 min remaining". A sidebar on the left contains a table of contents with 8 items, where "1 Introduction" is selected. The main content area displays the "1. Introduction" section, which explains that TensorFlow is an open source library for numerical computation, and the lab will teach how to install and run TensorFlow on a single machine to train a simple classifier for flower images. It also introduces the concept of transfer learning and mentions the Inception v3 network.

1 Introduction

2 Setting Up

3 Retrieving the images

4 (Re)training Inception

5 Using the Retrained Model

6 Optional Step: Trying Other Hyperparameters

7 Optional Step: Training on Your Own Categories

8 Next Steps

## 1. Introduction

[TensorFlow](#) is an open source library for numerical computation, specializing in machine learning applications. In this codelab, you will learn how to install and run TensorFlow on a single machine, and will train a simple classifier to classify images of flowers.

### What are we going to be building?

In this lab, we will be using transfer learning, which means we are starting with a model that has been already trained on another problem. We will then be retraining it on a similar problem. Deep learning from scratch can take days, but transfer learning can be done in short order.

We are going to use the Inception v3 network. Inception v3 is trained for the [ImageNet](#) Large Visual Recognition Challenge using the data from 2012, and it can differentiate between 1,000 different classes, like Dalmatian or dishwasher. We will use this same network, but retrain it to tell apart a small number of classes based on our own exa

What you will learn

Did you find a mistake? [Please file a bug.](#)



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# Video Classification

[video]





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Learn More!

[ *Deep Learning with TensorFlow LiveLessons* ]



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INT. SHIP

We see H pull a book from a shelf, flip through it while speaking, and then put it back.

H  
In a future with mass unemployment, young people are forced to sell blood. That's the first thing I can do.

H2  
You should see the boys and shut up. I was the one who was going to be a hundred years old.

H  
I saw him again. The way you were sent to me... that was a big honest idea. I am not a bright light.

C  
Well, I have to go to the skull. I don't know.

H  
He picks up a light screen and fights the security force of the particles of a transmission on his face.

H  
(continuing)  
What do you mean?

C  
(smiling)  
I don't know anything about any of this.

H  
(to Hank, taking his eyes from his mouth)  
Then what?

H2  
There's no answer.

C  
(frowning)  
We're going to see the money.

H  
(reading)  
"All right, you can't tell me that."

H  
steps back. Coffey is still going through.

C  
I was coming to that thing because you were so pretty.

H  
I don't know. I don't know what you're talking about.

C  
That's right.

H  
So what are you doing?

H2  
I don't want to be honest with you.

H  
He looks at him for a moment, then smiles at him.

H2  
You don't have to be a doctor.

H  
I know that.

H2  
I don't know.

H  
(angry)  
It would be a good time. I think I could have been my life.

H  
He starts to shake.

H  
(COMPS)  
It may never be forgiven, but that is just too bad. I have to leave, but I'm not free of the world.

C  
The... Perhaps I should take it from here. I'm not going to do something.

H  
You can't afford to take this anywhere. It's not a dream. But I've got a good time to stay there.

C  
Well, I think you can still be back on the table.

H  
Man. It's a damn thing scared to say. Holding the pistol to be a thing, but I was the one that got on this rock with a child and then I left the other two.

H  
He is standing in the stars and sitting on the floor. He takes a seat on the counter and pulls the camera over to his back. He stares at it. He is on the phone. He cuts the shotgun from the edge of the room and puts it in his mouth. He sees a black hole in the floor leading to the sea on the roof.

H  
He comes up behind him to protect him. He is still standing next to him.

H2  
He looks through the door and the door closes. He looks at the bag from his backpack, and starts to cry.

T  
Well, there's the situation with me and the light on the ship. The guy was trying to stop me. He was like a baby and he was gone. I was worried about him, but even if he would have done it all. He couldn't come any more. I didn't want to be a virgin. I mean, he was weak. And I thought I'd change my mind. He was crazy to take it out. It was a long time ago. He was a little late. I was going to be a moment. I just wanted to tell you that I was much better than he did. I had to stop him and I couldn't even tell. I didn't want to hurt him. I'm sorry. I know I don't like him. I can go home and be an bad and I love him. So I can get him all the way over here and find the square and go to the game with him and she won't show up. Then I'll check it out. But I'm going to see him when he gets to me. He looks like he and he knows me out of his eyes. Then he said he'd go to bed with me.



## Antecedents

Vision Case Study  
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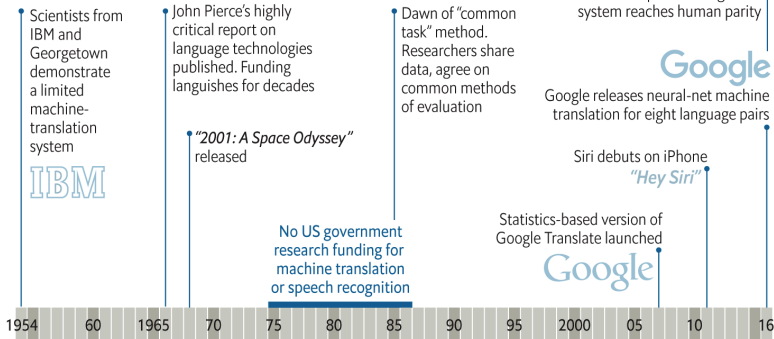
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## A history of language technologies



## Antecedents

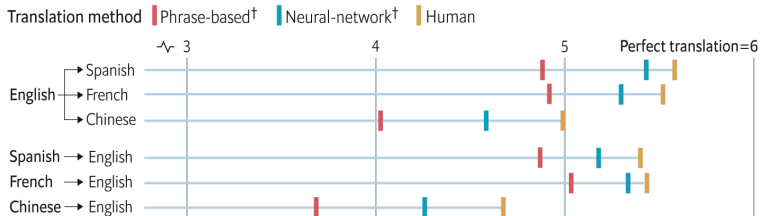
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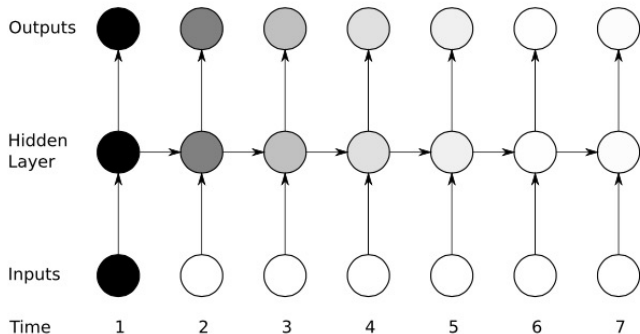
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# RNNs; *LSTM* RNNs

Hochreiter & Schmidhuber (1997)

Graves, ... & Schmidhuber (2009)



# Vector Space Embedding

Word2Vec: Mikolov, ... & Dean (2013)

## Antecedents

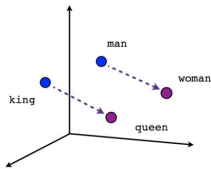
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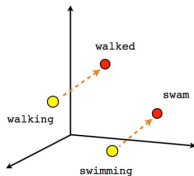
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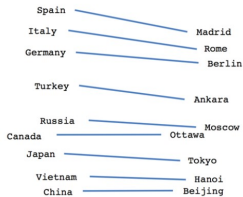
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Male-Female



Verb tense



Country-Capital





Hinton & van der Maaten (2008)

## Antecedents

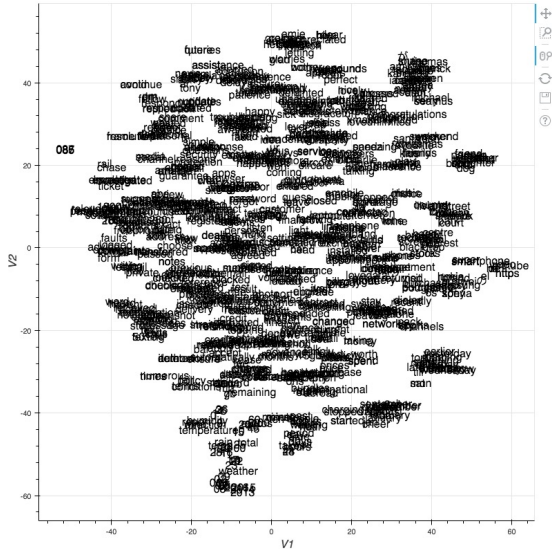
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# Word2Vec + t-SNE

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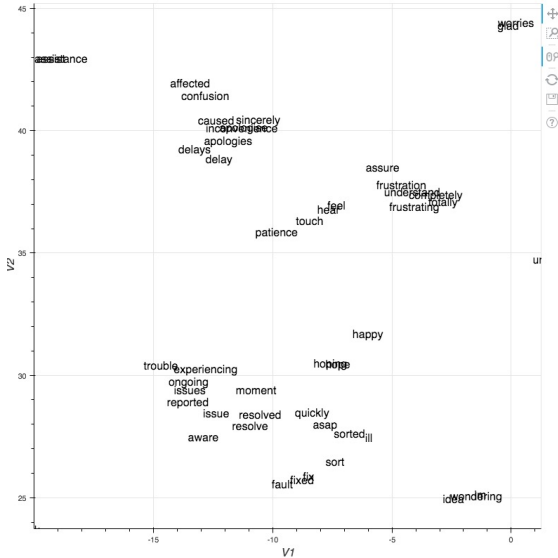
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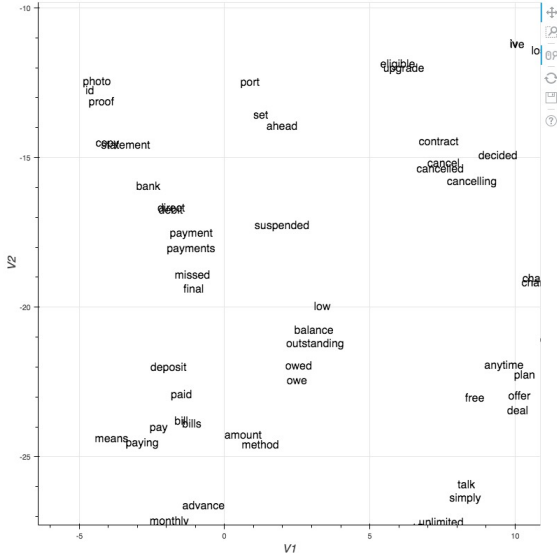
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# Word2Vec + t-SNE



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# 'Understand' Language

with Word2Vec features in your model

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```
model.most_similar(positive=['angular'])

[('angularjs', 0.9534549117088318),
 ('backbonejs', 0.9315043687820435),
 ('ember', 0.905410647392273),
 ('emberjs', 0.9029799103736877),
 ('reactjs', 0.896049439907074),
 ('requirejs', 0.8759748339653015),
 ('coffeescript', 0.8645504713058472),
 ('bootstrap', 0.8554328083992004),
 ('nodejs', 0.8515532612800598),
 ('backbone', 0.8443130254745483)]
```

```
model.most_similar(positive=['managed'])

[('oversaw', 0.8659406900405884),
 ('directed', 0.8491166234016418),
 ('supervised', 0.8058902621269226),
 ('coordinated', 0.7858685851097107),
 ('led', 0.7539615035057068),
 ('orchestrated', 0.7211644649505615),
 ('supported', 0.7198437452316284),
 ('comanaged', 0.6774874925613403),
 ('encompassing', 0.6726169586181641),
 ('administered', 0.6706464886665344)]
```

[even with small corpora]



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# Quick, Draw!

## ConvNet + LSTM

[link]



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# untapt

## Deep Neural Net

```
Give me one bullet-point from your resume:  
>> • Sat around all day checking my Facebook feed  
I predict a 0.0% chance of interview
```

```
Give me one bullet-point from your resume:  
>> • Developed trading applications in Python  
I predict a 24.6% chance of interview
```

```
Give me one bullet-point from your resume:  
>> • Developed python solution for Monte Carlo risk calculation using numpy,  
    scipy and pandas, with a Javascript frontend in AngularJS and React  
I predict a 98.1% chance of interview
```





Antecedents

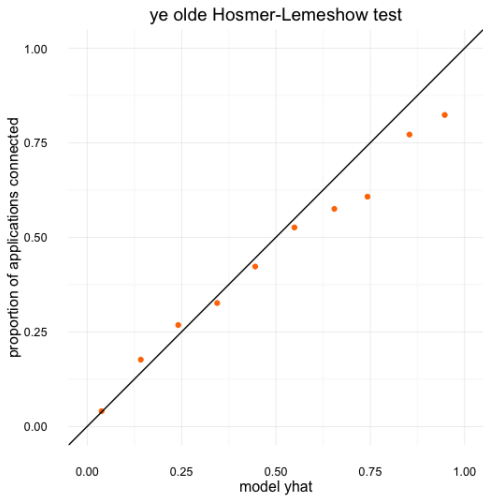
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# AlphaGO

Silver et al. (2016)

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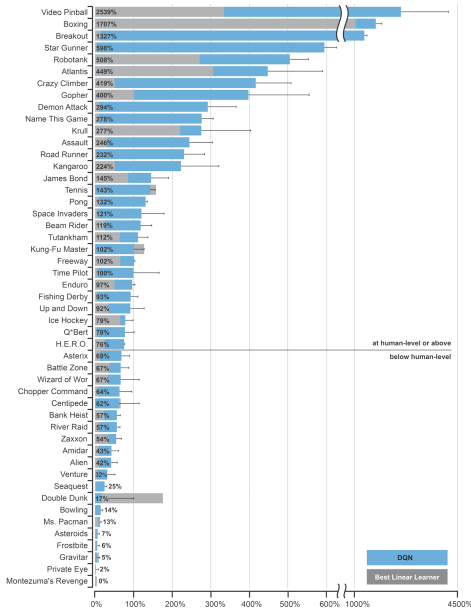
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[Atari Games]



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[OpenAI Universe]

[Google DeepMind Lab]



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