

Overview of Deep Learning in Medical Image Computing

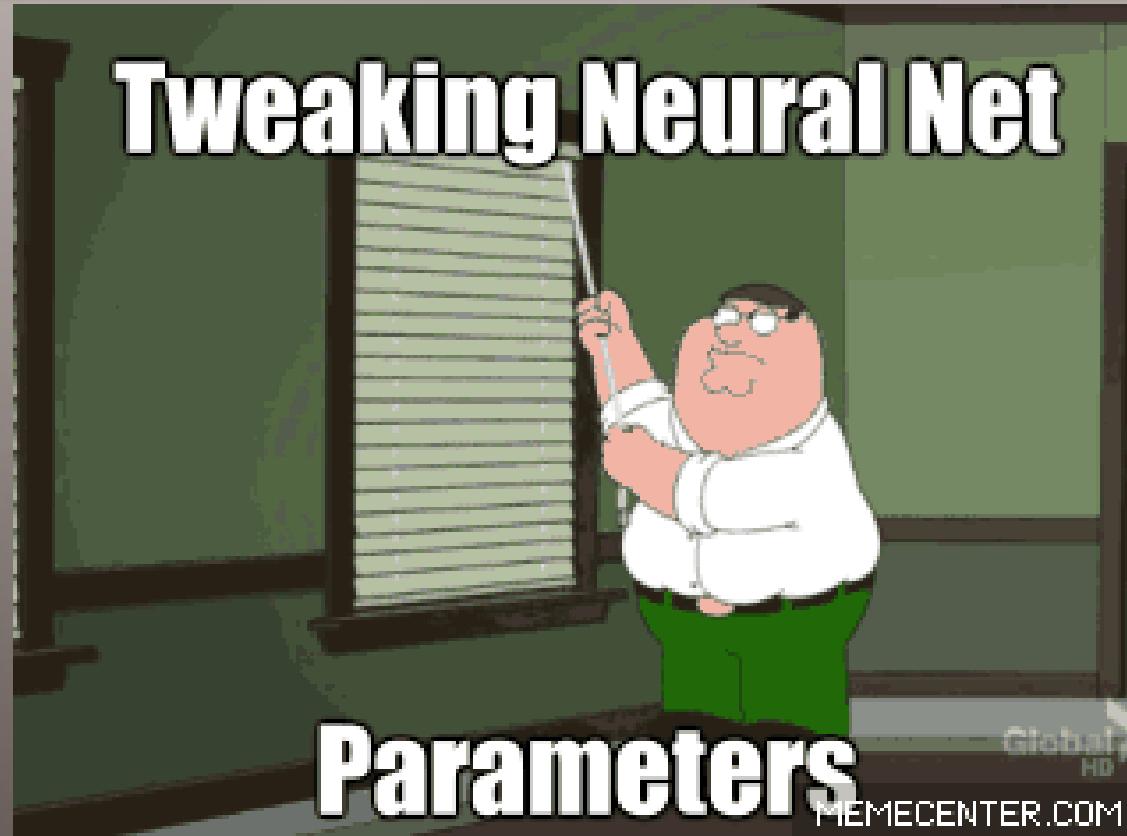
[Spring 2020 CS-8395-02 Deep Learning in Medical Image Computing]

Instructor: Yuankai Huo, Ph.D.
Department of Electrical Engineering and Computer Science
Vanderbilt University

Topics

- Introduction
- About the Class
- Syllabus
- Deep Learning in Medical Image Computing
- Assignment o

Why this class



<http://bbabenko.tumblr.com/post/83319141207/convolutional-learnings-things-i-learned-by>

I see “Deep Learning” in the class name!

Goals for This Course



<https://fyspringfield.com/post/94170889415>

- Cover Basic Concepts in Deep Learning
- Applications in Medical Image Computing
- Write Deep Learning Code on Basic Tasks
- Deep Learning Project for Medical Imaging
- Find a Good Job & Do Better Research
 - For basic math:
 - CS 6362: Machine Learning
 - CS 5260: Artificial Intelligence
 - CS 3891: Deep Learning

Organization of Course



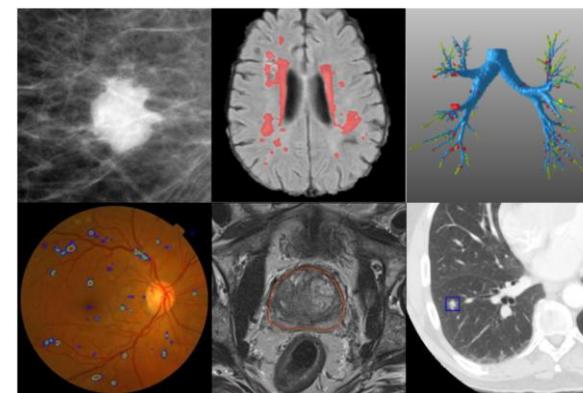
Overview

Overview of Deep Learning in Medical Image Computing
Neural Networks and CNN

Key Techs

Classification (Medical Image Diagnosis)
Detection (Landmark Localization and Detection)
Segmentation (Medical Image Segmentation)
GAN (Medical Image Synthesis)

Topics in Medical Image Computing



Overview of “Deep Learning”



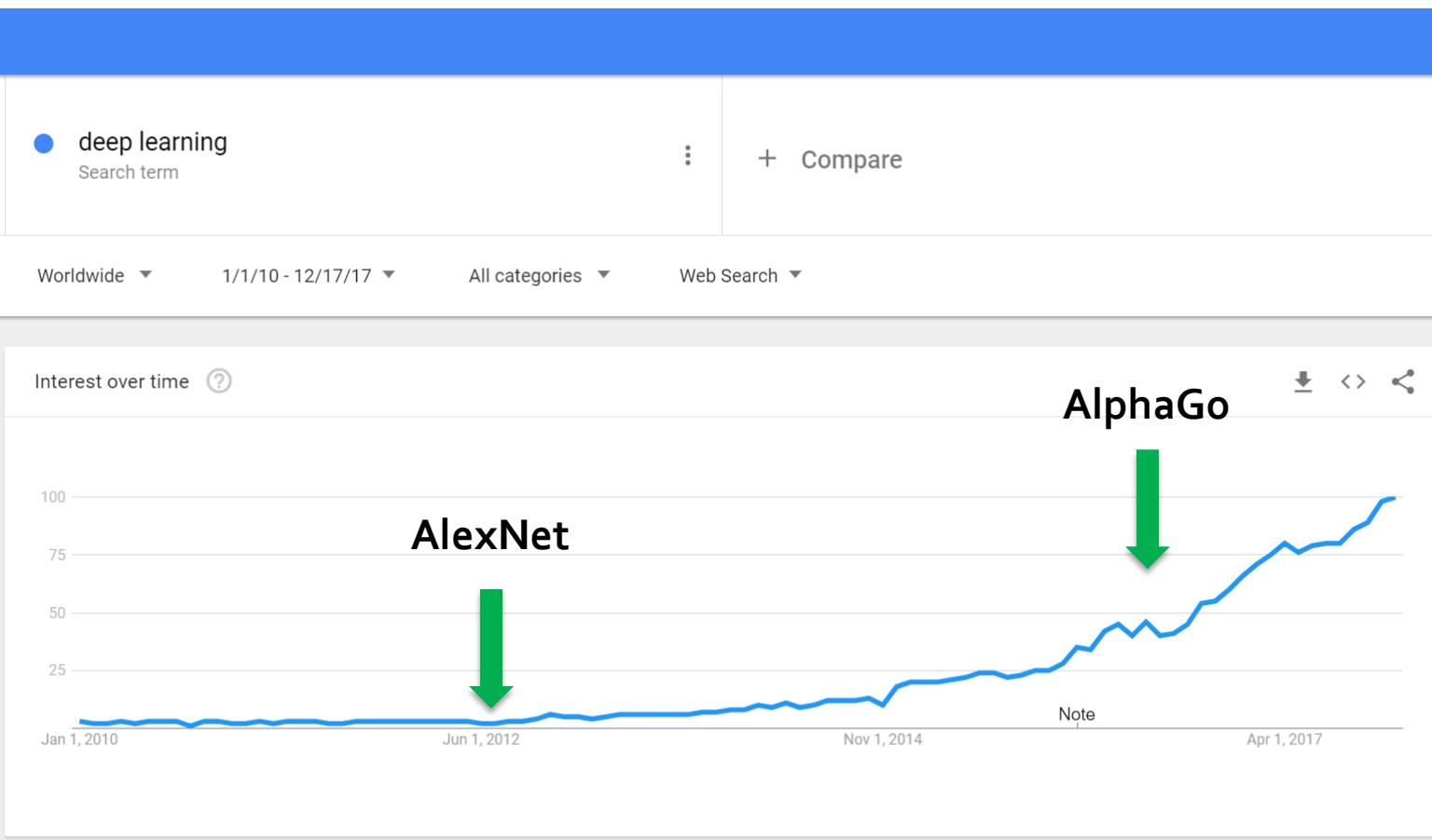
[PDF] [A Fast Learning Algorithm for Deep Belief Nets - University of Toronto...](http://www.cs.toronto.edu/~fritz/absps/ncfast.pdf)
www.cs.toronto.edu/~fritz/absps/ncfast.pdf ▾
by GE Hinton - Cited by 8687 - Related articles
fast, greedy algorithm that can **learn deep**, directed belief networks one layer at a time, provided the ...
Neural Computation 18, 1527–1554 (2006). C 2006 ... contrastive version of the wake-sleep algorithm
(Hinton, Dayan, Frey,.. & Neal, 1995) ...

People also search for

bengio et al., 2007 yee-whye teh
deep belief nets deep learning research papers pdf
neural network reading deep learning survey

“I want to call SVM shallow learning”
-- Geoffrey Everest Hinton

Deep Learning Became Hot

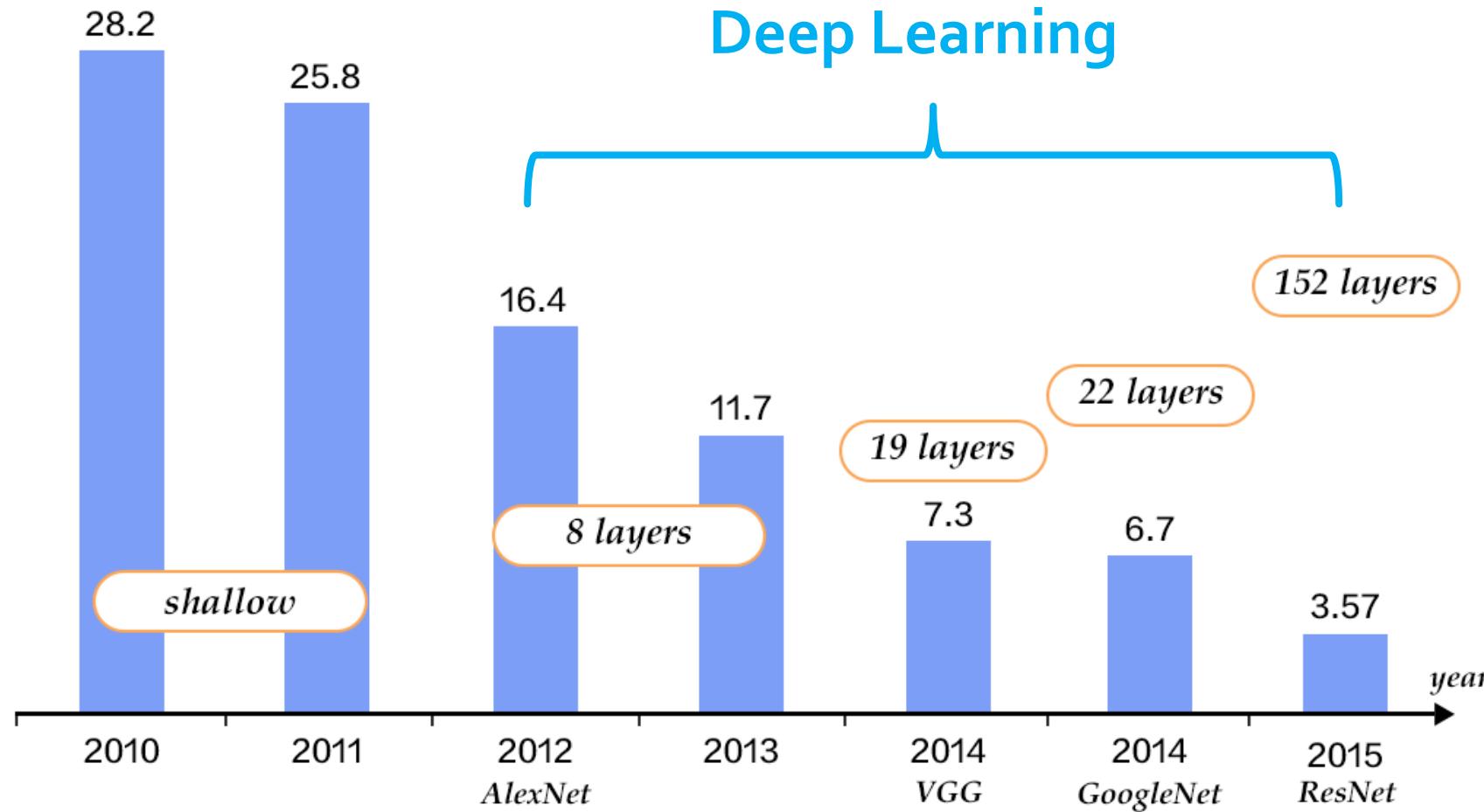


2016
MARCH
DeepMind's
AlphaGo, using
deep learning,
defeats world
champion **Lee
Sedol** in the
Chinese game of
go, four games to
one.



<http://fortune.com/ai-artificial-intelligence-deep-machine-learning/>

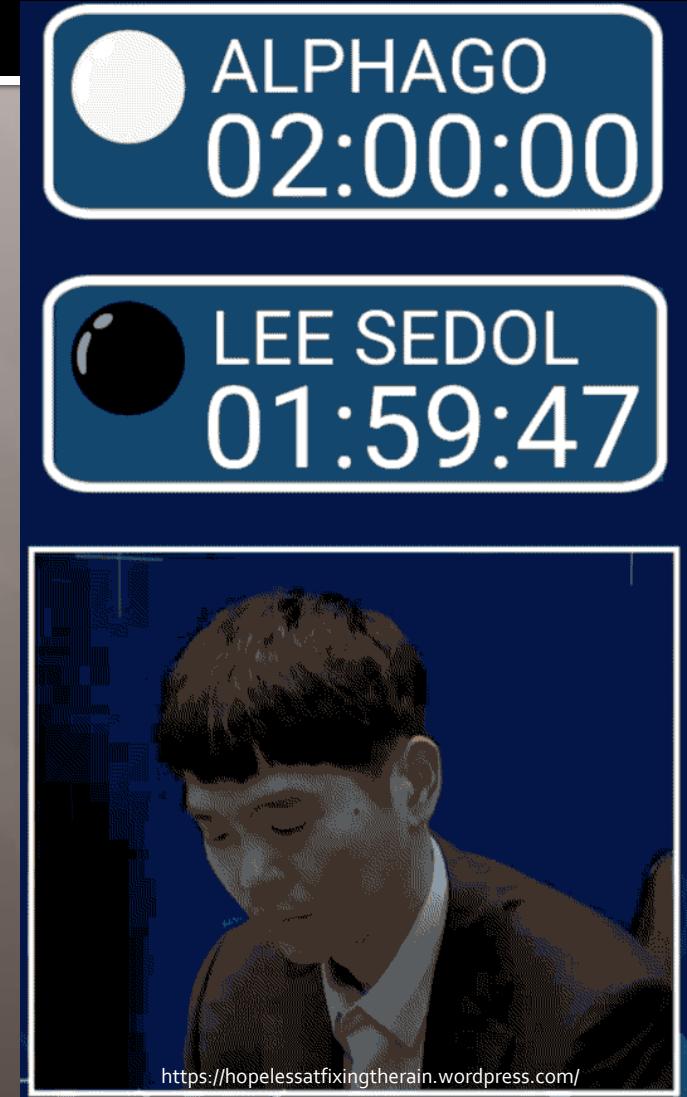
Performance on ImageNet



Top 5 error rate:
Computer > 25%
Human ~ = 5%

AlexNet
Computer ~ = 16 %

Great Marking!



Success on Vision



mite

container ship

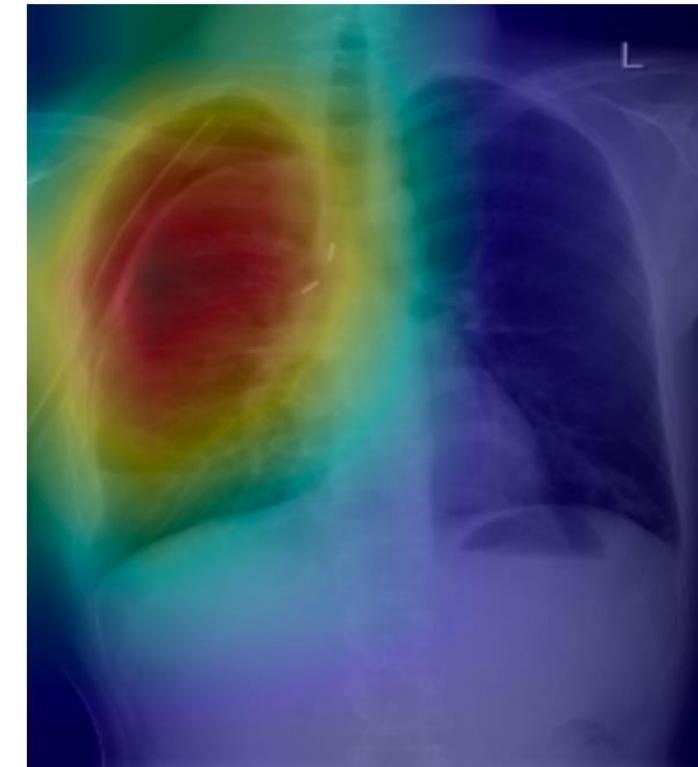
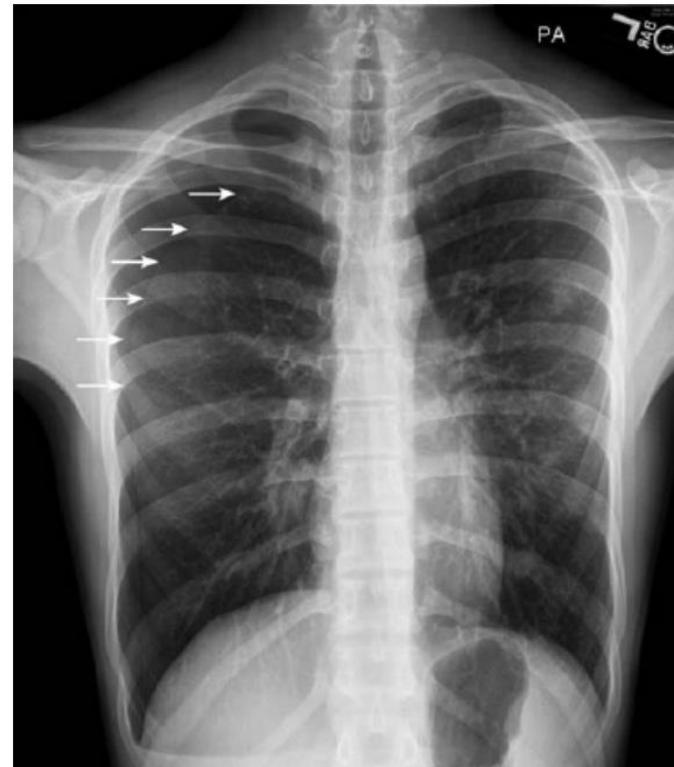
motor scooter

leopard

mite	black widow	lifeboat	motor scooter	leopard
cockroach	amphibian	fireboat	go-kart	jaguar
tick	drilling platform	bumper car	moped	cheetah
starfish		golfcart		snow leopard
				Egyptian cat

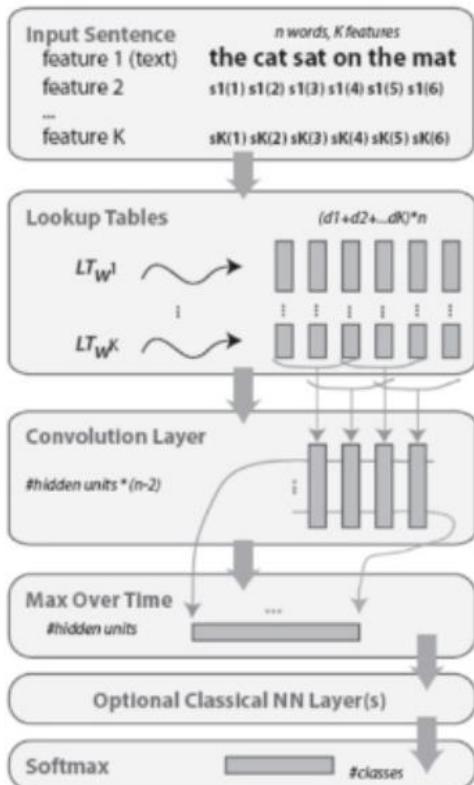
Success on Medical Imaging

Detect pneumothorax in real X-Ray scans



Success on NLP

General Deep Architecture for NLP



Basic features

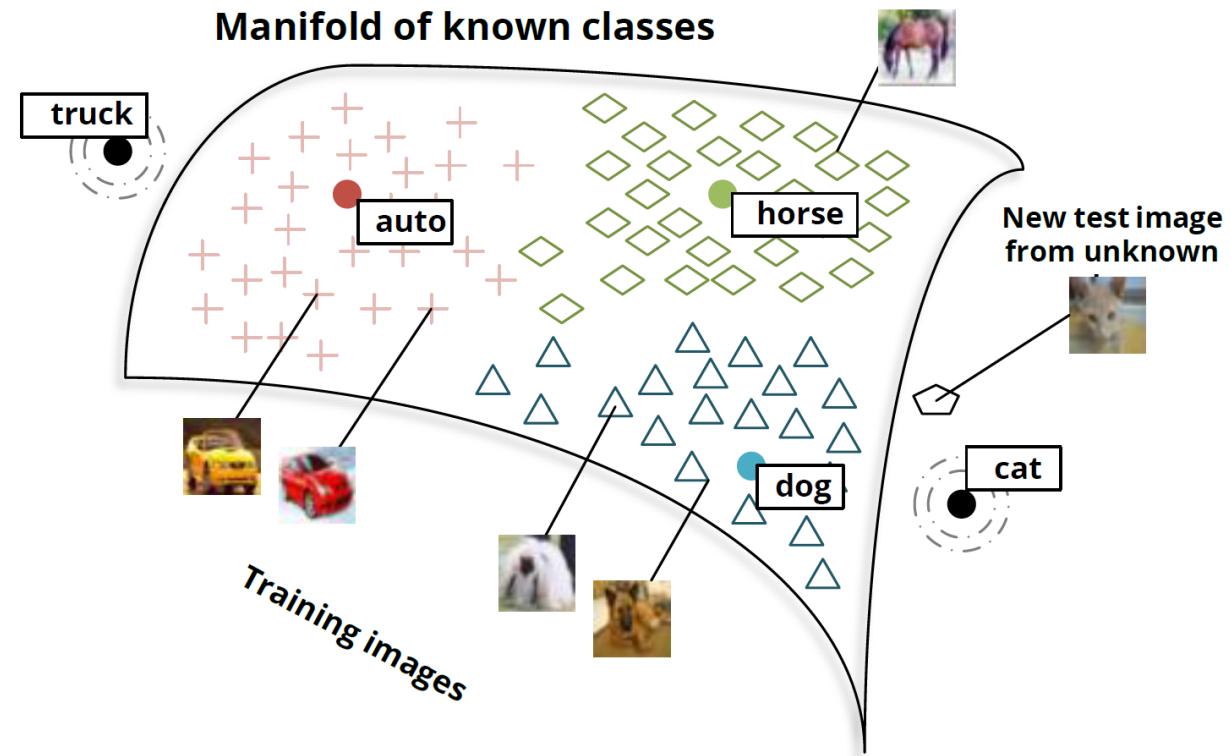
Embeddings

Convolution

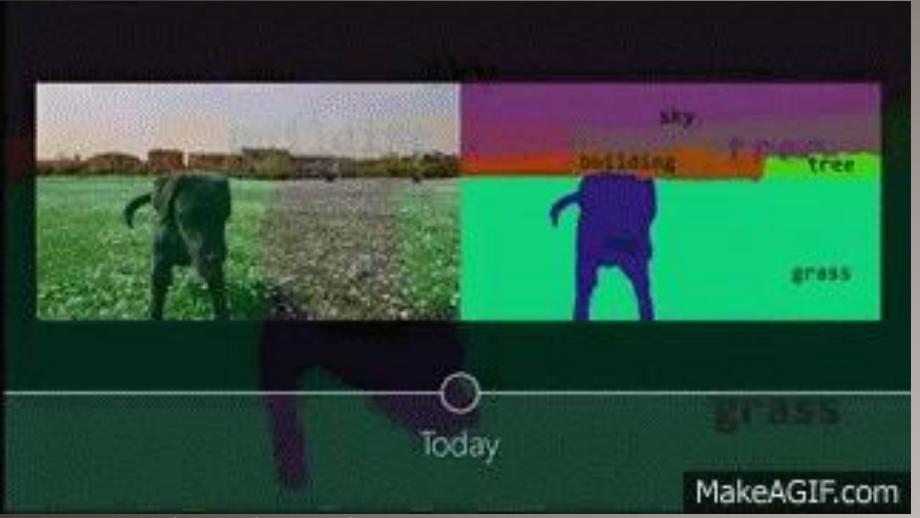
Max pooling

“Supervised” learning

source: Collobert & Weston, Deep Learning for Natural Language Processing. 2009 Nips



<http://colah.github.io/posts/2014-07-NLP-RNNs-Representations/>



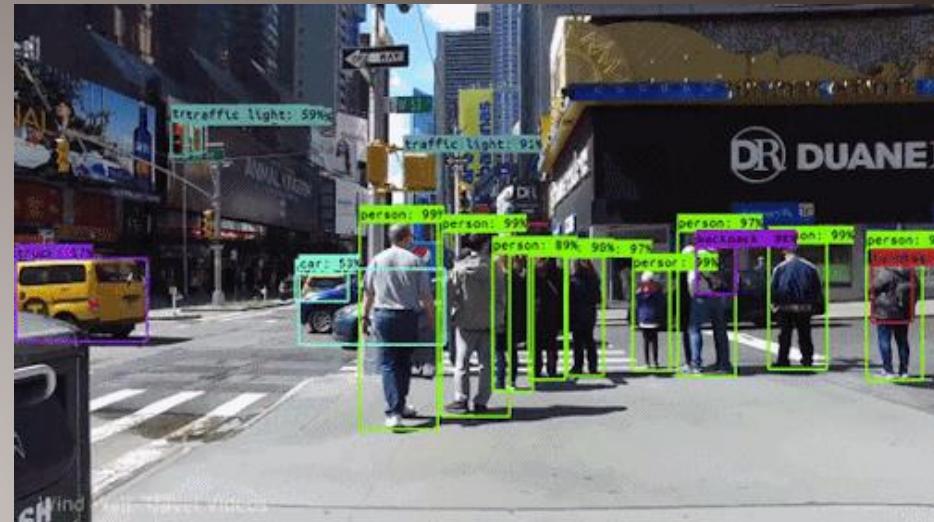
https://makeagif.com/gif/microsoft-deep-learning-semantic-image-segmentation-G_HiNa



<http://www.thesarahshow.com/>



<https://gifer.com/en/SRL8>



<https://mc.ai/deep-learning-for-object-detection-from-the-start-to-the-state-of-the-art-2-2/>



<https://www.theverge.com/2016/2/29/11133682/deep-learning-ai-explained-machine-learning>



<http://gizmodo.com/researchers-just-created-the-most-amazing-lip-reading-s-1788748163>



<https://www.technologyreview.com/s/542921/robot-toddler-learns-to-stand-by-imagining-how-to-do-it/>



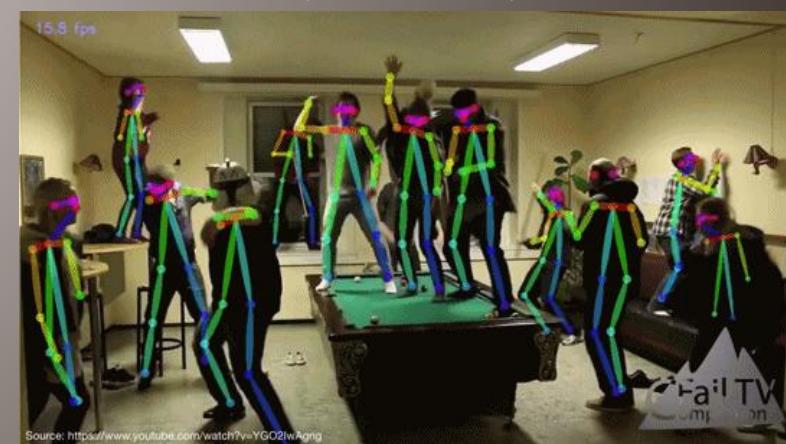
<https://people.eecs.berkeley.edu/~pathak/>



<https://www.instadeep.com/>



<https://kotaku.com/blizzard-wants-to-know-if-googles-deepmind-ai-can-conqu-1788615284>



<http://d.hatena.ne.jp/hanecci/20170513/p8>

Over Marketing



<http://do-androidsdreamof-electricsheep.tumblr.com/post/47576669265/ai-artificial-intelligence-2001>

In Public & Media



https://www.reddit.com/r/reactiongifs/comments/7077cb/mrw_i_read_the_news_that_the_us_will_end_the/

AI Scientists

Still Limited

Emergent Tech ▶ Artificial Intelligence

OpenAI bots smashed in their first clash against human Dota 2 pros

AI can react faster than humans, but don't play well enough to beat the masters yet

By Katyanna Quach 23 Aug 2018 at 02:19

1 SHARE ▼



<http://wqo326.tumblr.com/post/93428031577/push-until-the-end-ti4-newbee-vs-eg>

Two Attitudes

Optimist

The glass is half-full,
we just keep on
drinking, when it's
empty, it's empty.



Pessimist

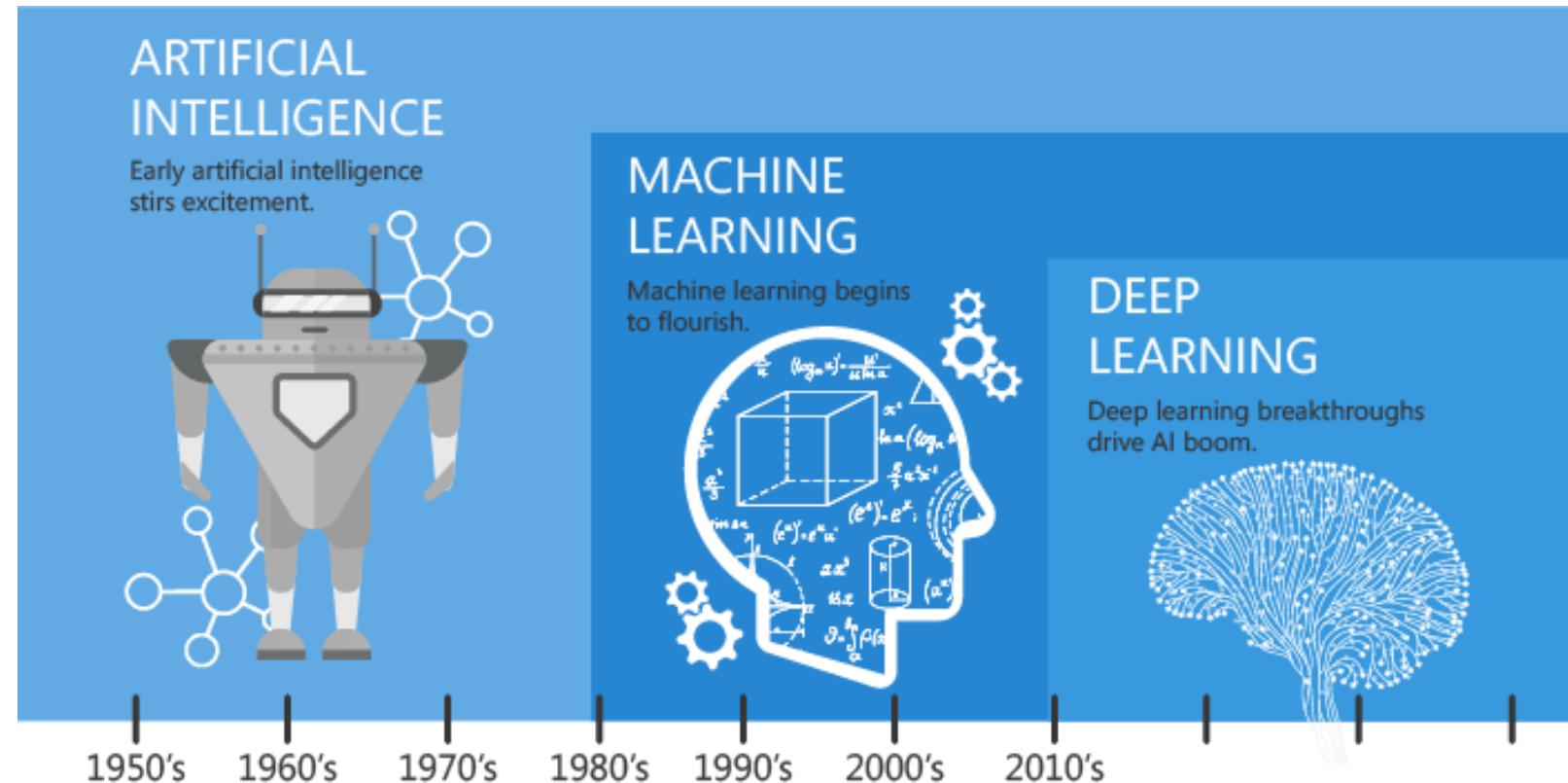
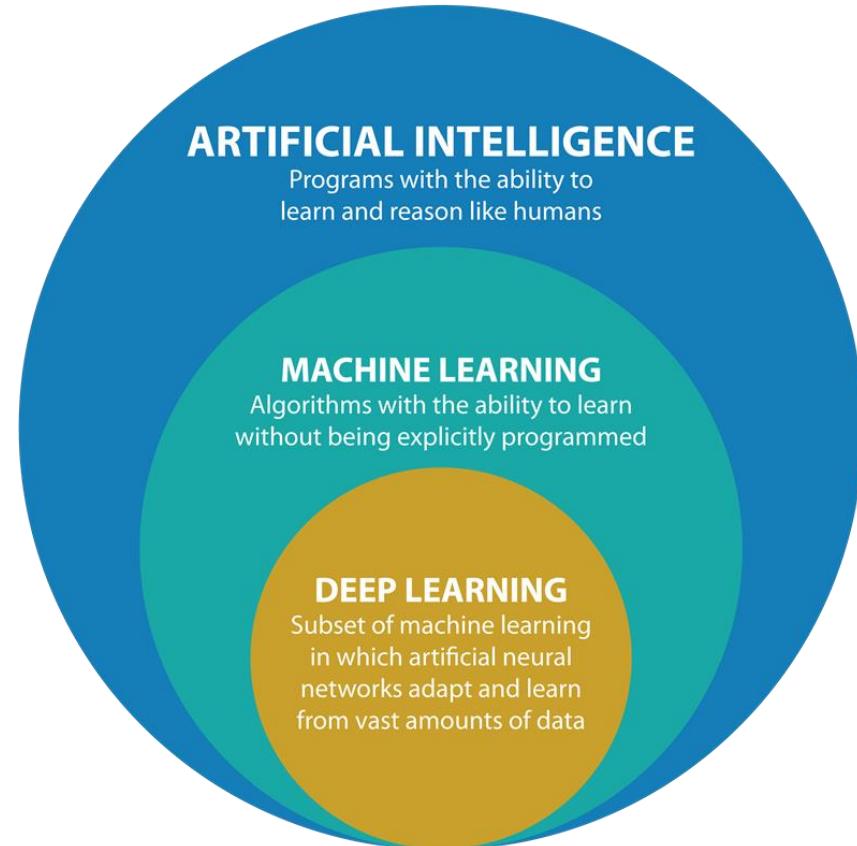
The glass is half-empty,
we should take it slow
and really enjoy our
drink.

The Realist looks at it both ways.

Concept & Illustration: Sietse Visser - Zietze.nl - 2011

<https://visual.ly/community/infographic/humor/optimist-or-pessimist>

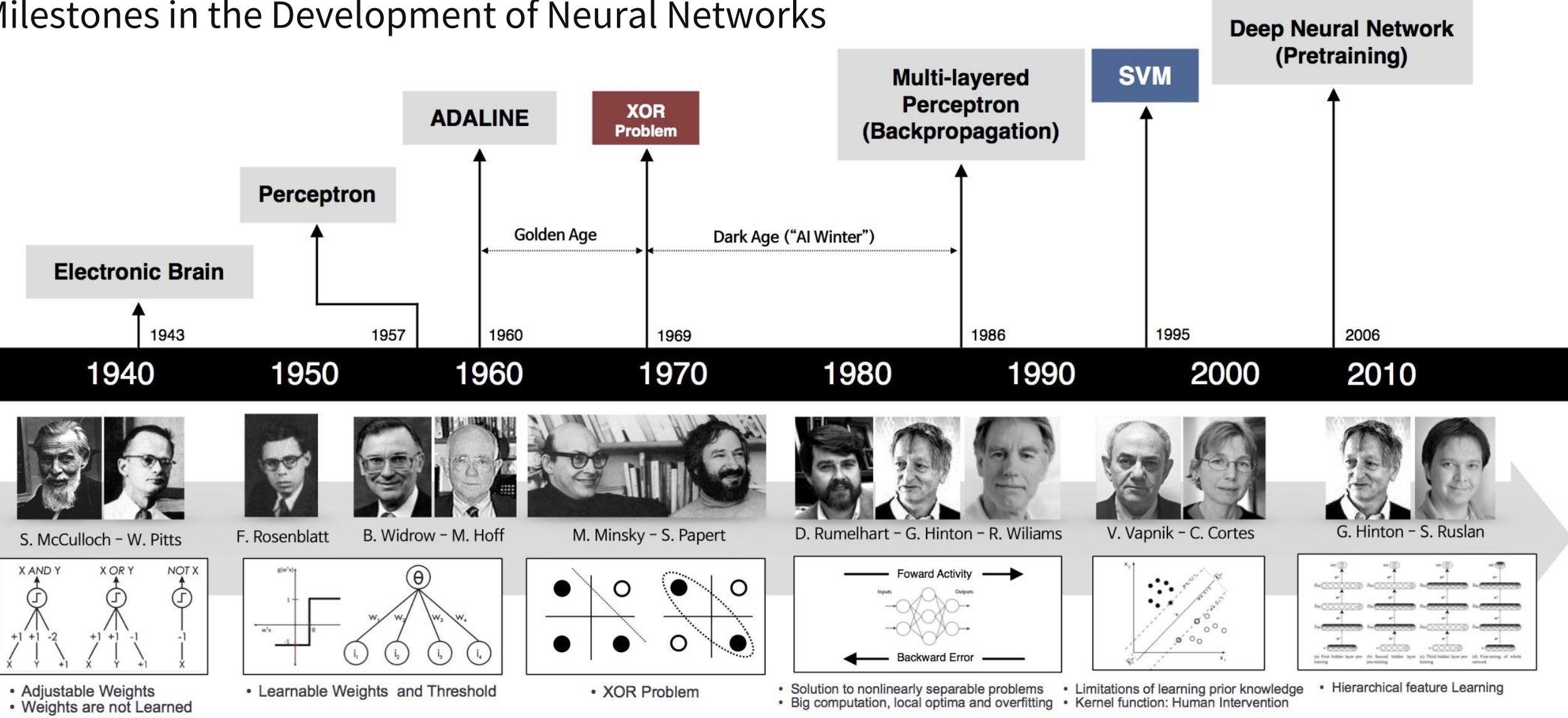
Deep Learning, Machine Learning, AI



Since an early flush of optimism in the 1950's, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.

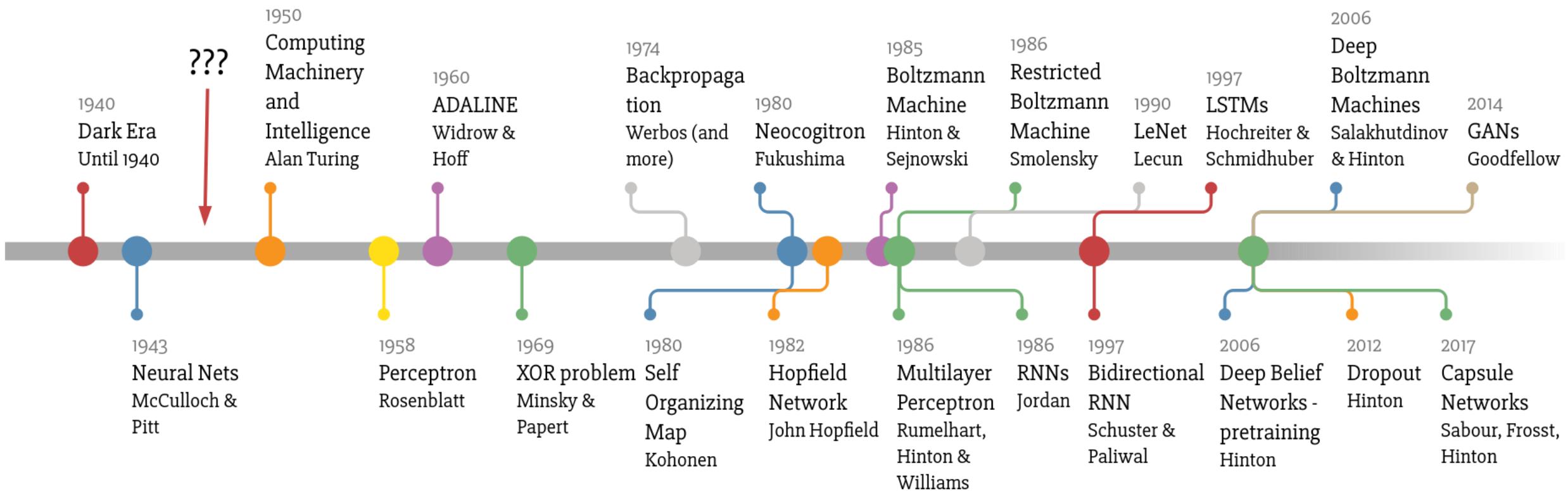
<http://blog.aimagnifi.com/index.php/2017/10/13/what-is-the-difference-between-machine-learning-and-deep-learning/>

Milestones in the Development of Neural Networks



https://beamandrew.github.io/deeplearning/2017/02/23/deep_learning_101_part1.html

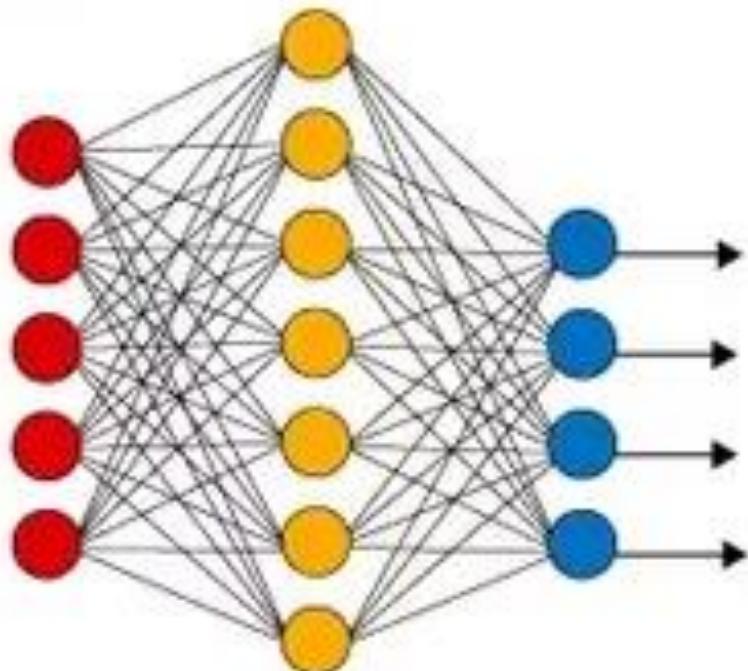
Deep Learning Timeline



<https://towardsdatascience.com/a-weird-introduction-to-deep-learning-7828803693bo?gi=6a575eef7393>

Deep Learn = Deep Neural Network

Simple Neural Network

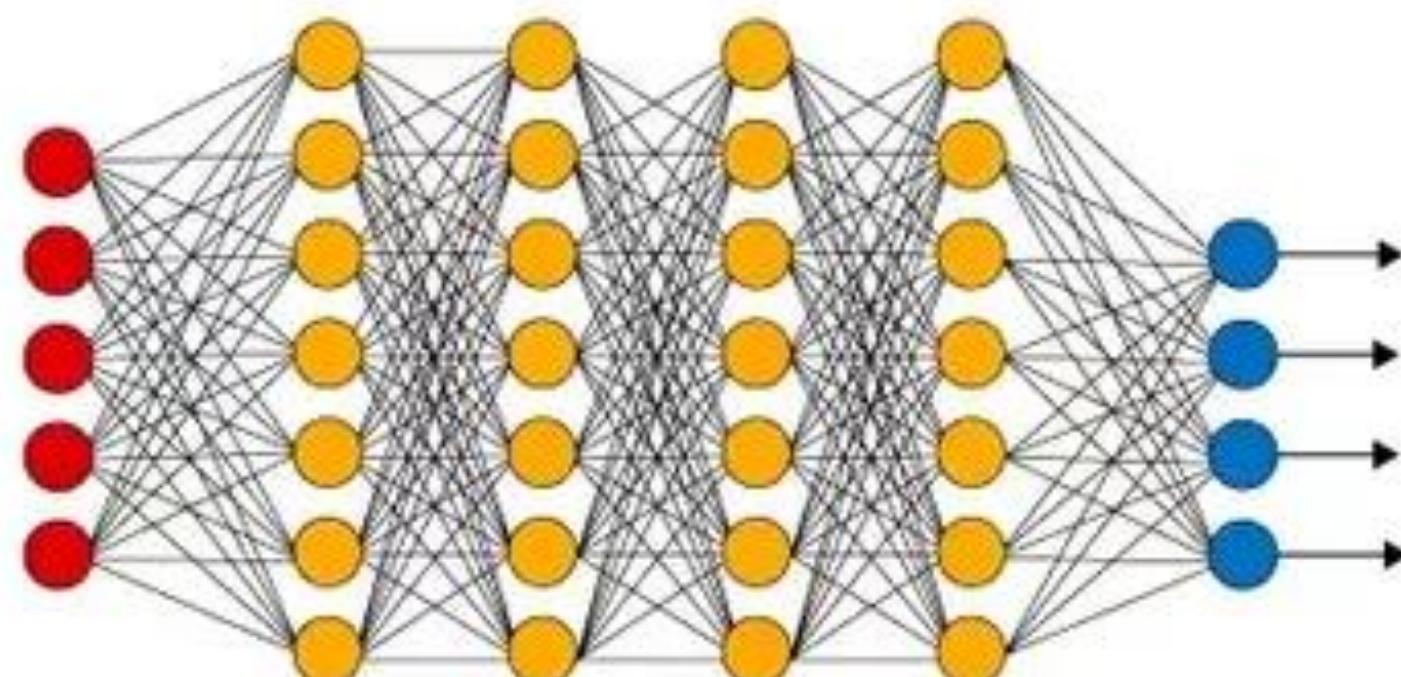


● Input Layer

○ Hidden Layer

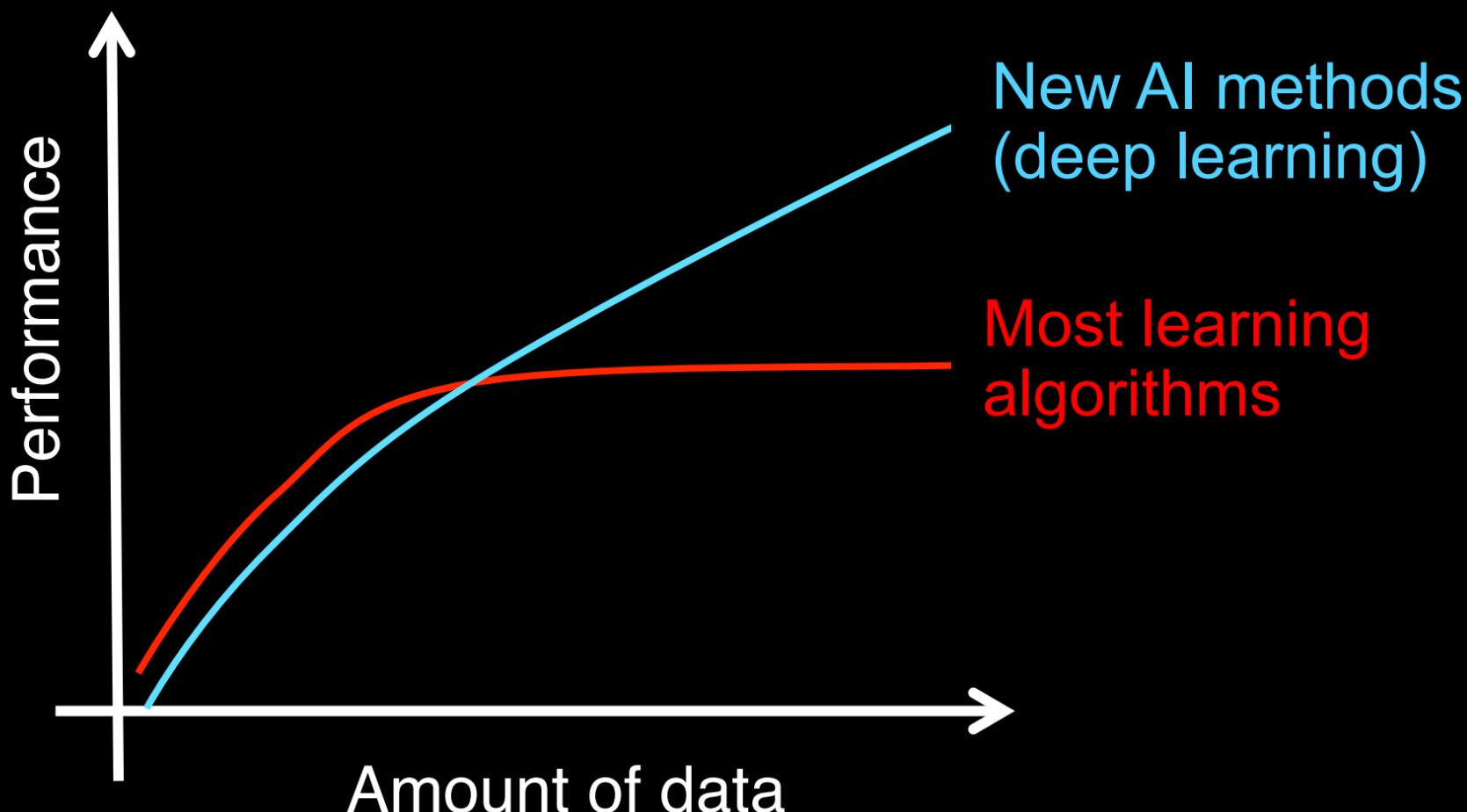
● Output Layer

Deep Learning Neural Network

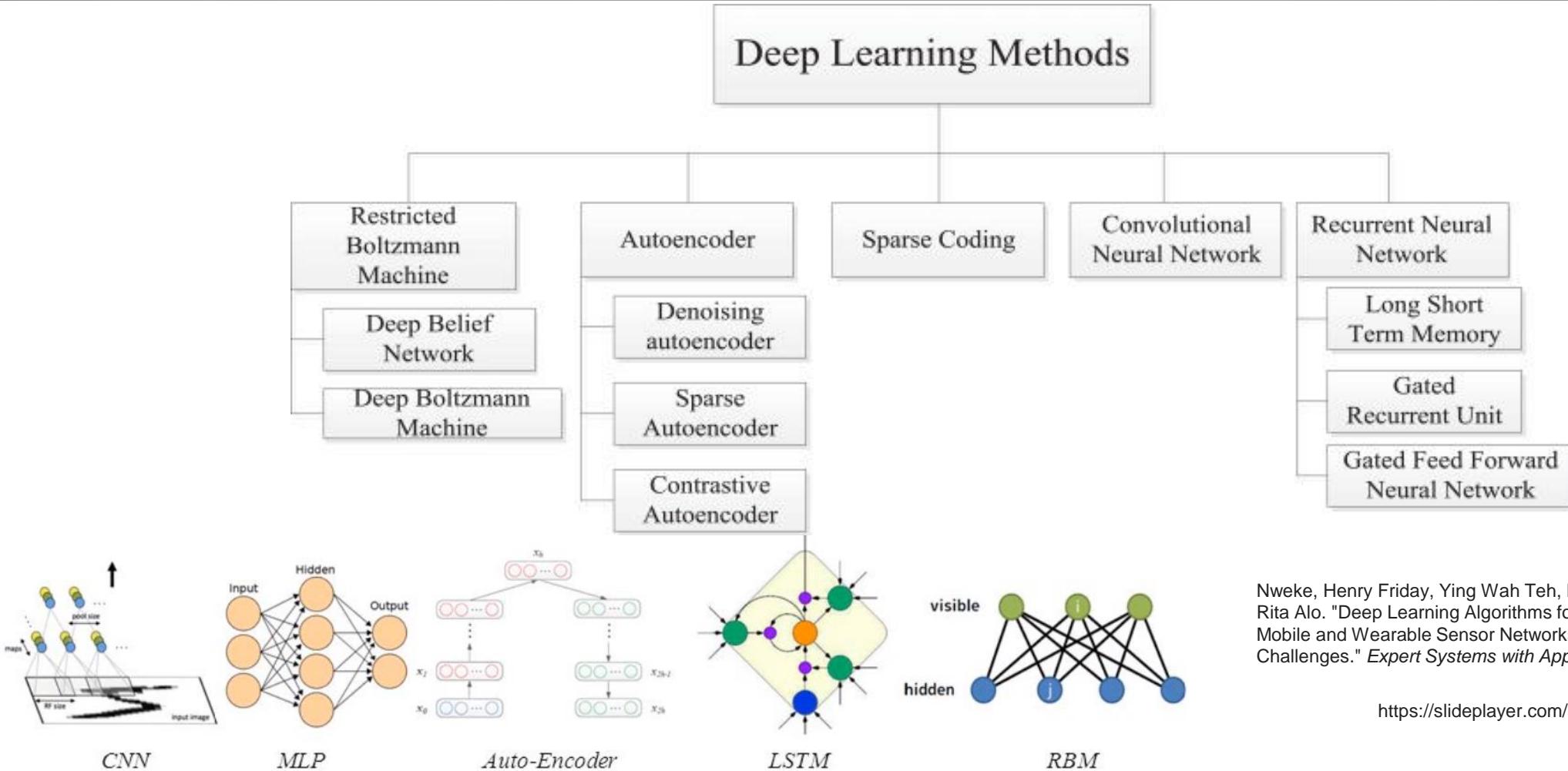


<https://www.quora.com/What-is-the-difference-between-Neural-Networks-and-Deep-Learning>

Advantages



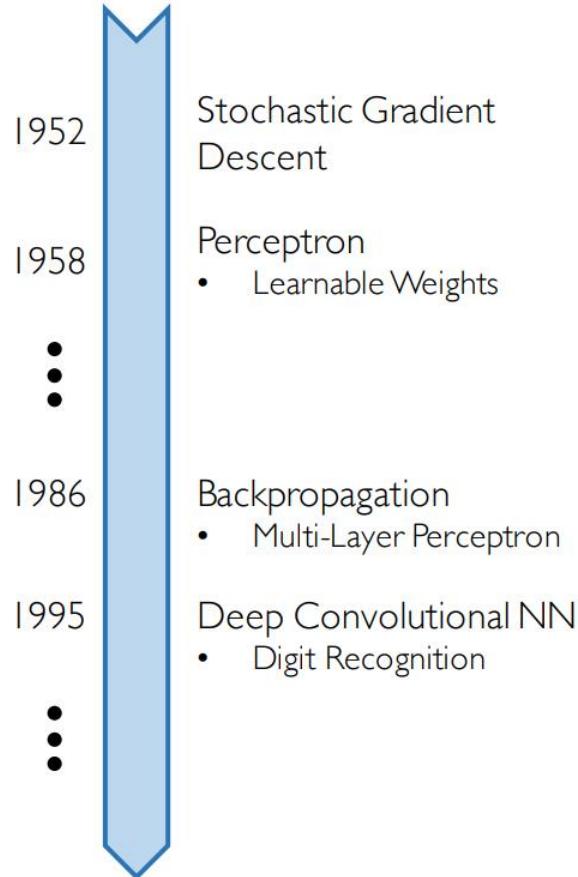
Overview of Method



Nweke, Henry Friday, Ying Wah Teh, Mohammed Ali Al-Garadi, and Uzoma Rita Alo. "Deep Learning Algorithms for Human Activity Recognition using Mobile and Wearable Sensor Networks: State of the Art and Research Challenges." *Expert Systems with Applications* (2018).

<https://slideplayer.com/slide/7483751/>

New but also Old



Nothing new !

- **Alexnet 2012**

- **based on CNN (LeCunn, 1989)**

- **Alpha Go**

- **based on Reinforcement learning and MCTS (Sutton, 1998)**

<https://elearningindustry.com/artificial-intelligence-will-shape-elearning>

njkim@Jamonglab.com, 2016

Ian Goodfellow, Deep Learning (MIT press, 2016).

Why Now?

I. Big Data

- Larger Datasets
- Easier Collection & Storage

IM³GENET



WIKIPEDIA
The Free Encyclopedia



2. Hardware

- Graphics Processing Units (GPUs)
- Massively Parallelizable

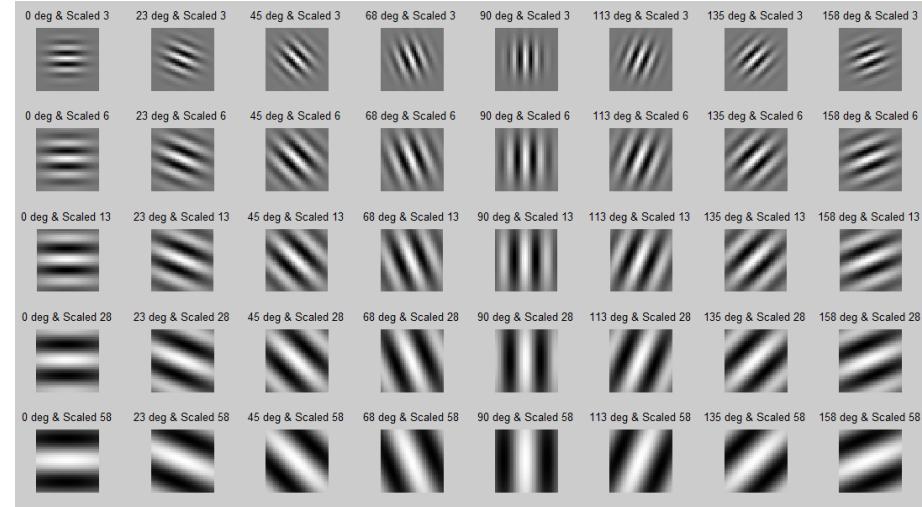


3. Software

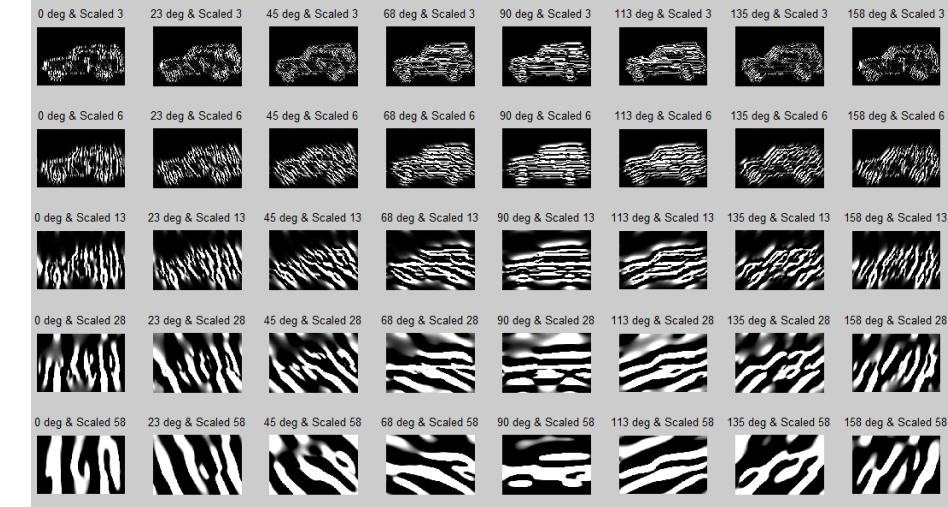
- Improved Techniques
- New Models
- Toolboxes



Handcrafted Feature



Gabor Filter

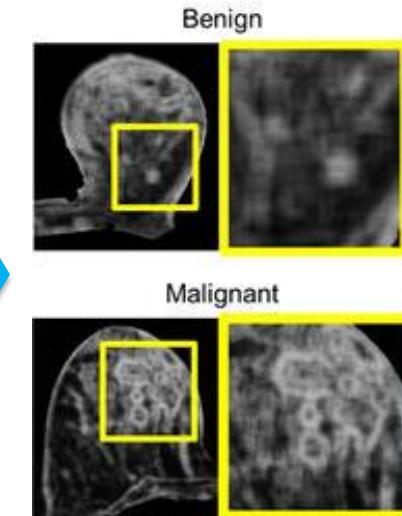
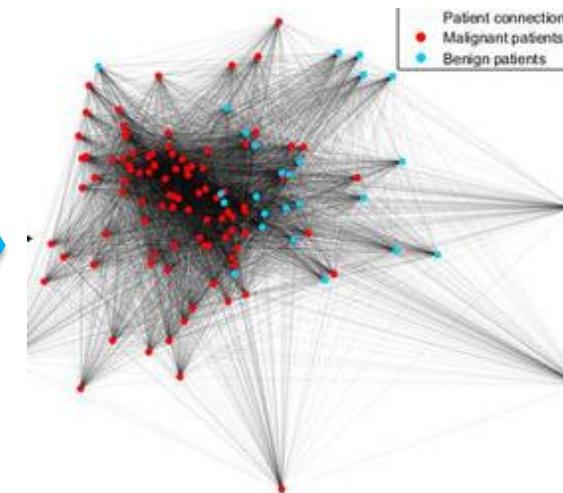
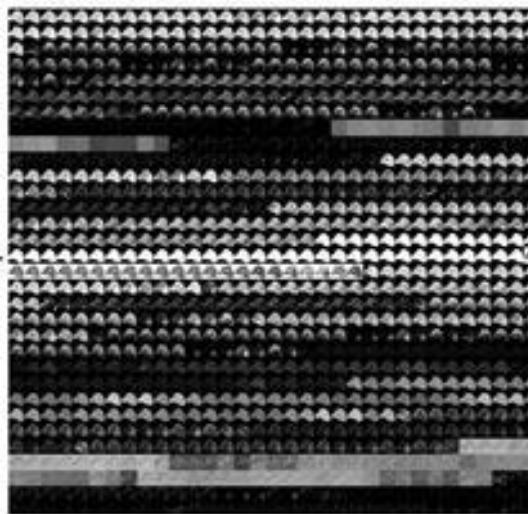
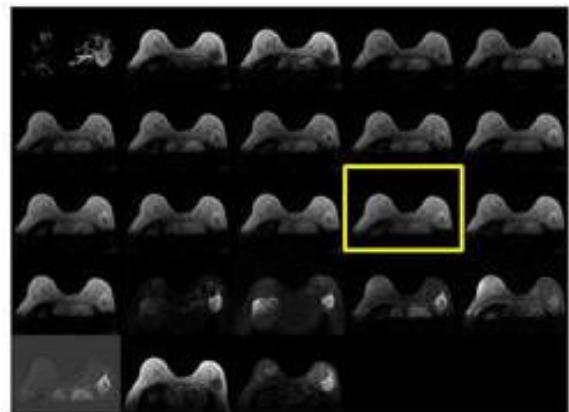


Features

Not as good as human's performance

Traditional Framework of Understanding Medical Image

Parekh et al., NPJ breast Cancer, 2017



Imaging Data

- MRI T1w
- MRI T2w
- CT

...

Feature Extraction

- Gabor
- Handcrafted Features
- Histogram
- Fourier

...

Classification

- SVM
- Shallow Models
- Network
- Random Forest

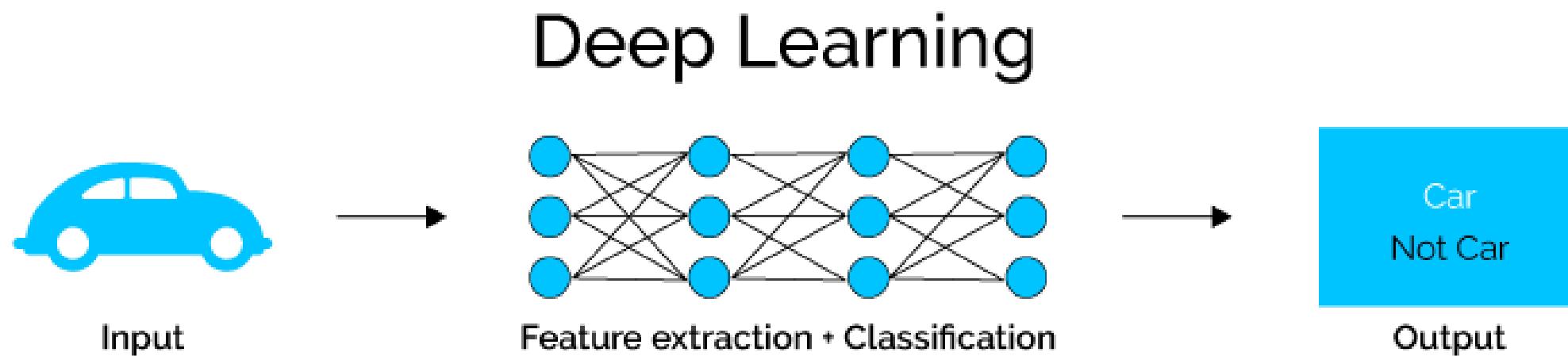
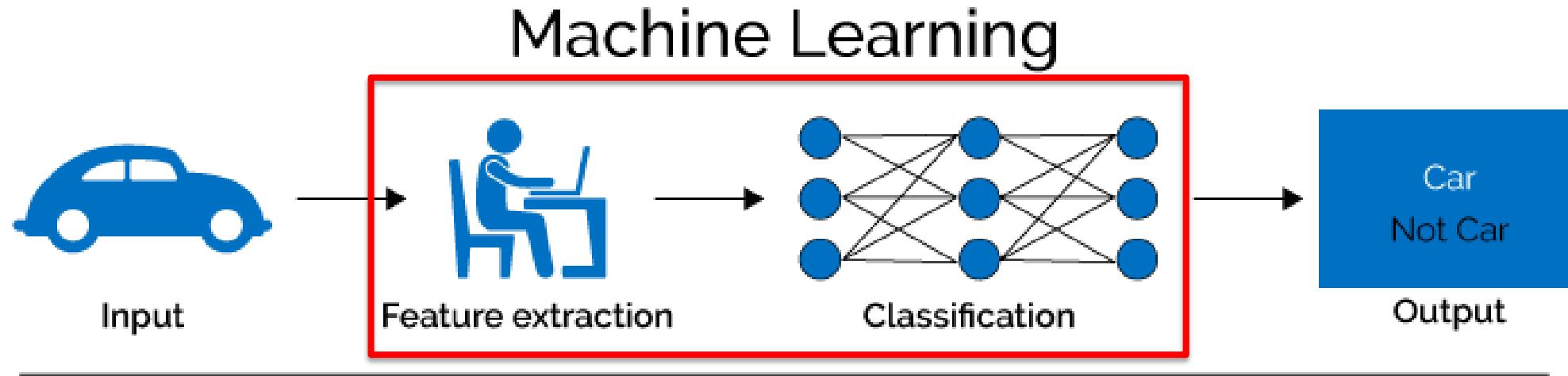
...

Results

- Classification
- Detection
- Segmentation

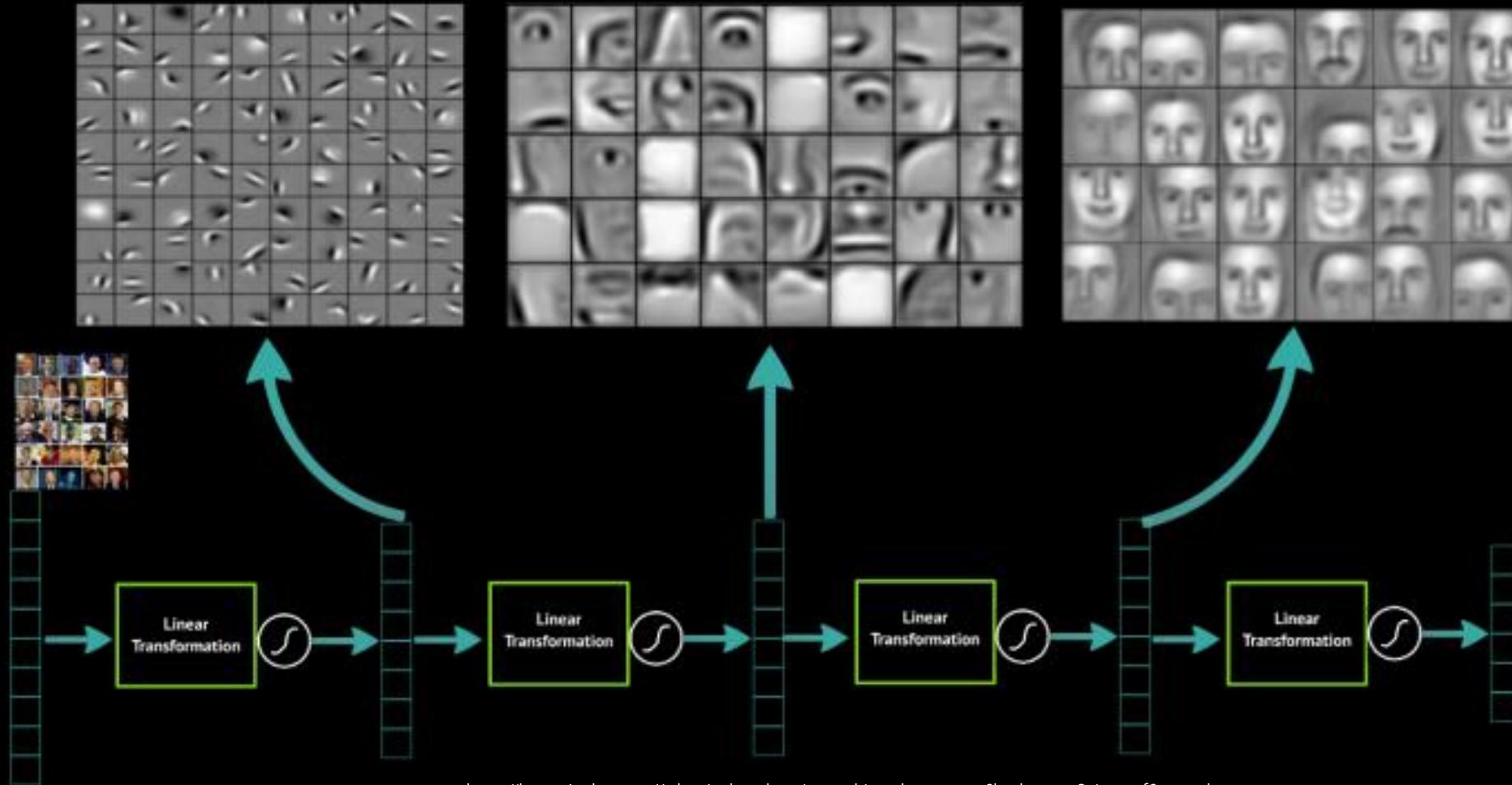
...

Deep Learning vs. Machine Learning



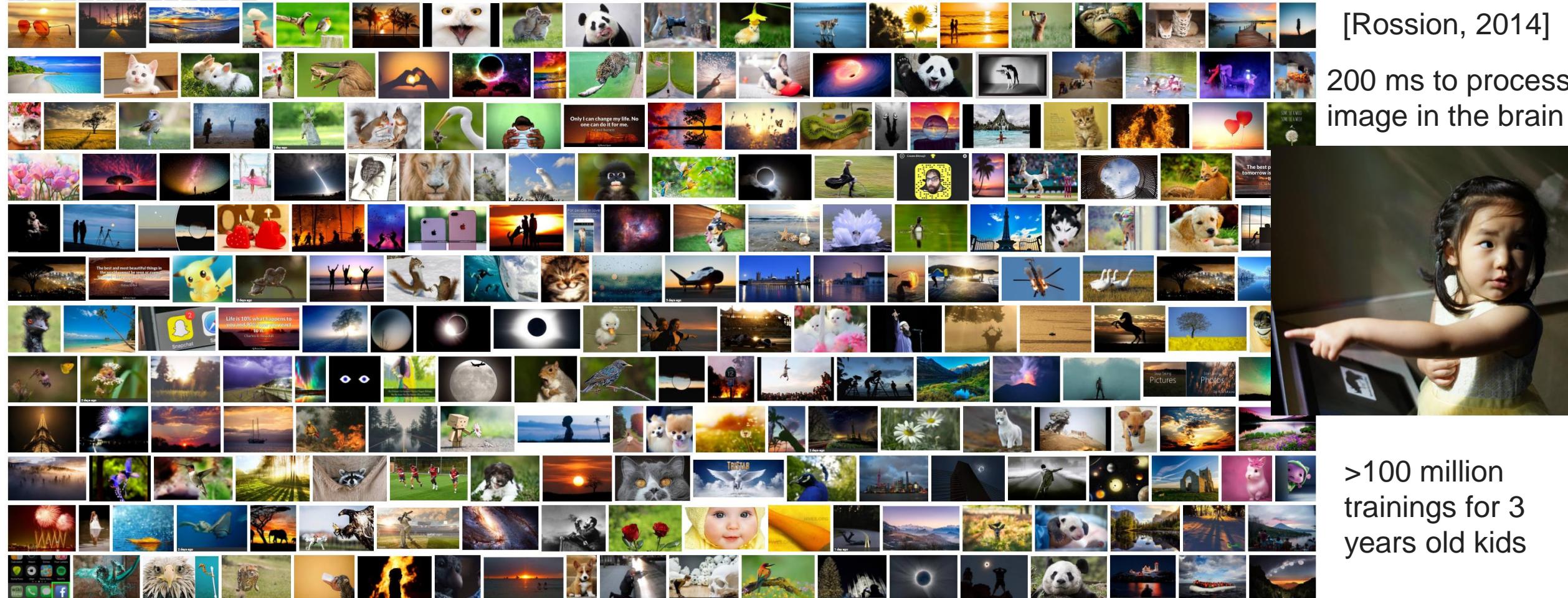
<https://www.quora.com/How-do-I-start-learning-machine-learning-and-deep-learning-using-C++>

Deep Features Learned by Computer



<https://becominghuman.ai/what-is-deep-learning-and-its-advantages-16b74bc541a1?gi=7220f649e19d>

Miss Another Reasons



[Rossion, 2014]

200 ms to process
image in the brain

>100 million
trainings for 3
years old kids



ImageNet

2007

Fei-Fei Li founds ImageNet and begins assembling a database of 14 million labeled images that can be used for machine-learning research. →



IMAGENET

- 1,000 object classes (categories).
- Images:
 - 1.2 M train
 - 100k test.



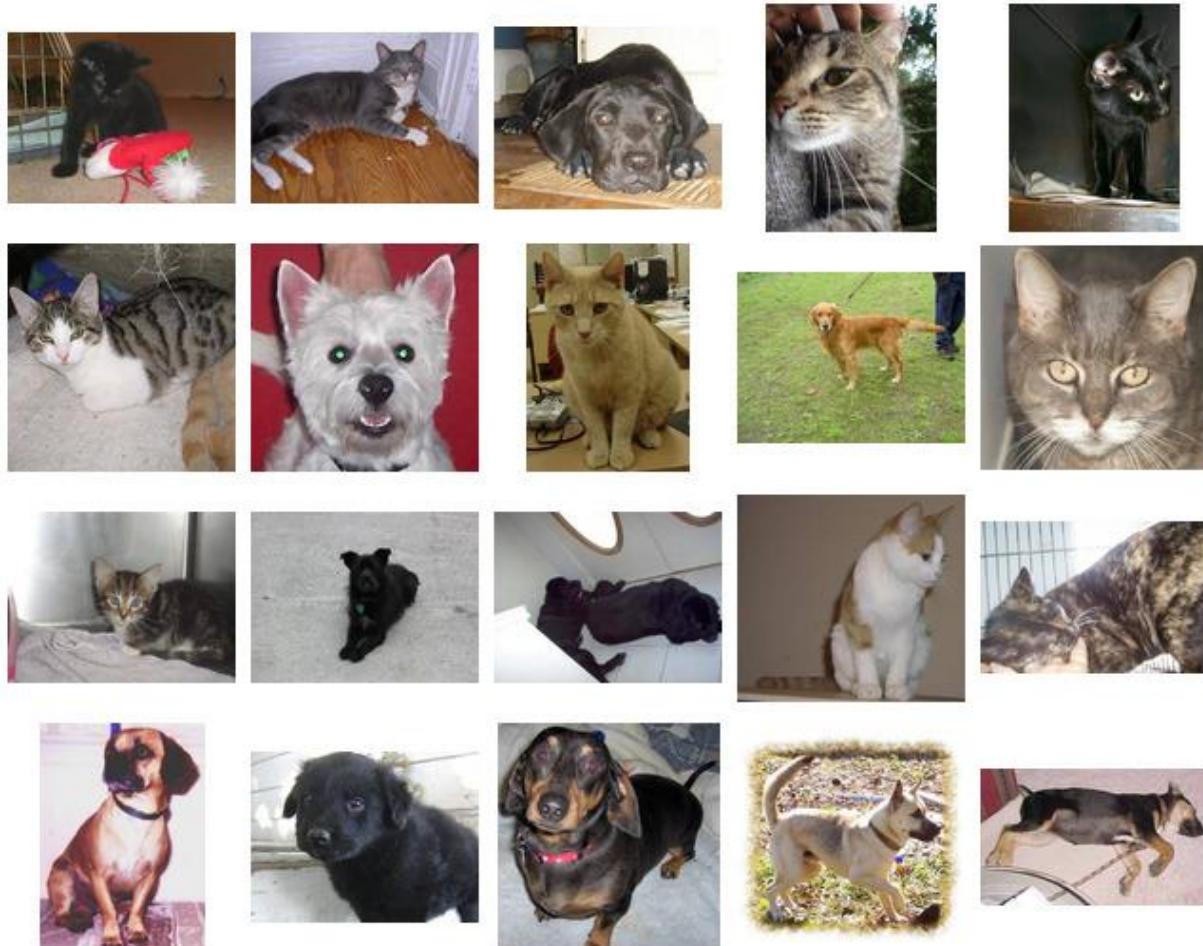
ImageNet Dataset

IMAGENET



Russakovsky, O., Deng, J., Su, H., Krause, J., Satheesh, S., Ma, S., ... & Fei-Fei, L. (2015). [Imagenet large scale visual recognition challenge](#). arXiv preprint arXiv:1409.0575. [\[web\]](#)

Amazing Human Vision

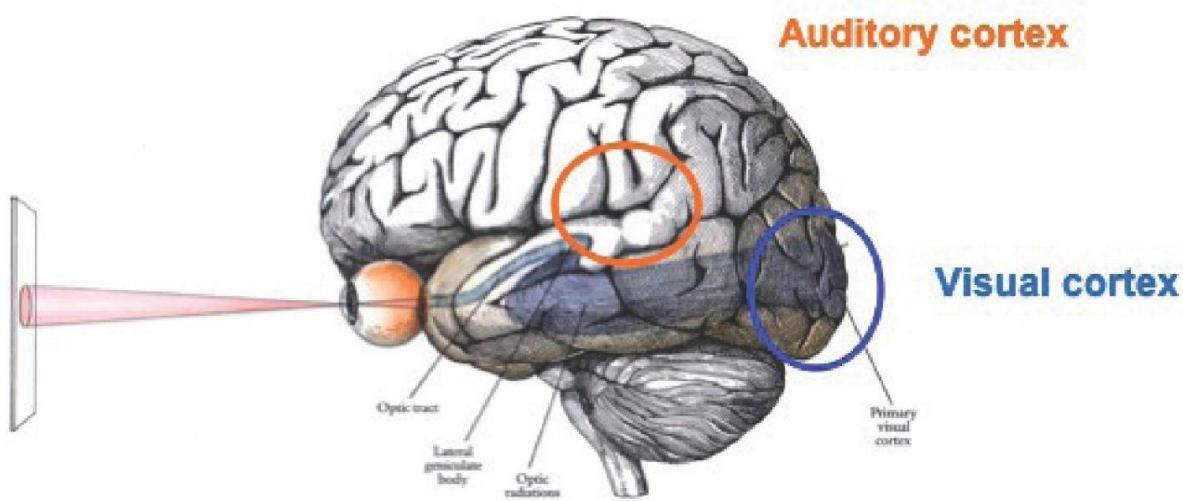


<https://www.pyimagesearch.com/2016/08/08/k-nn-classifier-for-image-classification/>

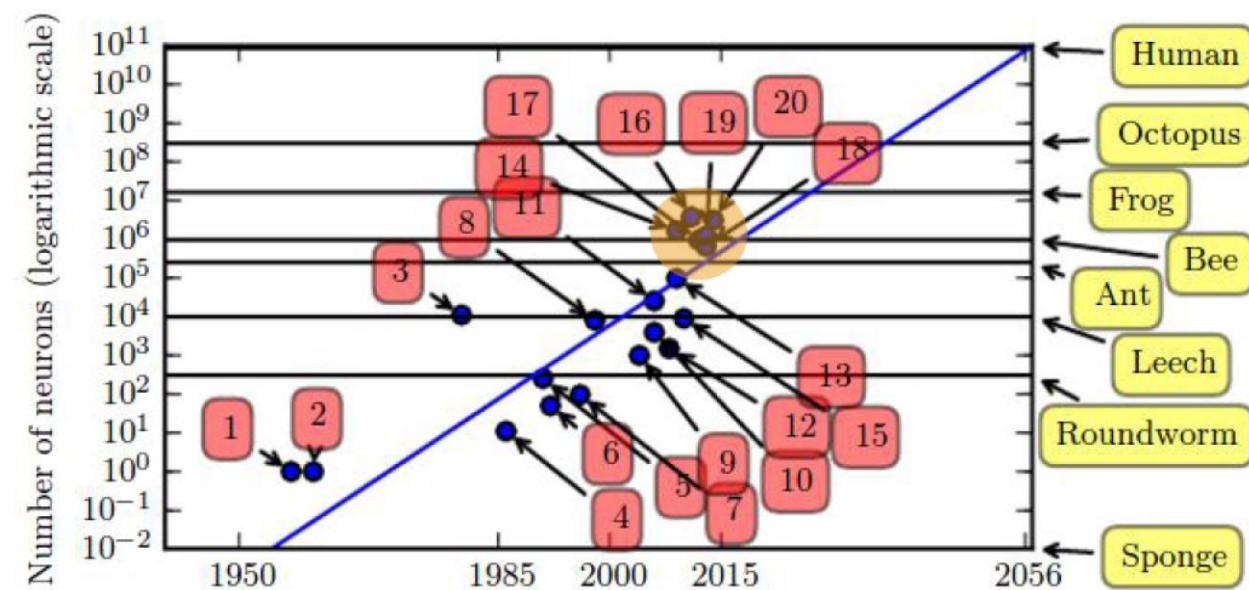


<https://www.bcm.edu/healthcare/care-centers/diagnostics>
<https://www.chestercountyhospital.org/services-and-treatments/radiology>

Why Human is Better?



<http://what-when-how.com/wp-content/uploads/2012/05/tmpD1.jpg>



Ian Goodfellow, Deep Learning (MIT press, 2016).

Super Hot in Healthcare



AI IS STORMING THROUGH THE \$6.5 TRILLION HEALTHCARE INDUSTRY

2

MICCAI 2017 brought together some of the world's best biomedical scientist, engineers and clinicians, and AI became a central theme for the medical imaging conference.

One sure sign of this: the number of AI-focused papers submitted by the health research community is surging. Of the 800 manuscripts submitted at MICCAI, 60 percent are focused on machine learning. And of those papers, 80 percent use deep learning

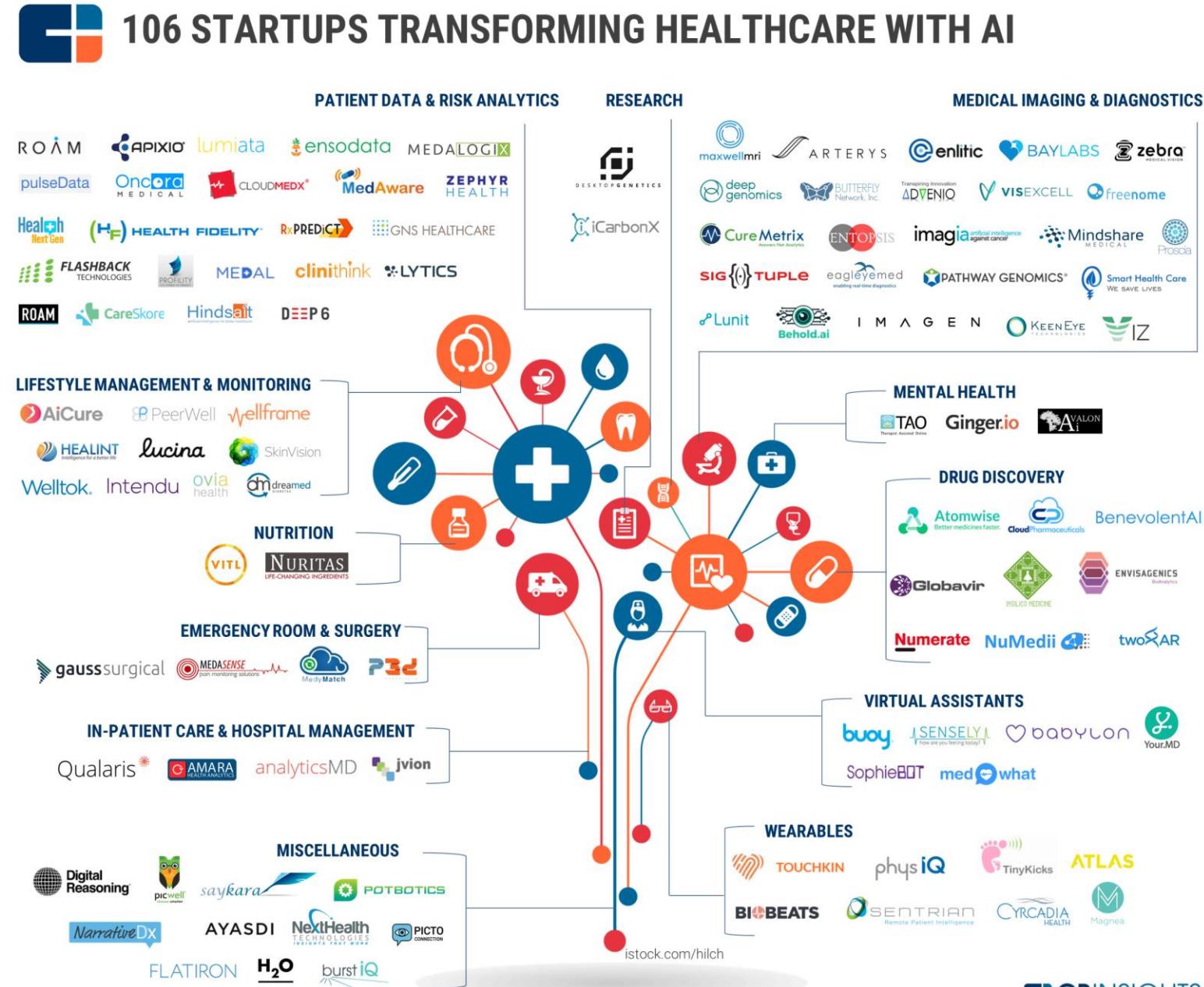
READ ARTICLE

Source : <https://www.slideshare.net/NVIDIA/top-5-deep-learning-and-ai-stories-september-22-2017>

Industry



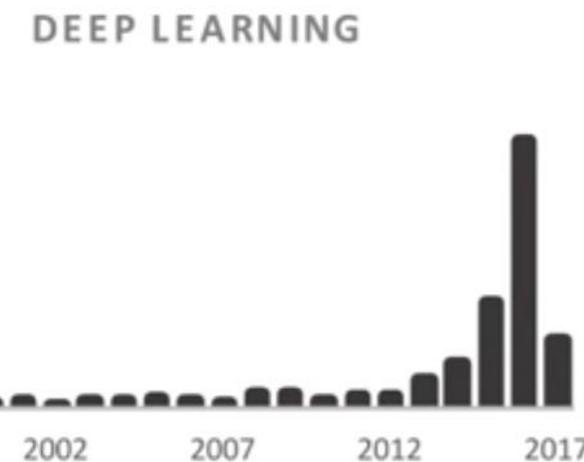
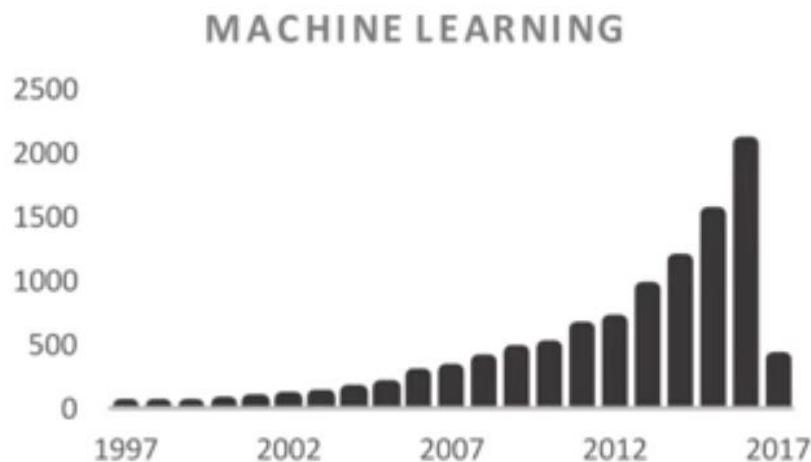
<https://triteq.com/news/10-largest-medical-device-companies-world/>



<https://www.cbinsights.com/research/artificial-intelligence-startups-healthcare/>

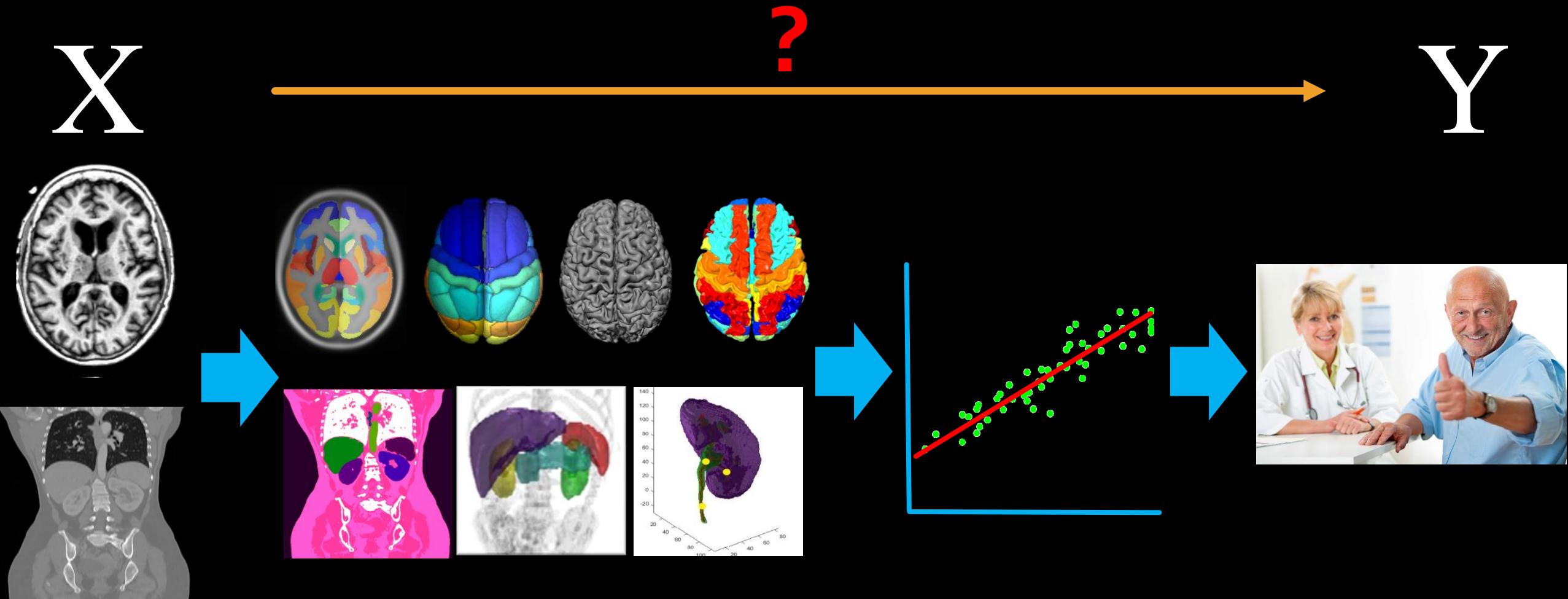
Academia

the rise of deep learning



Number of publications on PubMed containing these terms in title or abstract, queried 15/02/2017

Medical Image Analysis



Tasks In Medical Imaging

X → Y



Understand A Image

Understand A Region

Understand Every Pixel

Classification/Regression

Detection

Segmentation

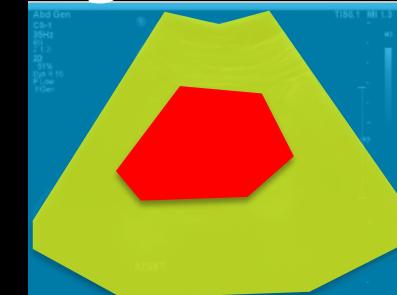
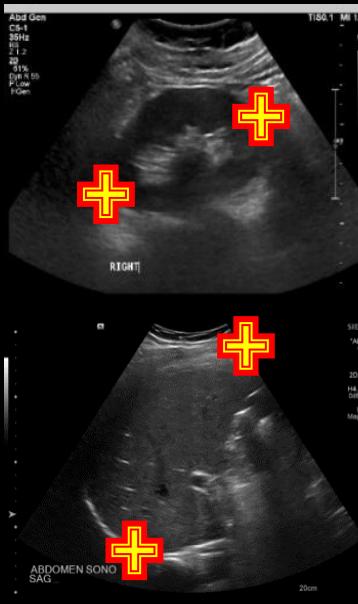
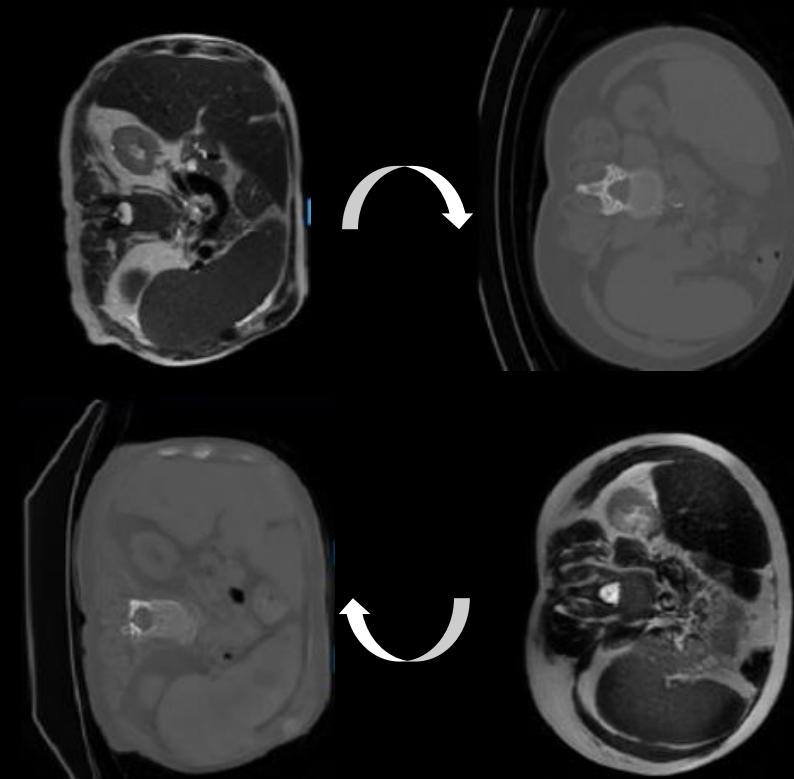
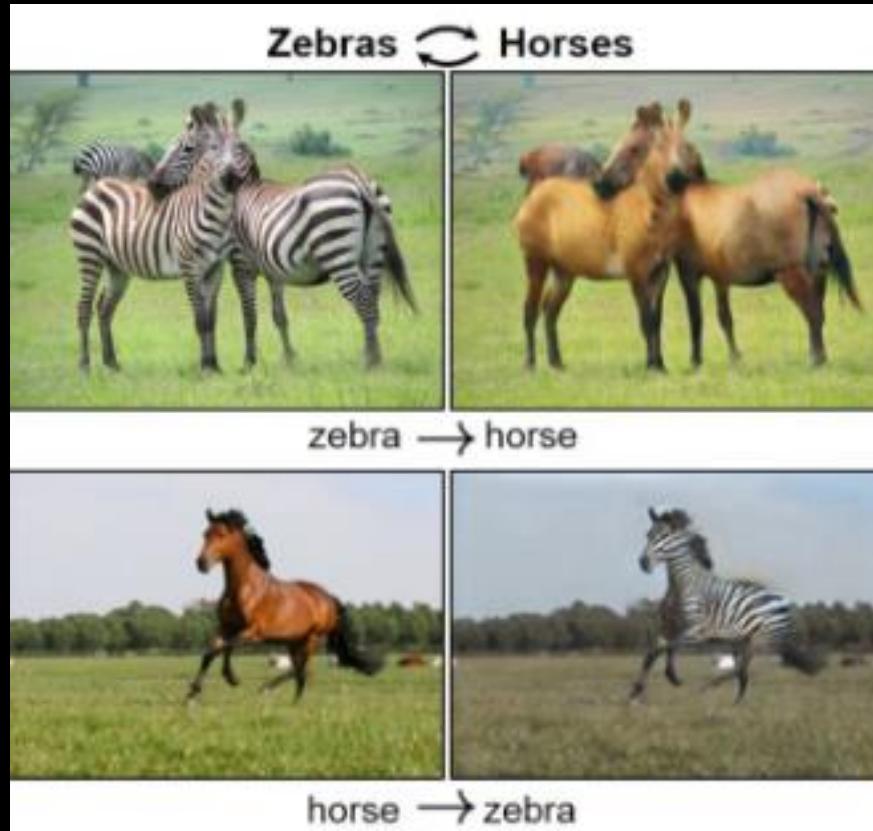


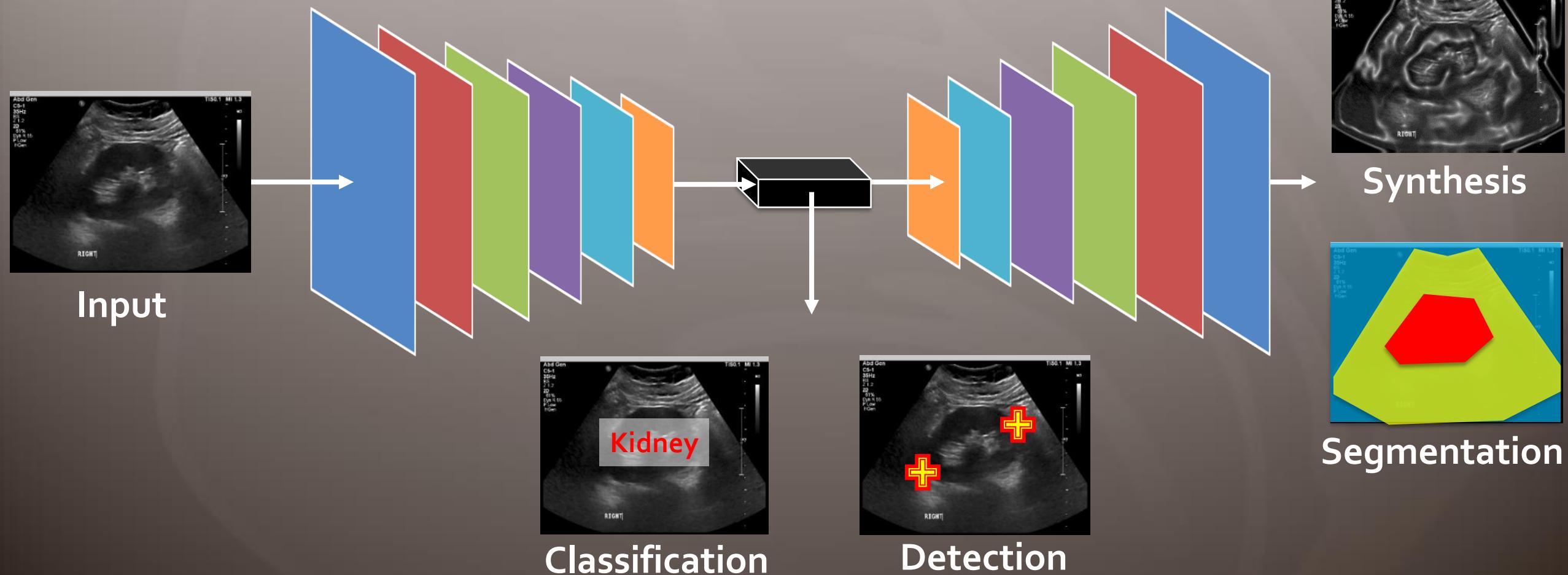
Image Synthesis Generative Adversarial Network (GAN)

X: CT/MRI —————→ Y: Synthesis/ Seg.

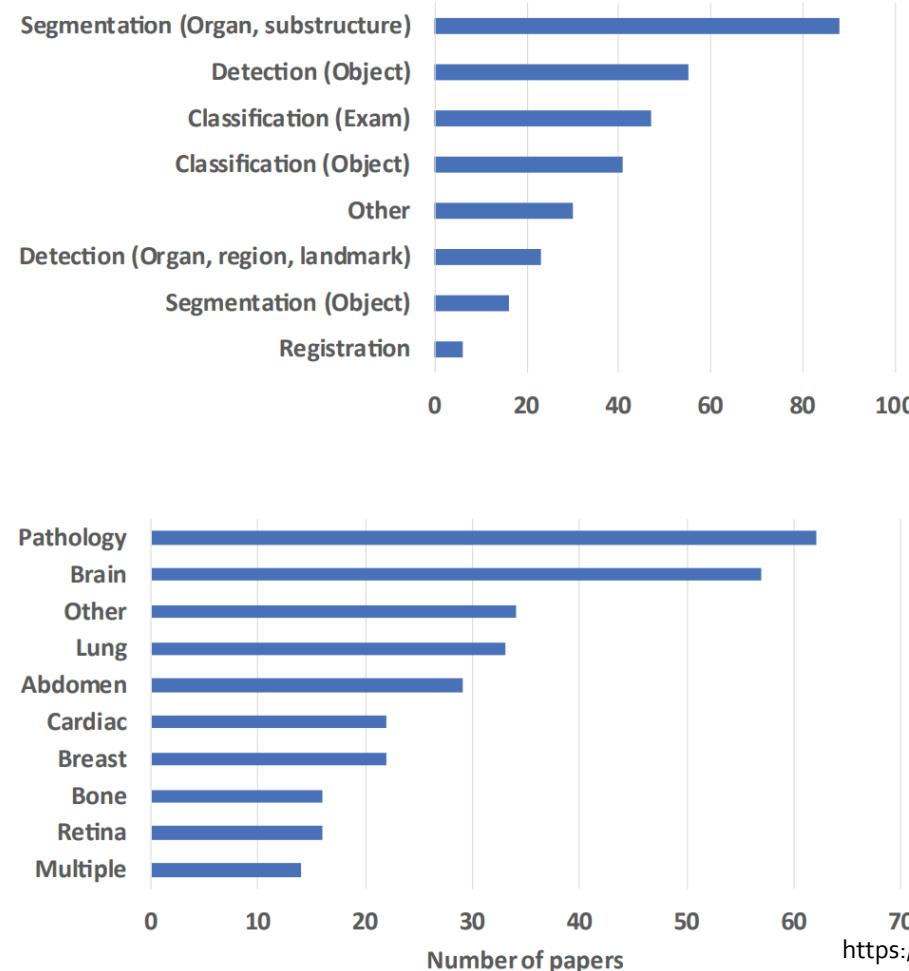
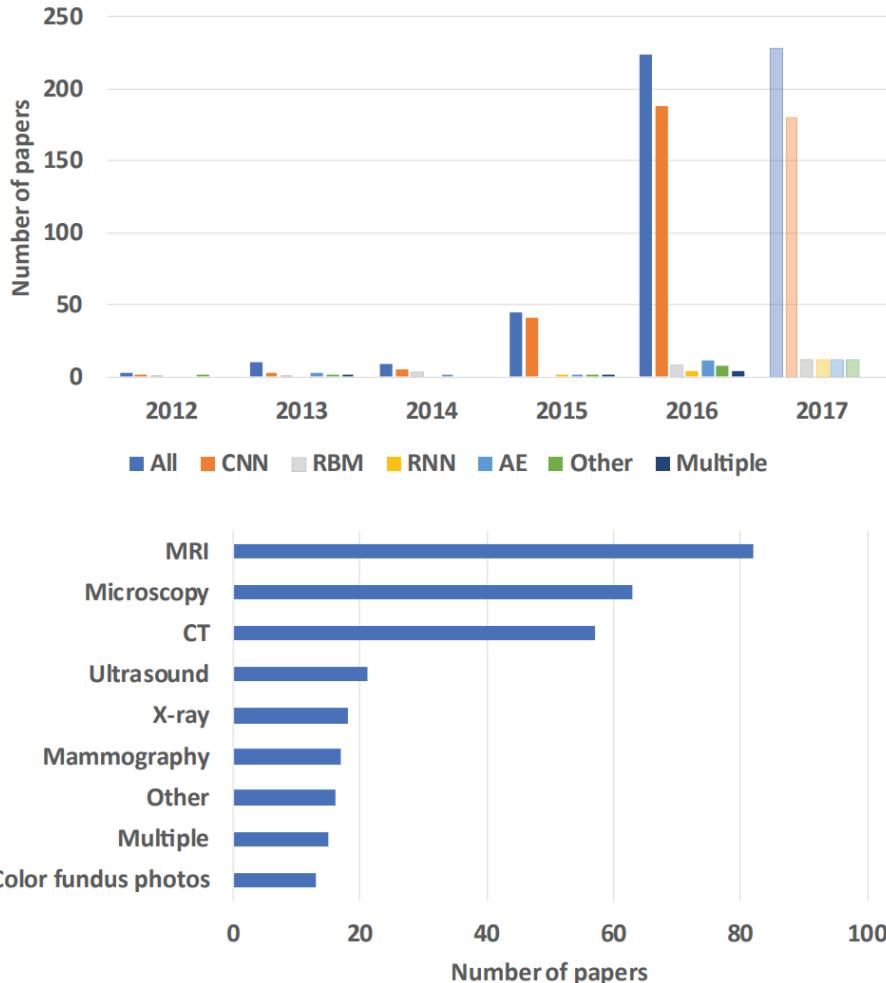


Summary

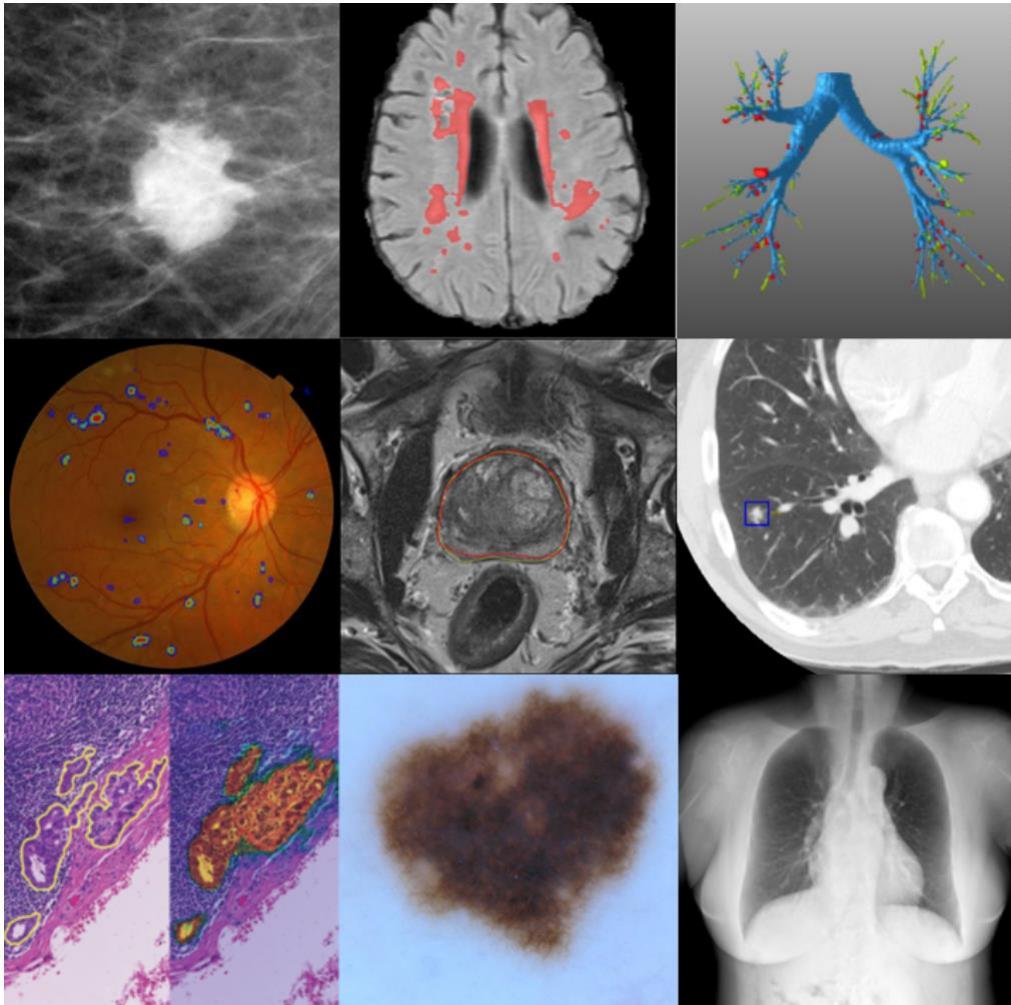
Jan 14	Classification (Medical Image Diagnosis)
Jan 16	Detection (Landmark and Computer Aided Detection)
Jan 21	Segmentation (Medical Image Segmentation)
Jan 23	GAN (Medical Image Synthesis)



Applications



Applications



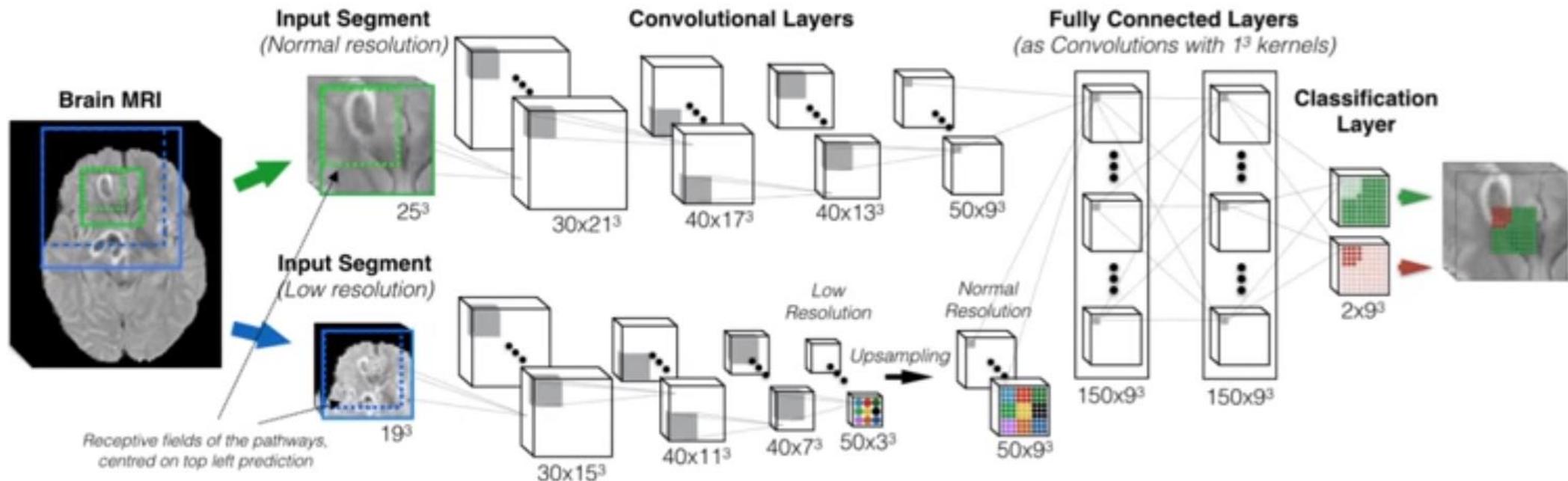
Some medical imaging applications in which deep learning has achieved state-of-the-art results. From top-left to bottom-right:

- mammographic mass classification (Kooi et al., 2016)
- segmentation of lesions in the brain (Ghafoorian et al. (2016b))
- leak detection in airway tree segmentation (Charbonnier et al., 2017)
- diabetic retinopathy classification (Kaggle challenge 2015)
- image from van Grinsven et al. (2016)
- prostate segmentation (top rank in PROMISE12 challenge)
- nodule classification (top ranking in LUNA16 challenge)
- breast cancer metastases detection in lymph nodes (CAMELYON16)
- human expert performance in skin lesion classification (Esteva et al., 2017)
- state-of-the-art bone suppression in x-rays, image from Yang et al. (2016c).

<https://arxiv.org/abs/1702.05747>

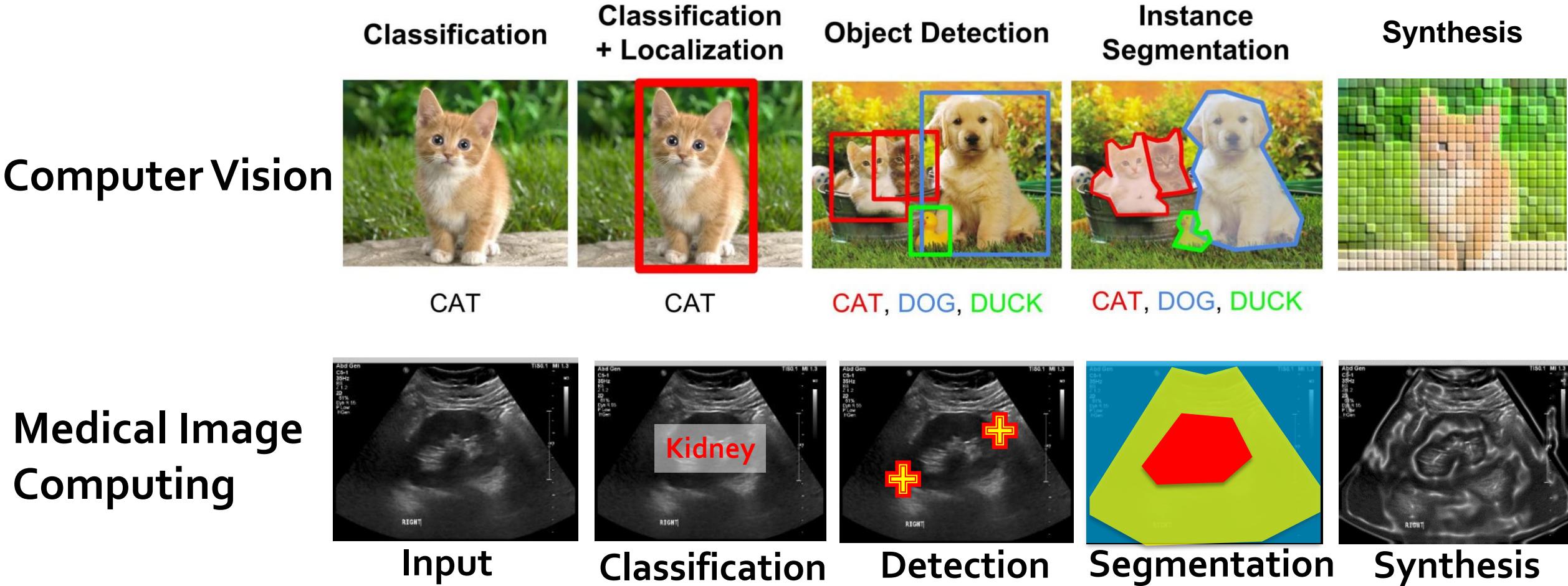
Success 1: Brain Cancer

brain lesions



Ben Glocker "Deep Learning in Medical Imaging" https://www.youtube.com/watch?v=2_Jv11VpOF4

Compare with Computer Vision



Simply Call APIs to Perform Deep Learning

Packages

- Caffe and Caffe2
 - NVIDIA Digits
- Theano
 - Lasagne
 - Keras
 - Blocks
- Torch
- TensorFlow
- MxNet
- MatConvNet
- Nervana Neon
- Leaf



Bar is Lower and Lower

GOOGLE CLOUD PLATFORM

How a Japanese cucumber farmer is using deep learning and TensorFlow

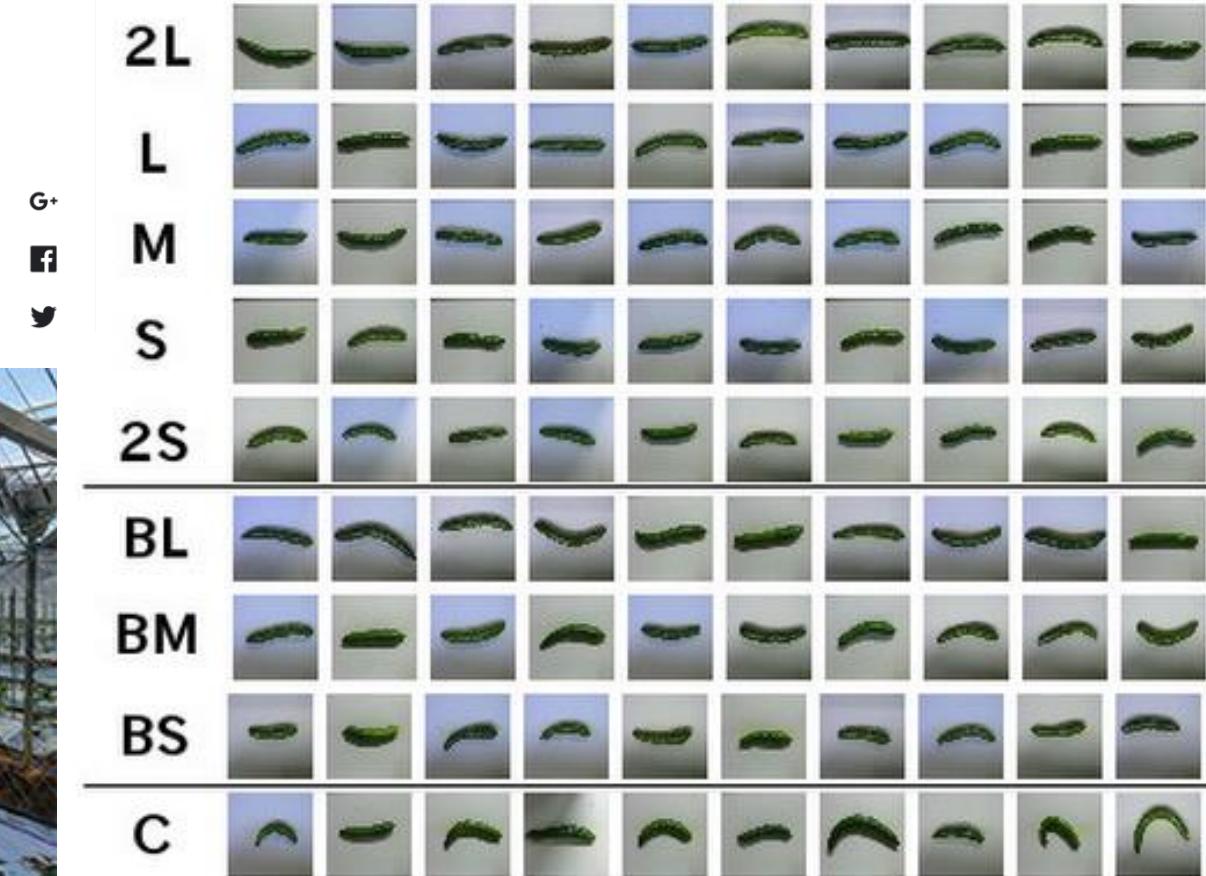
Kaz Sato
Developer Advocate, Google Cloud Platform

August 31, 2016

It's not hyperbole to say that use cases for machine learning and deep learning are only limited by our imaginations. About one year ago, a former embedded systems designer from the Japanese automobile industry named Makoto Koike started helping out at his parents' cucumber farm, and was amazed by the amount of work it takes to sort cucumbers by size, shape, color and other attributes.



<https://cloud.google.com/blog/products/gcp/how-a-japanese-cucumber-farmer-is-using-deep-learning-and-tensorflow>





**SO WHAT
SHOULD WE DO ?**

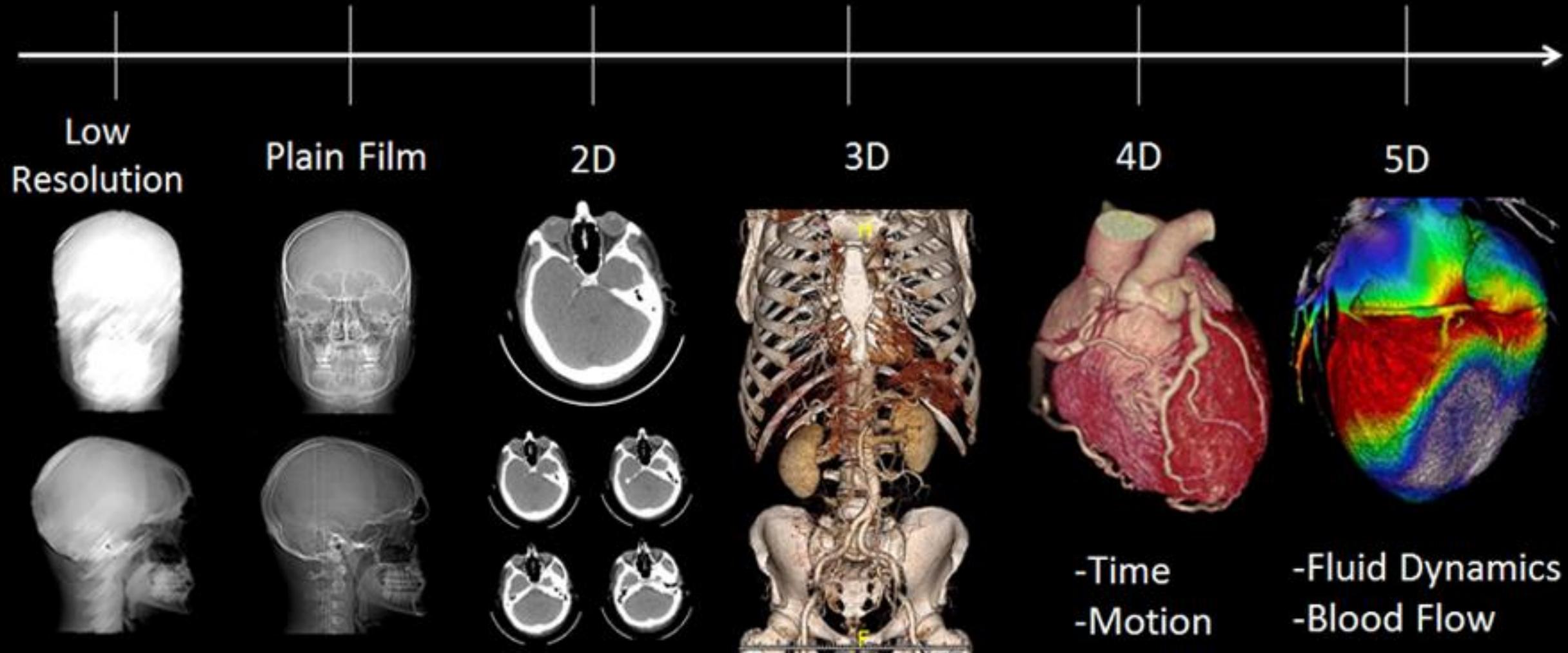


<https://www.nmr.mgh.harvard.edu/lab/laboratory-computational-imaging-biomarkers/miccai-2014-machine-learning-challenge>

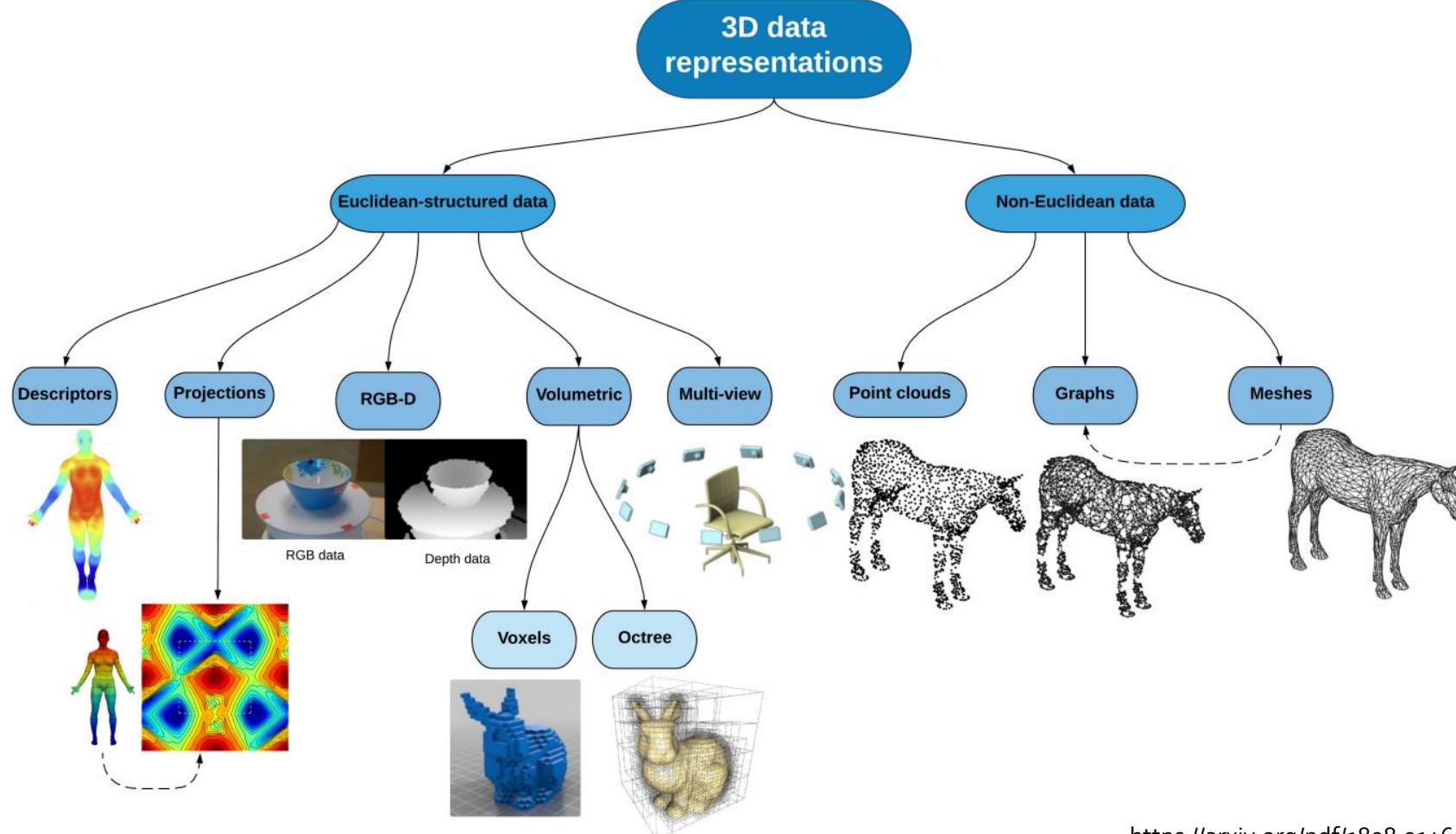
Solve the challenges in medical imaging!

- 1. How to use longitudinal images**
- 2. How to use multi-modal images**
- 3. Multi-task learning**
- 4. Medical images are mostly 3D, even 4D, 5D**
- 5. Learn from big unlabeled images**
- 6. Utilize previous efforts (registration, multi-atlas ...)**
- 7. Incorporate strong prior (shape, contour ...)**
- 8. Annotations are rare and expensive**
- 9. Co-learn from clinical data and other biomarkers**
- 10. How to add new annotations (online learning)**

Medical images are mostly 3D, even 4D, 5D



3D Data

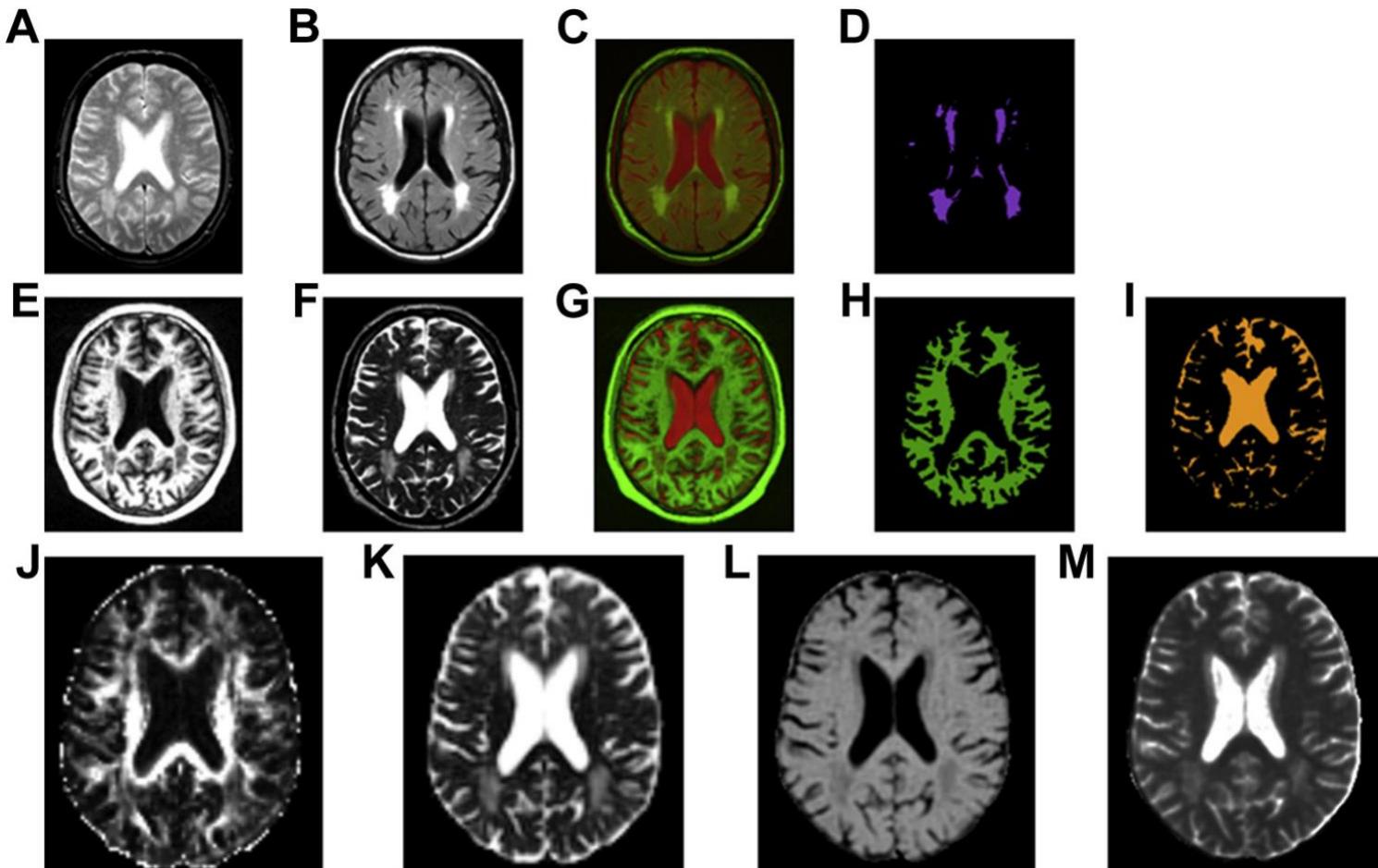


Feb 06

Beyond 2D:
3D and 4D
(RNN)
Networks

<https://arxiv.org/pdf/1808.01462>

How to use multi-modal images

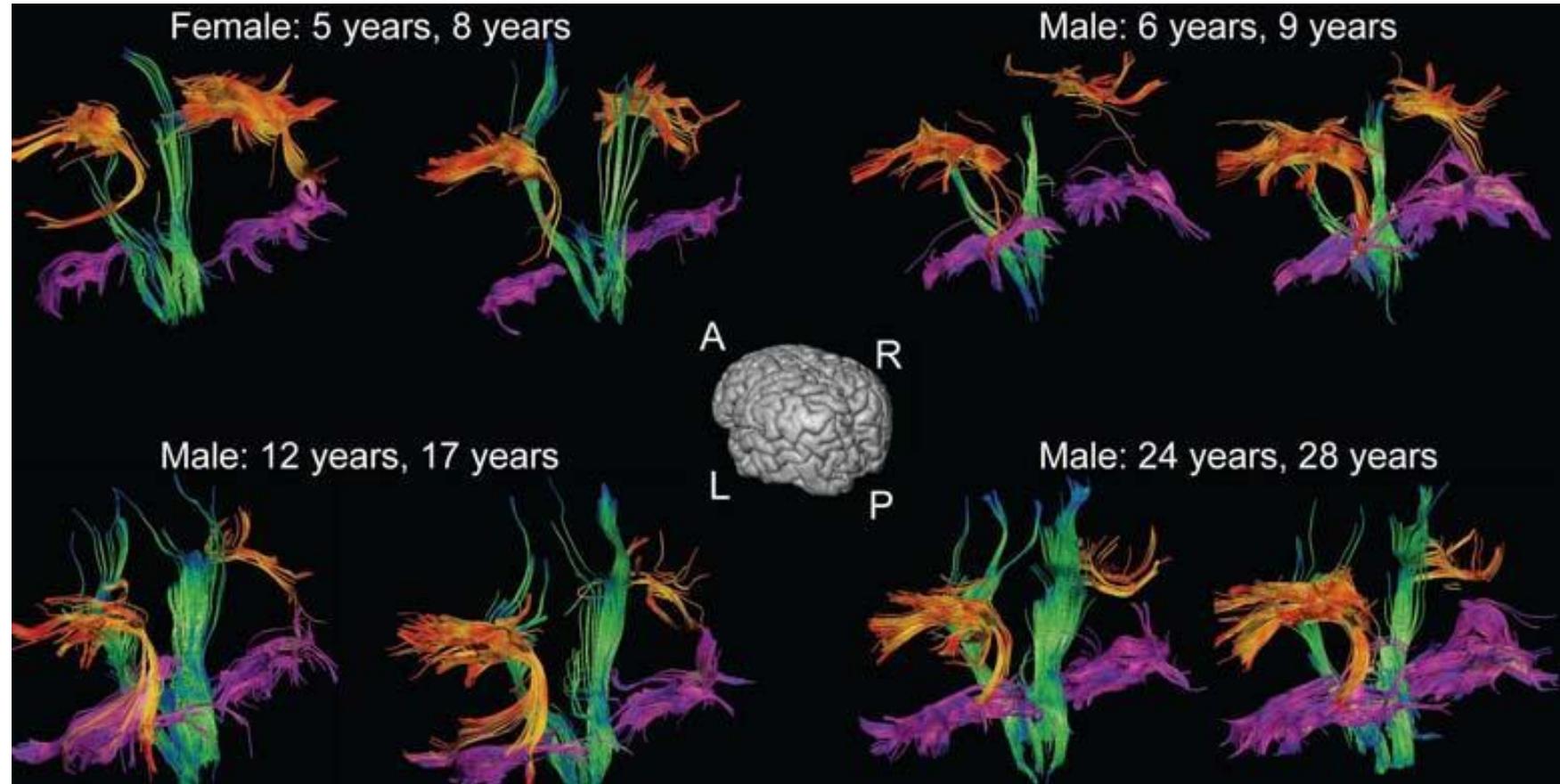


T₂*W (A) and FLAIR (B) structural scans are combined in red-green color space (C) to facilitate the extraction of WMH voxels (D). T₁W (E) and T₂W (F) structural scans are combined in red-green color space (G) to facilitate the extraction of NAWM (H) and CSF (I) voxels; the latter is subtracted from the WMH and NAWM masks to avoid CSF partial volume averaging within the measurement masks. The last row shows reconstructed parametric images of MRI biomarkers: FA (J), MD (K), MTR (L) and T₁ relaxation time (M)

Jan 30

Multi-model
Learning

How to use longitudinal images

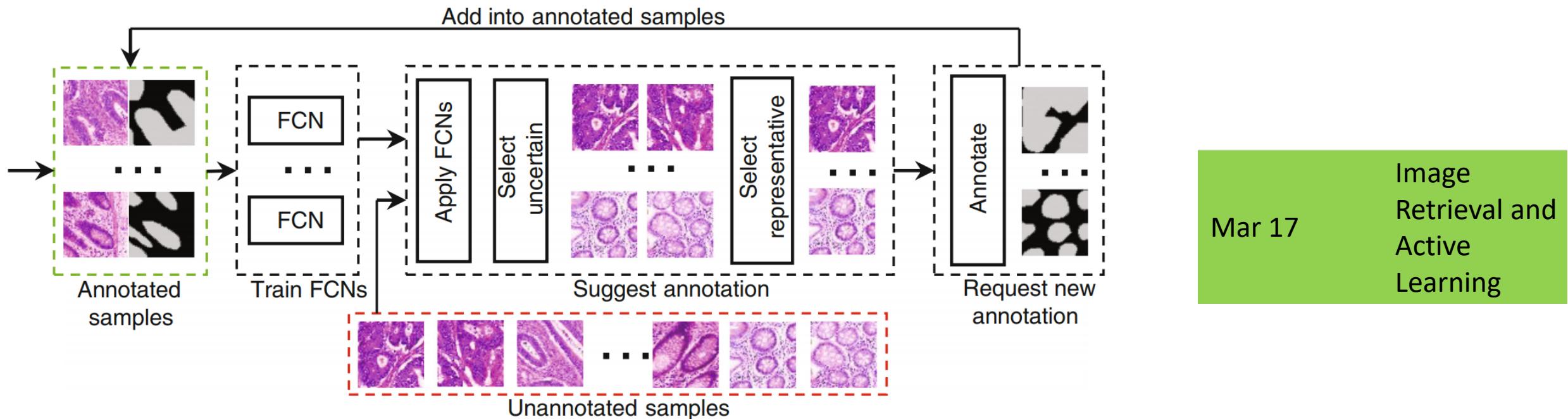


Mar 12

Spatial
Temporal
Learning

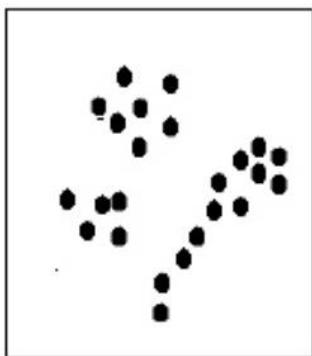
<https://brainchemist.wordpress.com/2011/09/23/development-of-human-brain-wiring-continues-into-adulthood-a-longitudinal-dti-study/>

Annotations are rare and expensive



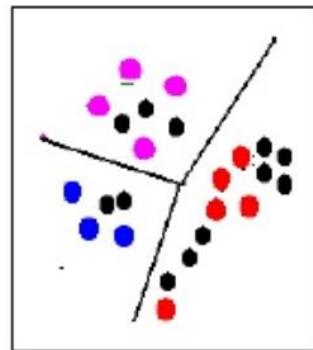
Learn from big unlabeled images

Unsupervised Learning



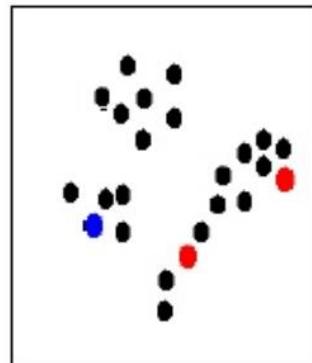
We know nothing about data structure

Supervised Learning



We know well data structure

Semi-Supervised Learning



We know something about data structure

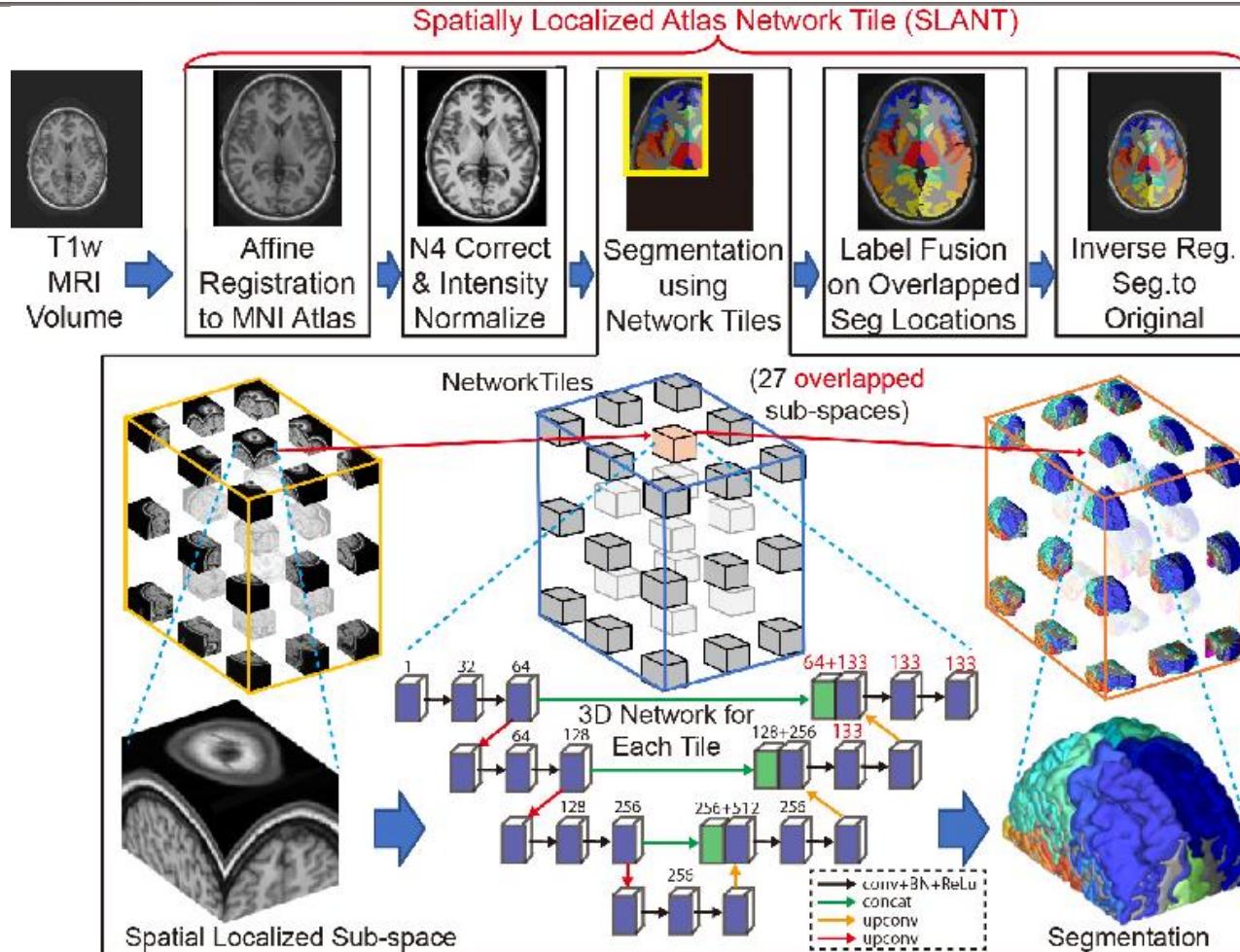
Feb 11

Semi-supervised Learning

Feb 13

Unsupervised Learning

Utilize previous efforts (registration, multi-atlas ...)



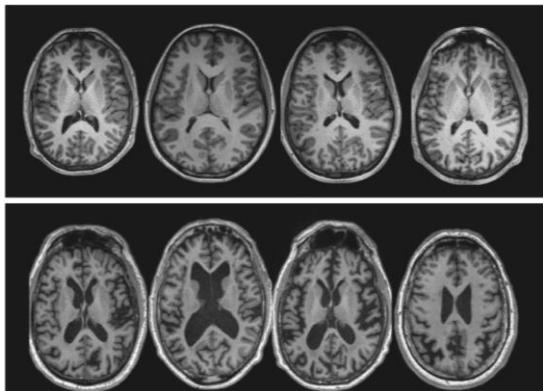
Feb 20

Data Augmentation and Preprocessing

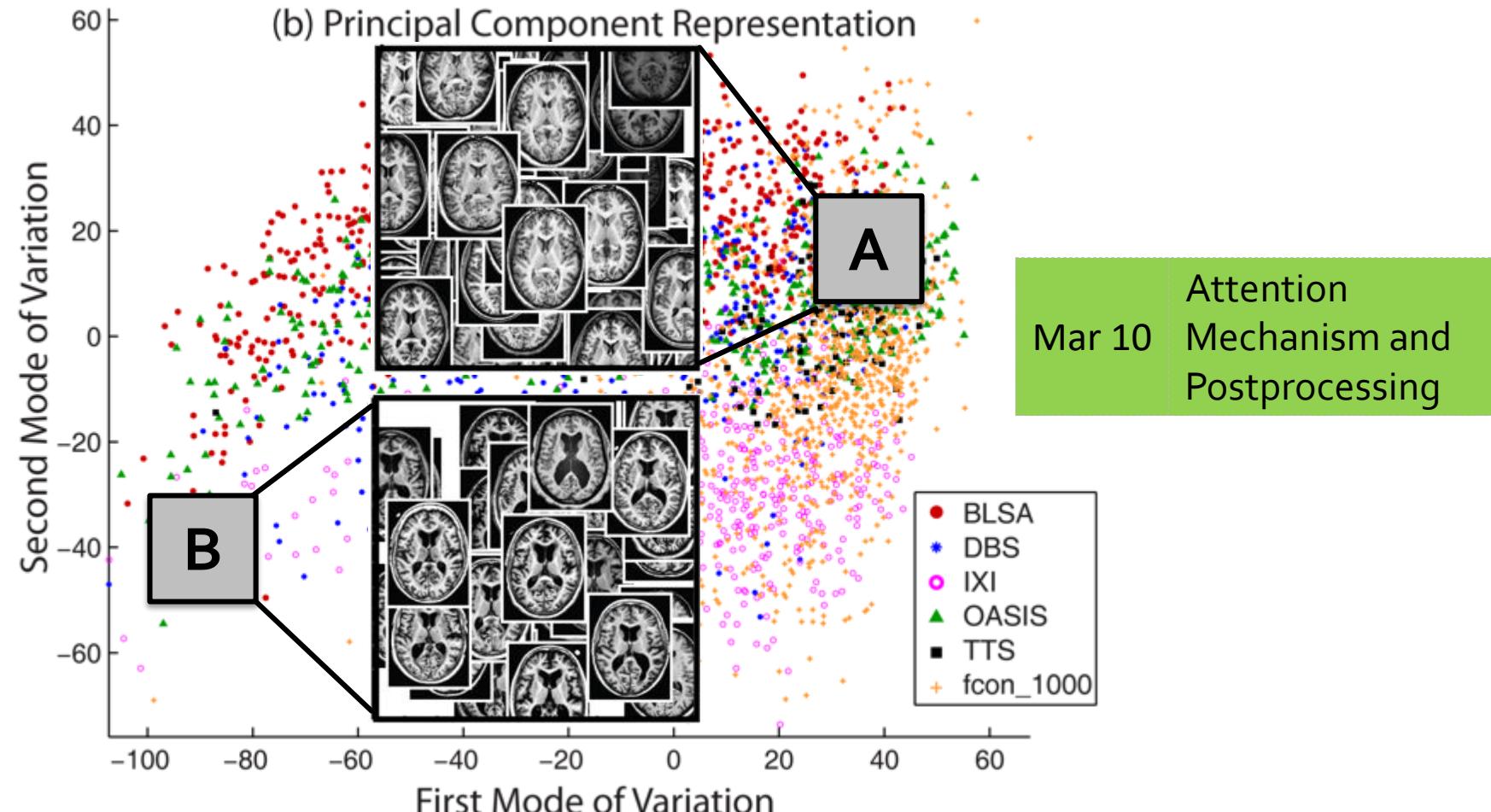
Incorporate strong prior and attention



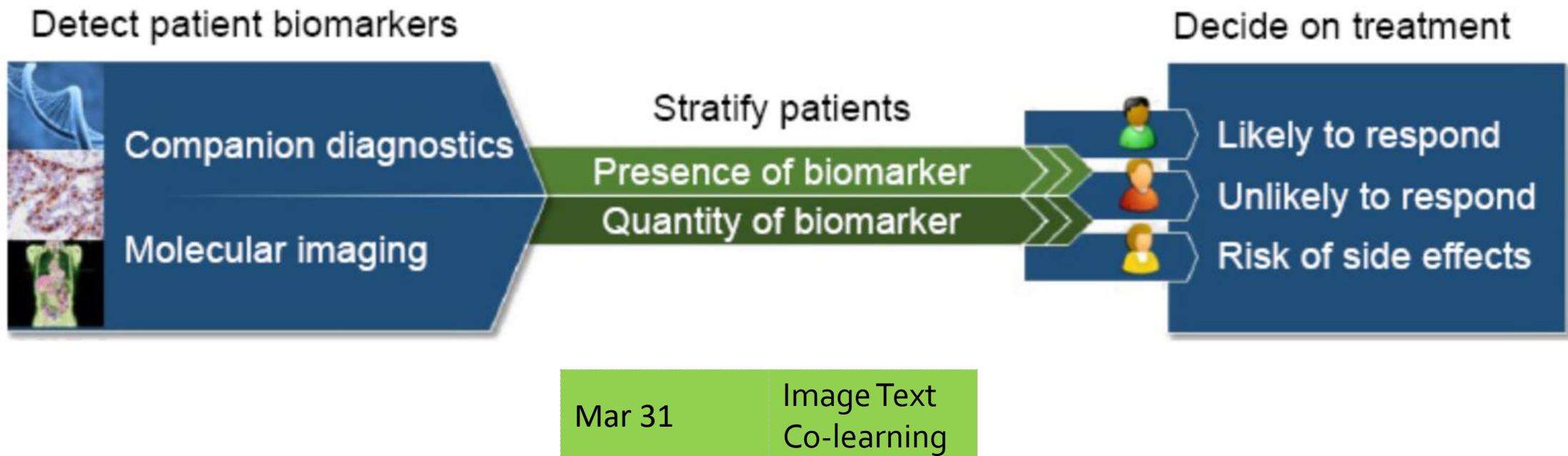
http://www.diegoantognini.com/projects/image_classification/



<http://www.mahoningvalleylanes.com/2425/car-picture-for-kids.html>

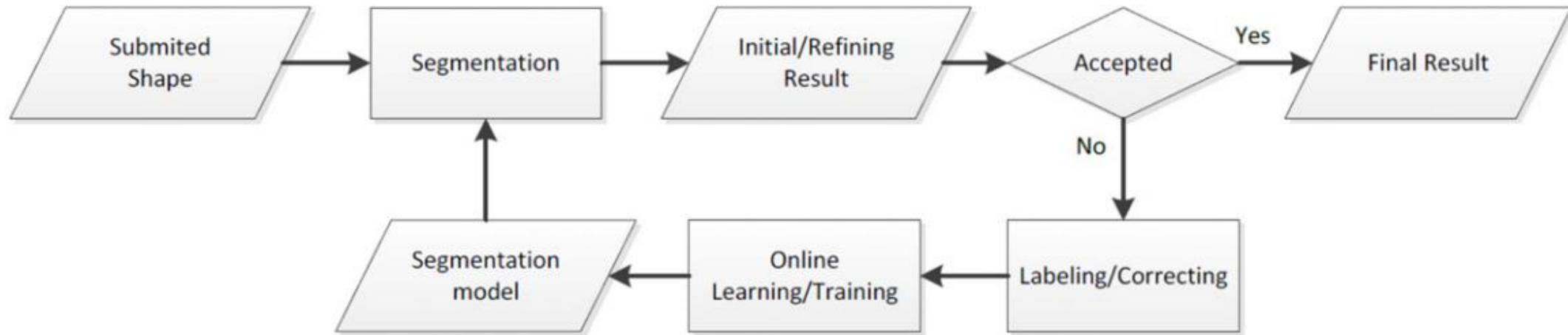


Co-learn from clinical data and other biomarkers



Van Heertum, Ronald L., Robert Scarimbolo, Robert Ford, Eli Berdugo, and Michael O'Neal. "Companion diagnostics and molecular imaging-enhanced approaches for oncology clinical trials." *Drug design, development and therapy* 9 (2015): 5215.

How to add new annotations (online learning)



Nov 15

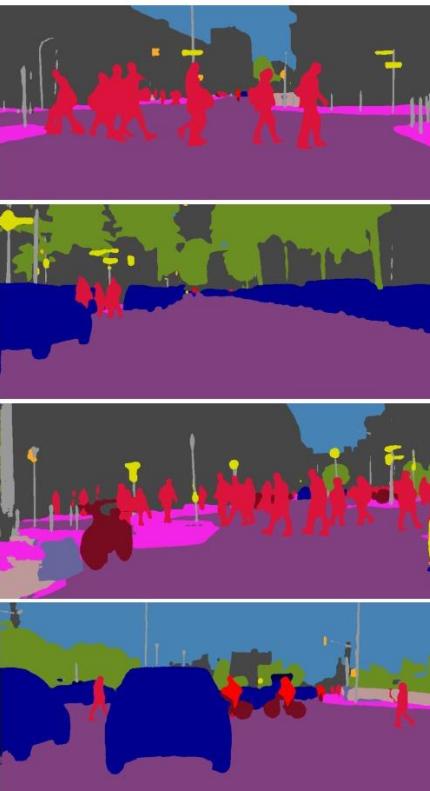
Online
Learning

Zhang, Feiqian, Zhengxing Sun, Mofei Song, and Xufeng Lang. "Progressive 3D shape segmentation using online learning." *Computer-Aided Design* 58 (2015): 2-12.

Multi-task Learning



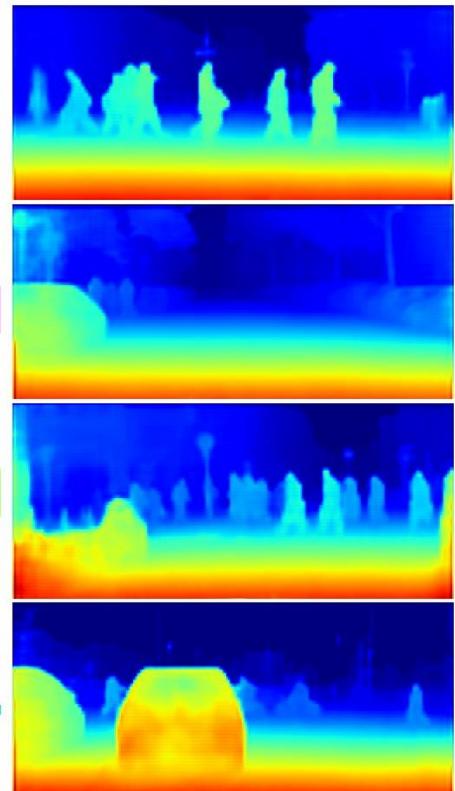
(a) Input image



(b) Segmentation output



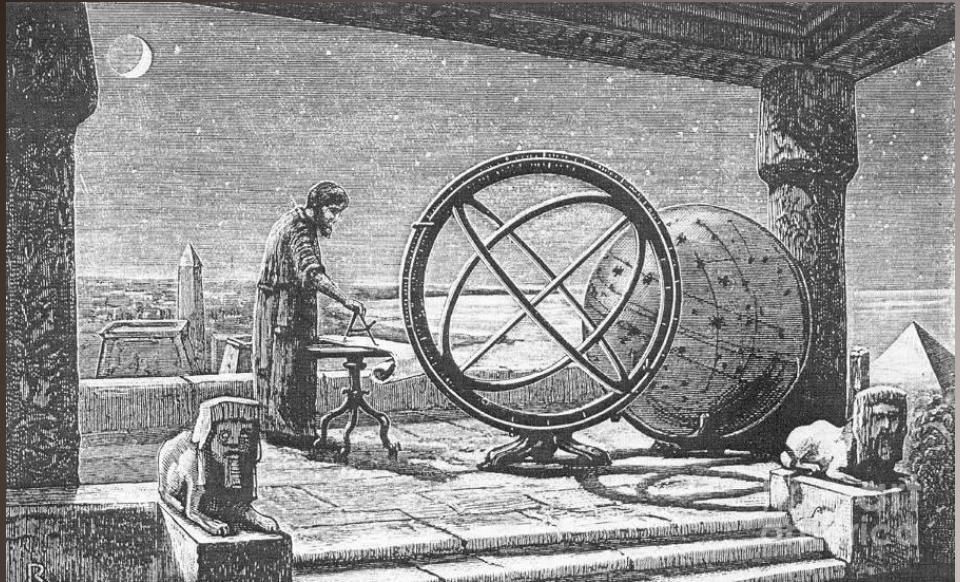
(c) Instance output



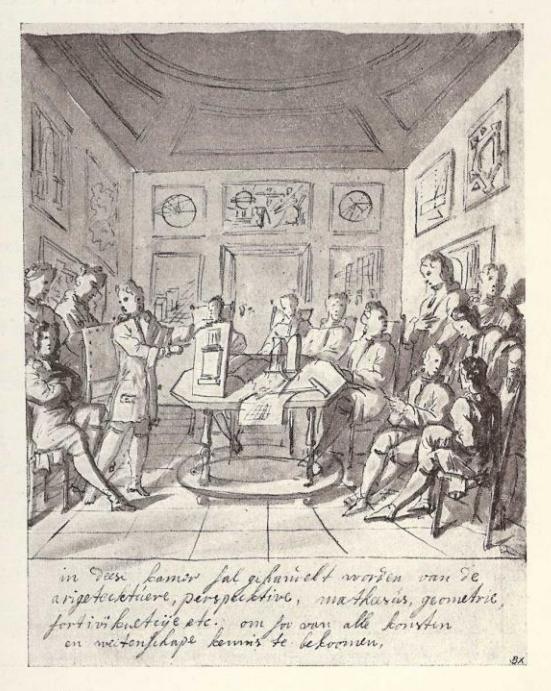
(d) Depth output

Feb 04

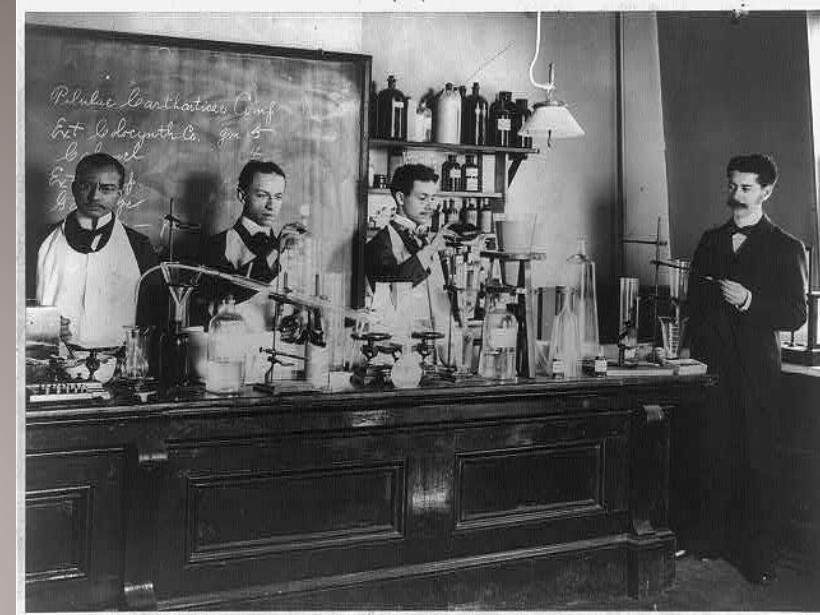
Multi-task
Learning



http://www.sohu.com/a/142293687_672687



<http://criticalarchitecture.org/en/research-projects/>



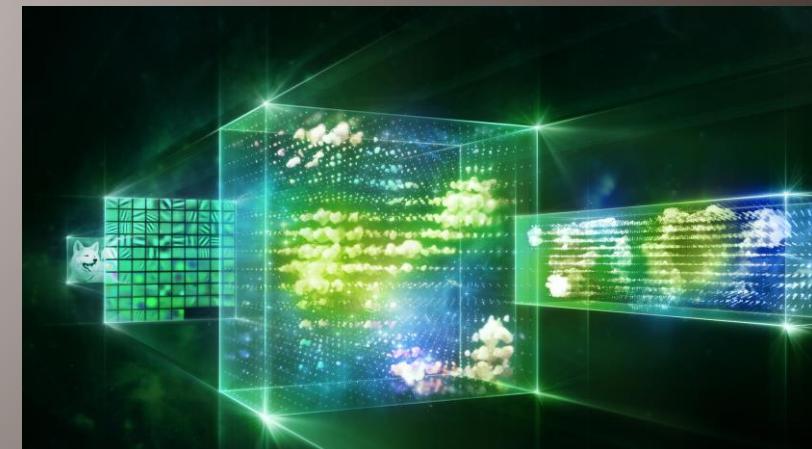
<http://libguides.law.umn.edu/druglaw>



<https://www.flickr.com/photos/l1nl/3094229534>



<https://www.scmp.com/lifestyle/travel-leisure/article/1983144/why-growing-army-digital-nomads-are-choosing-asia-their>



Contact

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TA Office Hours: Tuesday & Thursday, 1:00 pm – 4:00 pm, TA Office (FGH 385)

Instructor Office Hours: Tuesday & Thursday, 3:00 pm – 4:00 pm, FGH 376

Course Website: <https://my.vanderbilt.edu/cs8395>

Submission & Discussion: <https://www.vanderbilt.edu/brightspace>