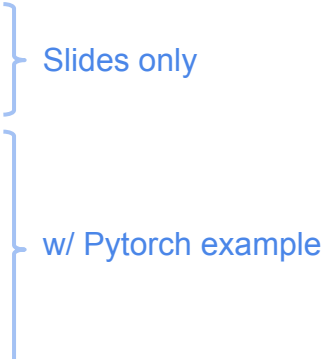


Learning Deep Learning with PyTorch

(1) Introduction

Qiyang Hu
UCLA IDRE
April 16, 2020

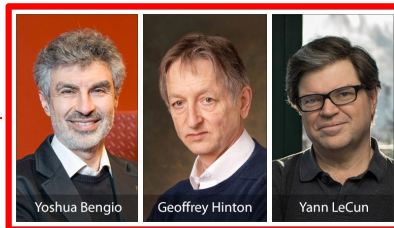
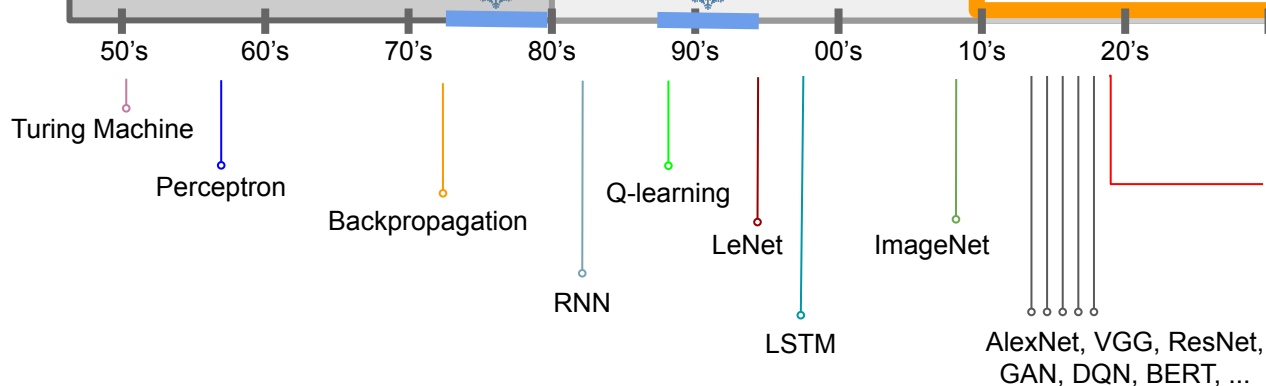
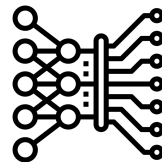
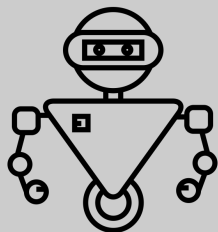
About the series

- Not a comprehensive course
 - **Broad:** high-level descriptive review
 - **Practical:** jupyter notebook examples
 - Workshop plan in this quarter
 - Introduction (Apr 16, 2020)
 - Learning Mechanics of Deep Learning (Apr 20, 2020)
 - Knowing PyTorch (Apr 23, 2020)
 - Convolutional Neural Networks (Apr 29, 2020)
 - Practical techniques in Deep Learning (May 4, 2020)
 - Recurrent Neural Networks and LSTM (May 7, 2020)
- 
- Slides only
- w/ Pytorch example

Artificial Intelligence

Machine Learning

Deep Learning



[2018 ACM Turing Award Recipients](#)

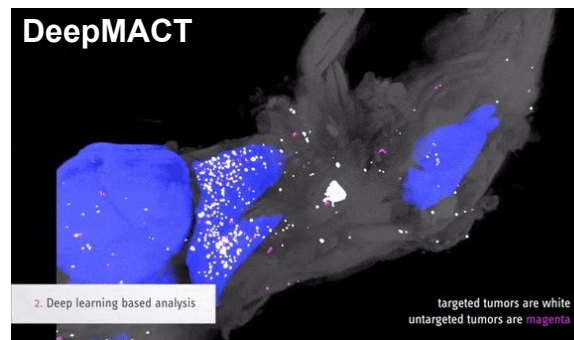
Amazing Deep Learning Achievements in 2019



Detectron2



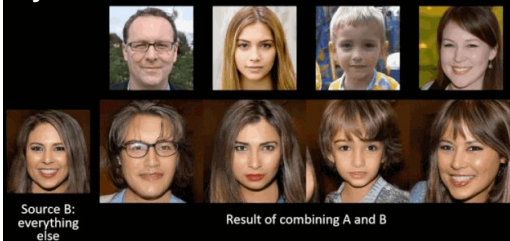
DeepMACT



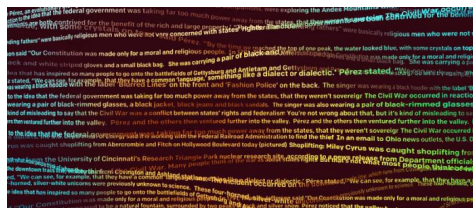
Equation-solving AI

StyleGAN2

Source A: gender, age, hair length, glasses, pose



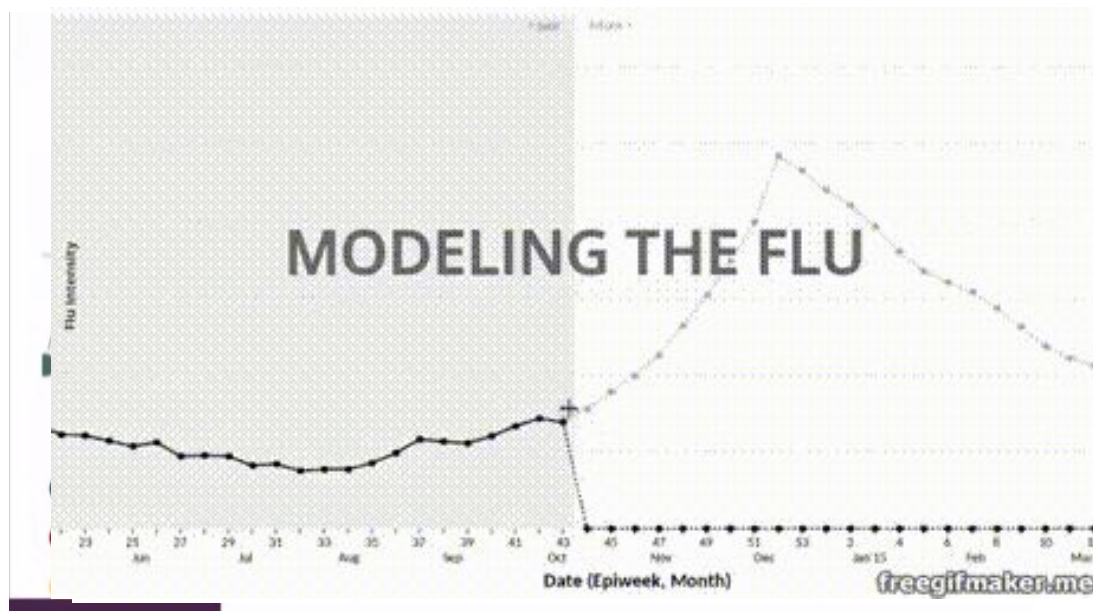
GPT-2



EQUATION	SOLUTION
$y' = \frac{16x^3 - 42x^2 + 2x}{(-16x^6 + 112x^7 - 204x^8 + 28x^9 - x^4 + 1)^{1/2}}$	$y = \sin^{-1}(4x^4 - 14x^3 + x^2)$
$3xy \cos(x) - \sqrt{9x^2 \sin(x)^2 + 1} y' + 3y \sin(x) = 0$	$y = c \exp(-\ln(3x \sin(x)))$
$4x^4 y'' - 8x^3 y'^2 - 8x^2 y y'' - 3x^2 y'' - 8x^2 y'^2 - 6x^2 y'' - 3x^2 y'' - 9x y' - 3y = 0$	$y = \frac{c_1 + 3x + 3 \log(x)}{x(c_2 + 4x)}$

Deep Learning Against COVID-19

Diagnosis, prognosis, drug discovery, and public health



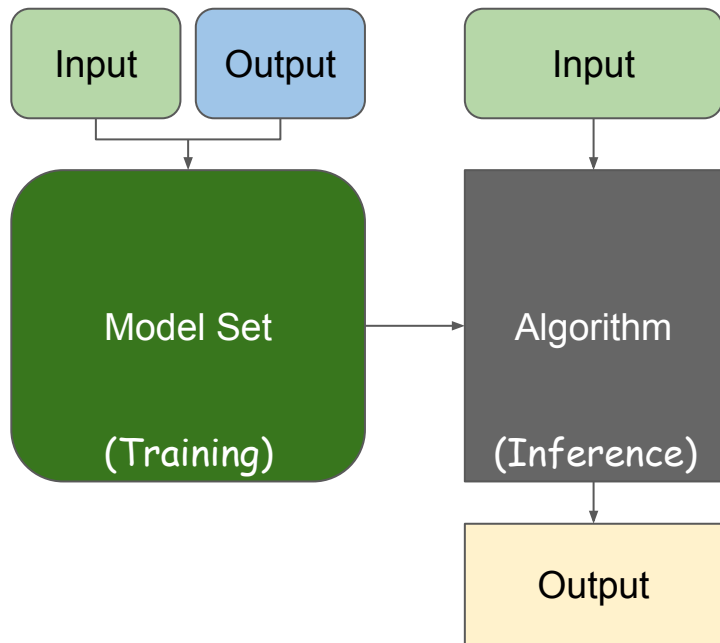
Research by Jennifer Lu and
<https://www.stanford.edu/>

What is Machine Learning?

Traditional Programming



Machine Learning

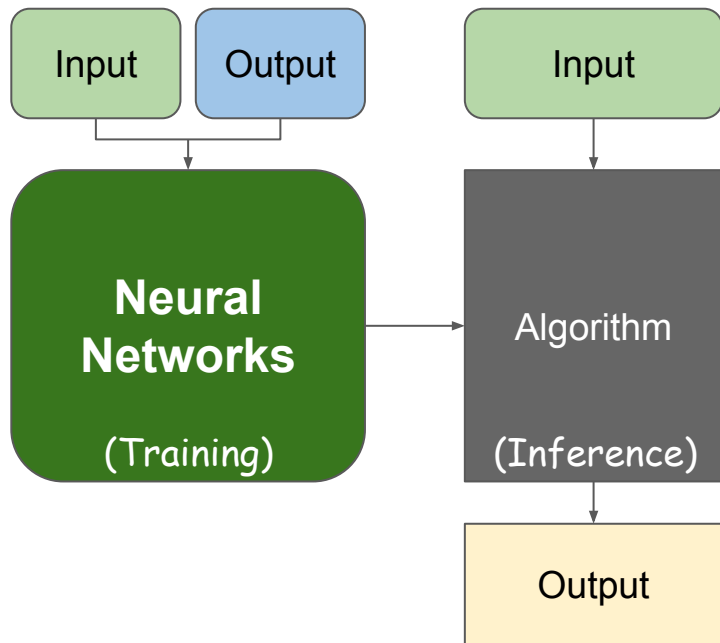


What is Deep Learning?

Traditional Programming

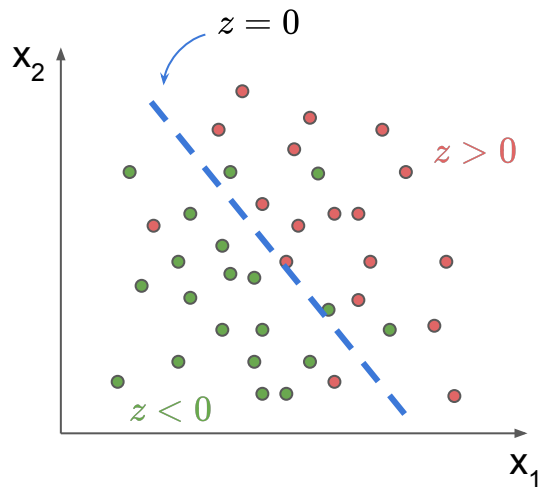


Deep Learning

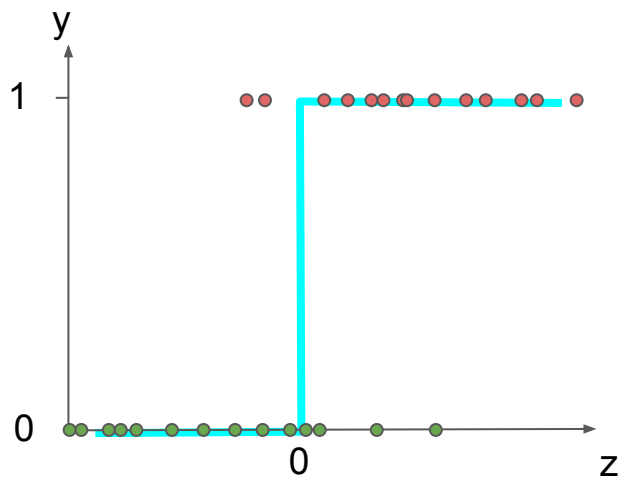


What is Neural Network?

- Recap for simple linear classification problem

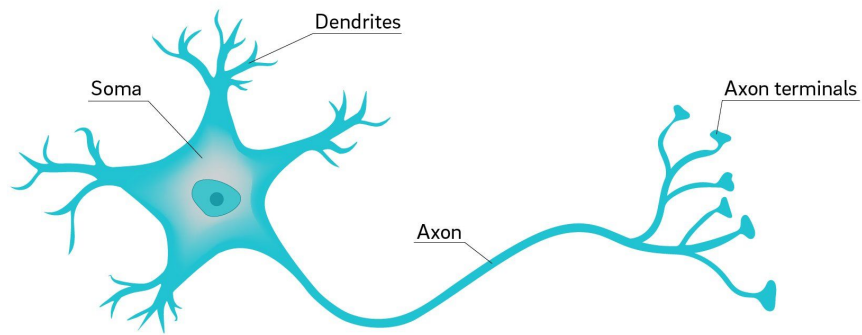
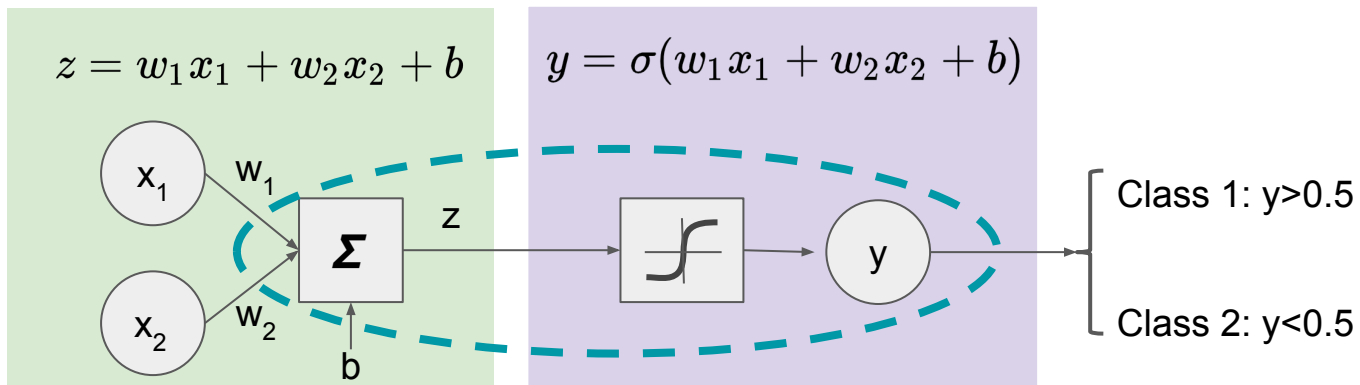


$$z = w_1 x_1 + w_2 x_2 + b$$

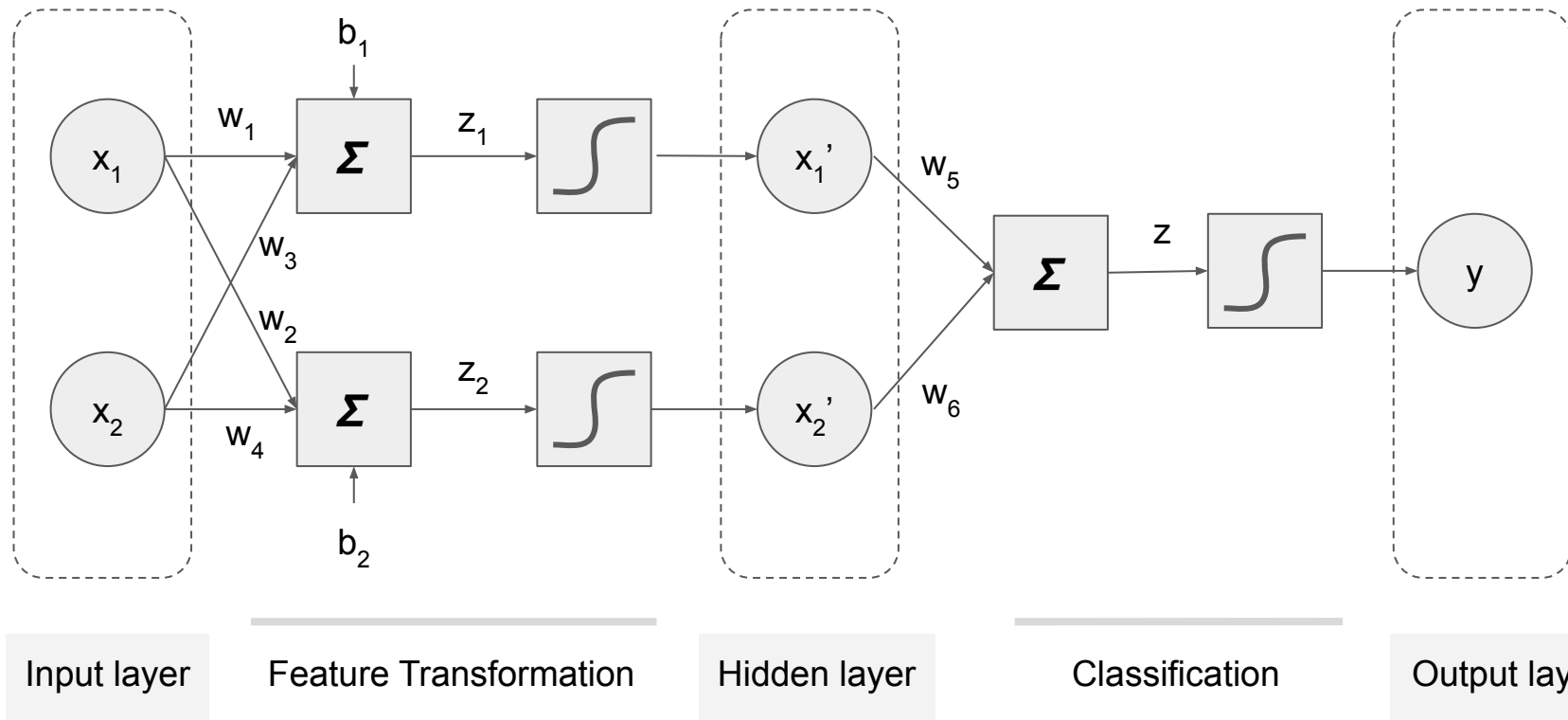


$$y = \sigma(z) = \begin{cases} 1 & \text{if } z > 0 \\ 0 & \text{if } z < 0 \end{cases}$$

A linear classifier ~ one artificial neuron



(Deep) Neural Networks ~ piling/stacking logistic-regression classifiers



How deep a deep learning network can be?

- LeNet-5 (1998)



Year	CNN	Developed by	Place	Top-5 error rate	No. of parameters
1998	LeNet(8)	Yann LeCun et al			60 thousand
2012	AlexNet(7)	Alex Krizhevsky, Geoffrey Hinton, Ilya Sutskever	1st	15.3%	60 million
2013	ZFNet()	Matthew Zeiler and Rob Fergus	1st	14.8%	
2014	GoogLeNet(19)	Google	1st	6.67%	4 million
2014	VGG Net(16)	Simonyan, Zisserman	2nd	7.3%	138 million
2015	ResNet(152)	Kaiming He	1st	3.6%	

Key Terminology in Deep Learning

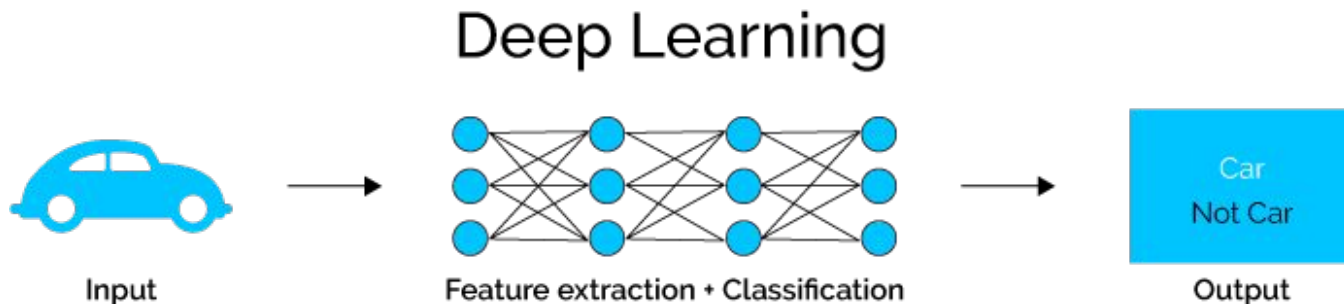
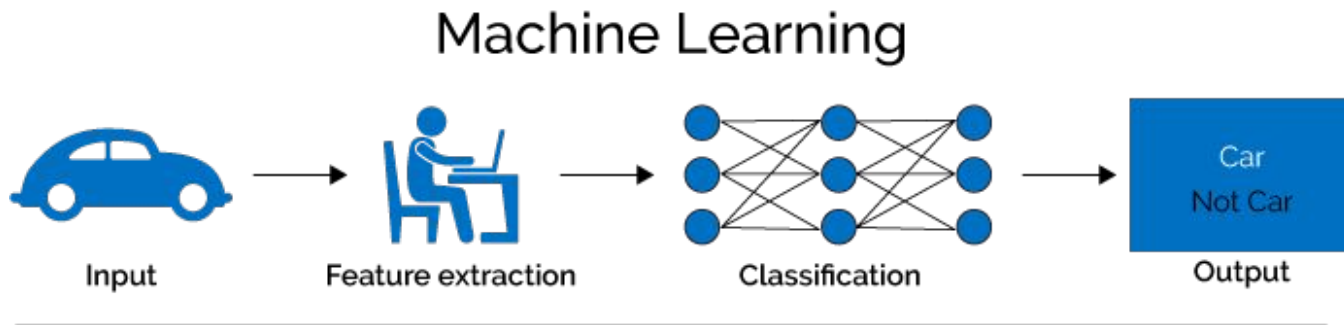
- Datasets:
 - Label: a desired output (e.g. house price)
 - Feature: a known input (e.g. address, condition, household income, etc)
- Model: relationship between input & output
 - Parameter: to be learned from data, e.g. weight, coefficients
 - Weight: a coefficient for a feature in linear model
 - Bias: an intercept or offset from an origin
 - Hyperparameter: often set by heuristics, e.g. learning rate, depth of trees, batch, epoch.
 - Batch: a subset from the division of training datasets
 - Epoch: all data in training sets has had an opportunity to update the internal model parameters

[Complete Glossary](#)

A lot of “Learning”s to learn

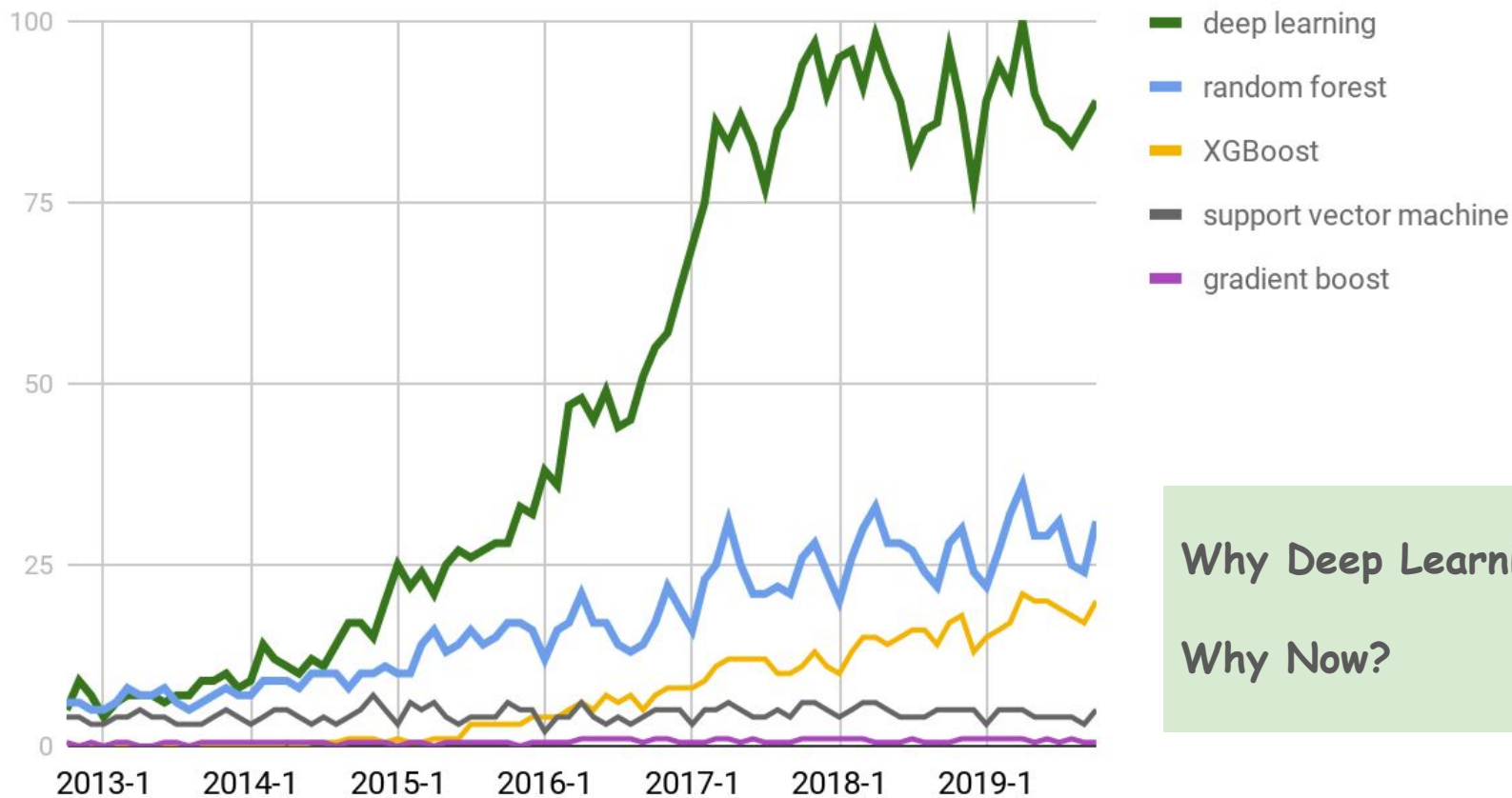
- Supervised Learning (data with labels)
 - Regression
 - Classification
- Unsupervised Learning (data without labels) (Auto Encoders)
- Semi-supervised Learning (data with partial labels)
- Reinforcement Learning (reward rules to get data) (PPO, Deep Q-learning)
- Self-supervised learning (no rules & no labels) 🔥
- Transfer Learning (data with unrelated labels)
 - (zero-shot learning, one-shot learning, few-shot learning, etc.)
 - ⇒ Continuous learning
 - ⇒ Meta Learning (MAML, LSTM)

Machine Learning vs. Deep Learning



Source: <https://www.xenonstack.com/blog/log-analytics-deep-machine-learning/>

Interest over time from Google Trends



Why Deep Learning?

Why Now?

Driving Forces in Deep Learning (1): *Algorithms*

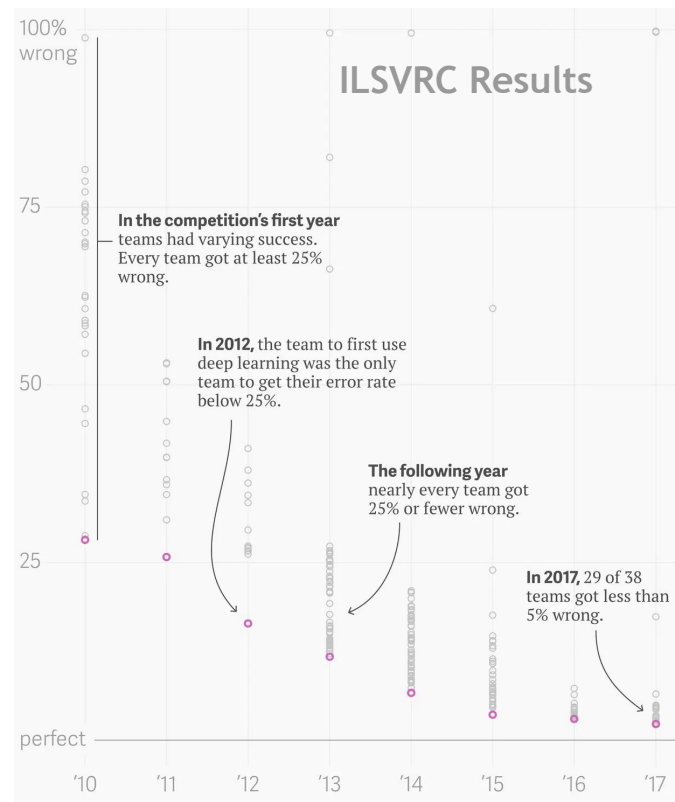
- Key Issues in Deep Learning:
 - Optimization for models with huge number of parameters
 - Gradient Propagation through stacks of layers
 - Gradient vanishing and exploding
- Algorithmic improvements in 2009-2010:
 - Better activation functions, weight-initialization schemes, optimization schemes
- Advanced techniques in 2014-2018:
 - Batch normalization/Drop-out, Residual/skip connections, Depth-wise separable convolutions
- Progress keeps accelerating! 2019-
 - Low precision neural networks
 - Graph neural networks

Driving Forces in Deep Learning (2): *Data*

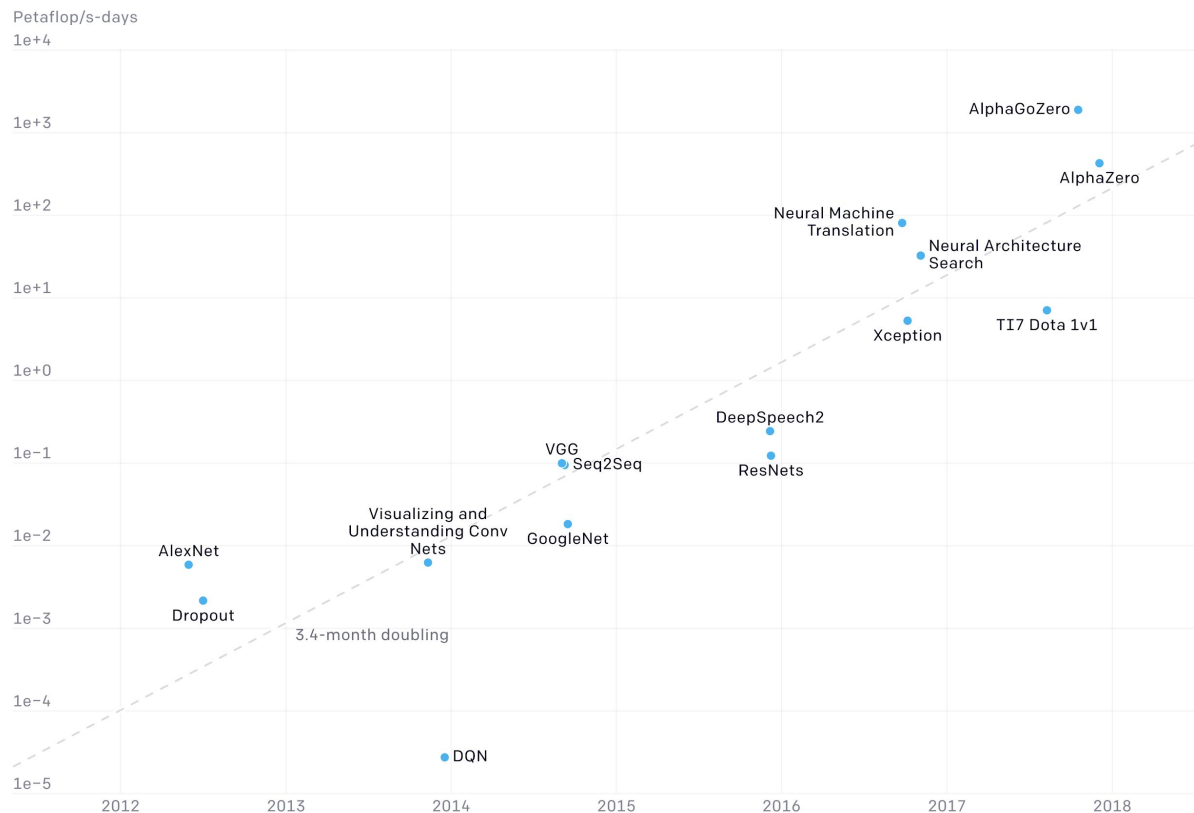
- **ImageNet** (a large dataset of annotated photographs built on 2009)
 - 14 million+ images
 - 21,000 groups or classes
 - ILSVRC competition (1.2 million image, 1000 classes)

- **Kaggle** (Founded in 2010, acquired by Google in 2017)
 - 1,000,000+ registered users in 194 countries in 2017
 - Hosts 19K+ of [datasets](#) and 200K+ code snippets
 - Famous for the high-rewards [competitions](#)

- **Datasets from the rise of internet**
 - User-generated image tags on Flickr
 - Video dataset/tags on Youtube
 - Data from Wikipedia for NLP

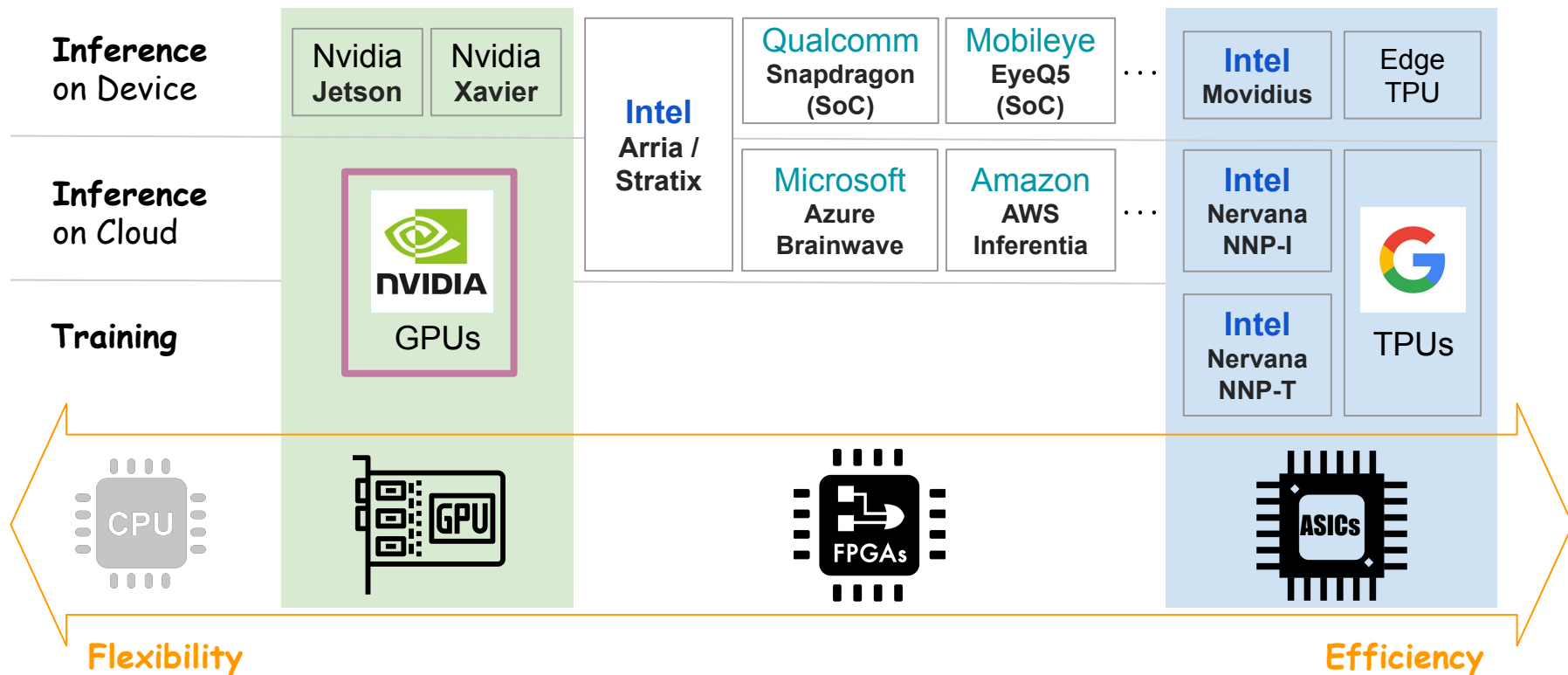


AI compute amount increases 10 times per year!



Source:
[OpenAI Report](#)

Driving Forces in Deep Learning (3): *Hardware*



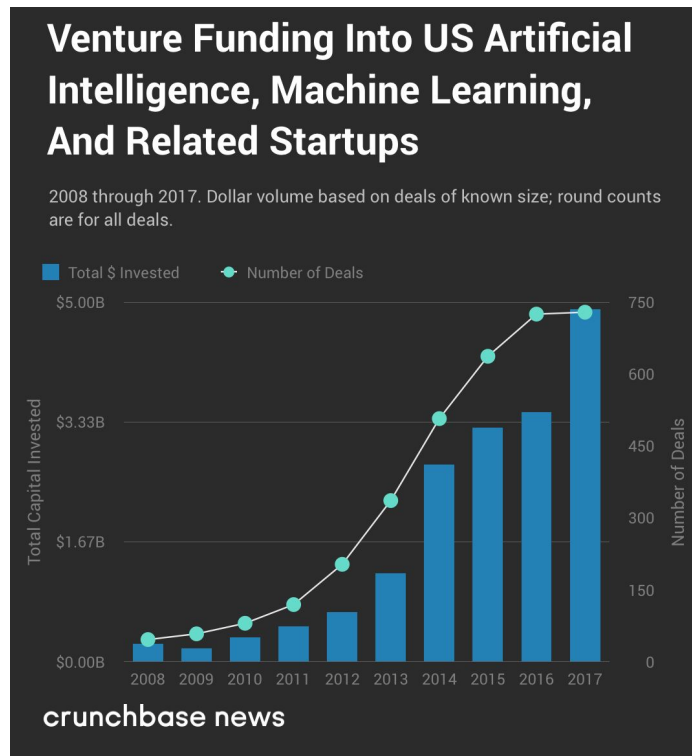
Free GPU Computation Resources

- Cloud-based resources (Google Colaboratory, Kaggle, Paperspace's Gradient)
 - A free Jupyter notebook env that requires no setup and runs entirely in the cloud.
 - Google Drive → New → More → Google Colaboratory
 - Kaggle.com → Log in → Kernel → New Kernel
- Hoffman2 ([GPU resources](#))
 - Work under python shell in terminal:
 - `qrsh -l gpu,P4`
`module load python/anaconda3`
`conda activate pytorch-1.3.1-gpu`
 - Work in Jupyter notebook session:
 - For the very first time, add the virtual env to kernel in the above qrsh session:
 - `python -m ipykernel install --user --name=pytorch-1.3.1-gpu`
 - Using [h2jupynb](#):
 - `./h2jupynb -v anaconda3 -g yes -c P4 -l 10.0`

	Colab	Kaggle	Hoffman2
CPU Type	Intel Xeon 2.30GHz	Intel Xeon 2.30GHz	Intel Xeon 2.80GHz
Slots/Threads available	1 core / 2 threads	1 core / 2 threads	8 cores / no hyper-threads
RAM available	12 GB	18 GB	24 GB
Disk available	311 GB	626 GB	1 TB
GPU Type	Tesla T4 (2018)	Tesla P100 (2018)	Tesla P4 (2016)
GPU SP Floating-Point Perf	8.1 TFLOPs	10.6 TFLOPs	5.5 TFLOPs
GPU Memory	16 GB	16 GB	8 GB
Active Time Limit	8 hours	6 hours	24 hours

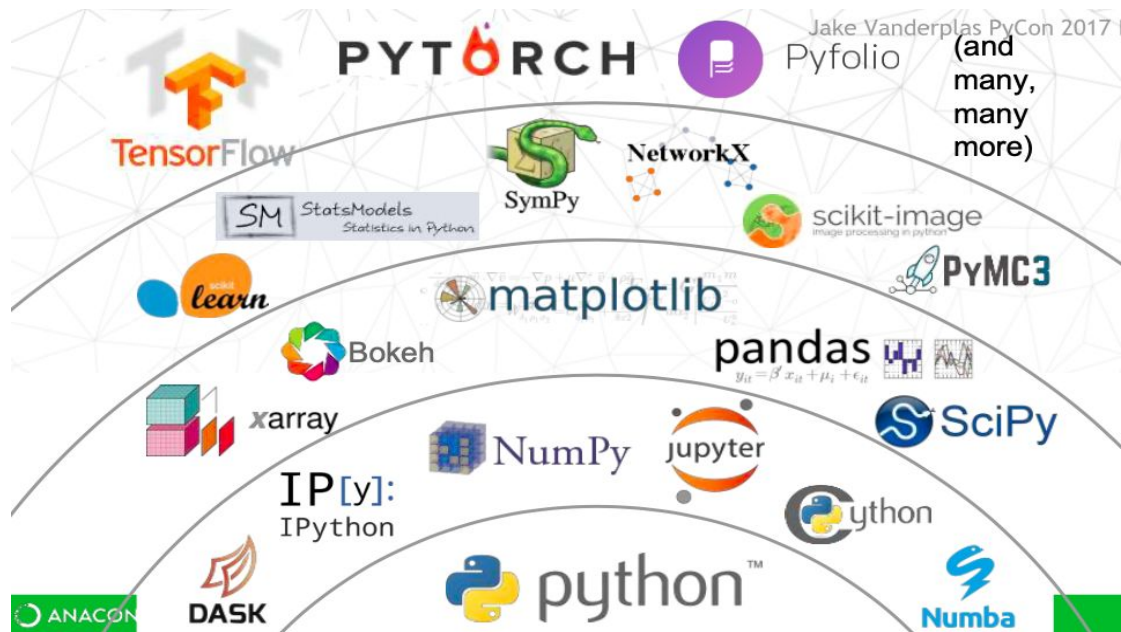
Driving Forces in Deep Learning (4): *Investment*

- Venture Capital Investment soars
 - [Source](#)
 - 20x increase in 8 years
 - Most for deep-learning
- AI acquisitions
 - Google: \$500M for DeepMind (2013)
 - Intel: \$400M for Nervana Systems (2016)
 - Tons of M&As undisclosed
- Demand drives supply
 - 100x more people working on deep learning



Driving Forces in Deep Learning (5): *Toolsets*

- From C++/Cuda to scripting languages (Python, R)

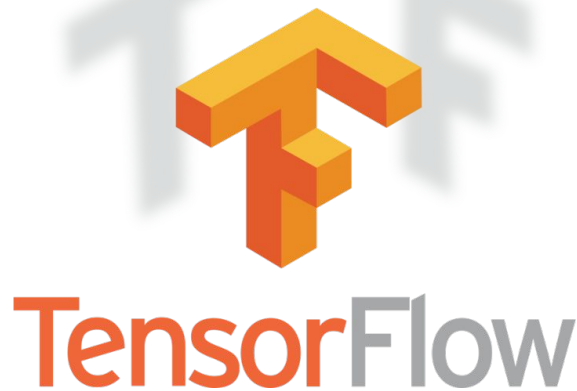


Deep Learning Framework Battles



Finals?

PYTORCH vs.



FYI

- Github Repo:
 - <https://github.com/huqy/idre-learning-deep-learning-pytorch>
- Slack workspace:
 - bit.ly/join-LDL
- Contact me
 - huqy@idre.ucla.edu
 - Direct message in Slack
- IF you don't have plan to attend the rest of workshops, :
 - Please fill out our series survey: bit.ly/2X2phyS