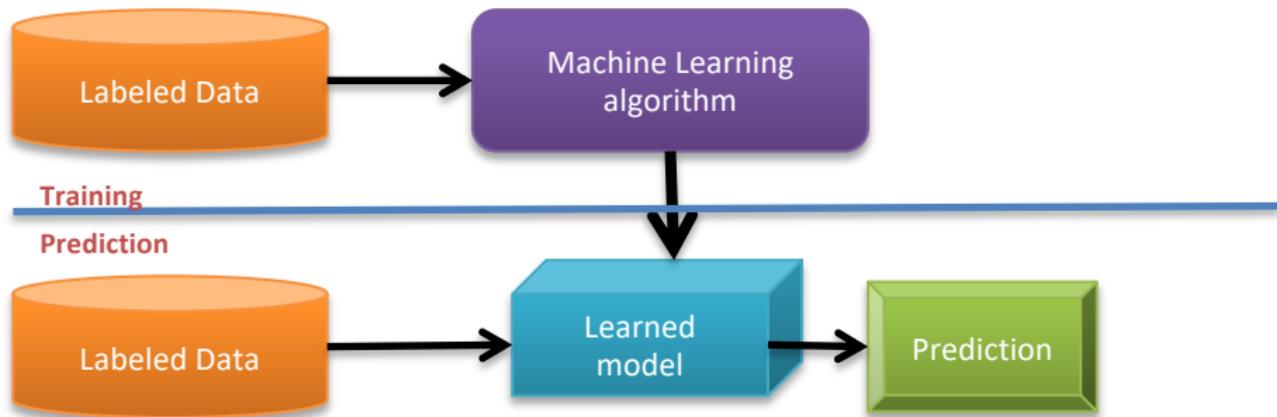


# Introduction to Deep Learning

# Machine Learning Basics

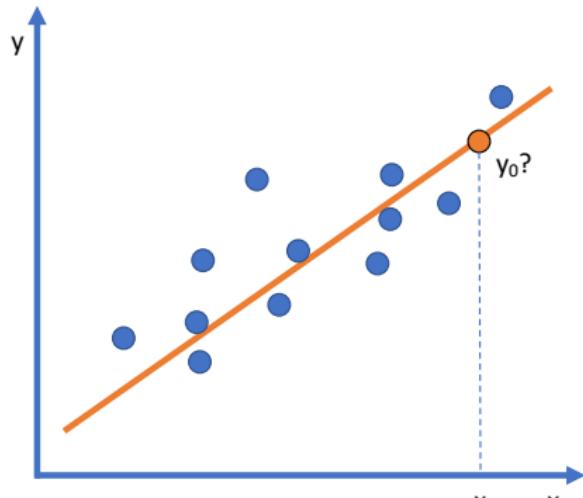
Machine learning is a field of computer science that gives computers the ability to **learn without being explicitly programmed**



Methods that can learn from and make predictions on data

# Machine Learning Basics

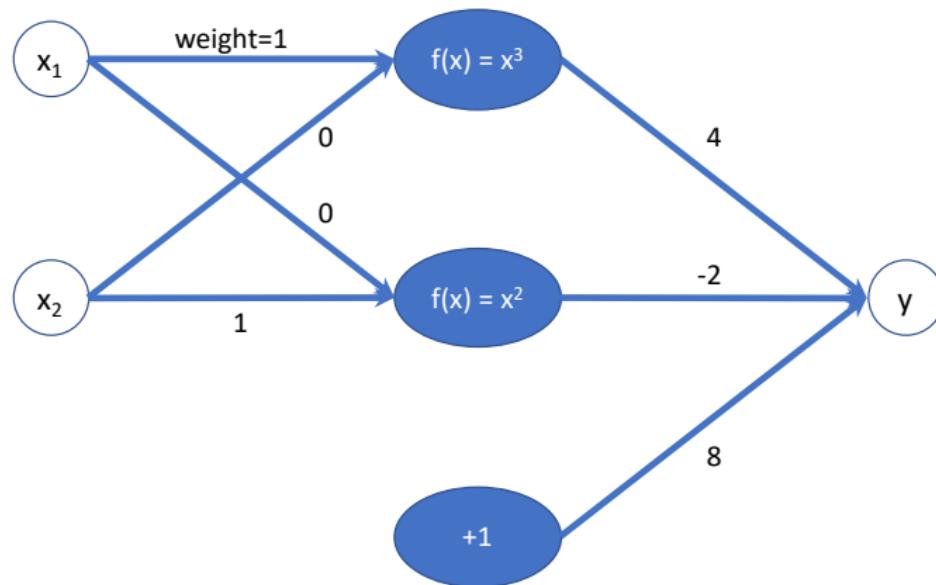
Input data → prediction model → output label



$$y = F(x)$$

# Machine Learning Basics

$$y = 4x_1^3 - 2x_2^2 + 8$$



# Types of Learning

**Supervised:** Learning with a **labeled training** set

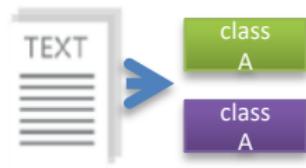
Example: email **classification** with already labeled emails

**Unsupervised:** Discover **patterns** in **unlabeled** data

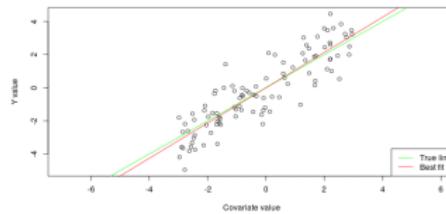
Example: **cluster** similar documents based on text

**Reinforcement learning:** learn to **act** based on **feedback/reward**

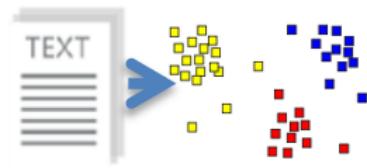
Example: learn to play Go, reward: **win or lose**



Classification



Regression



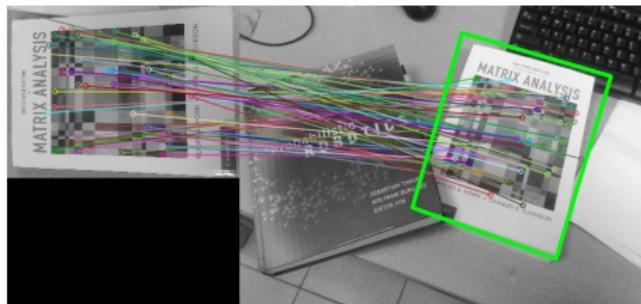
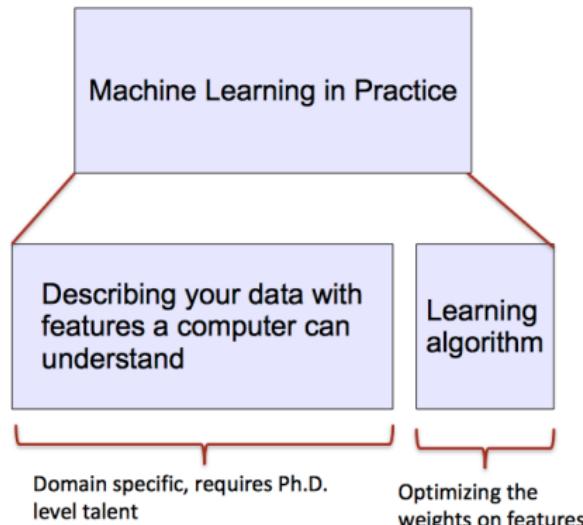
Clustering

Anomaly Detection  
Sequence labeling

# ML vs. Deep Learning

Most machine learning methods work well because of **human-designed representations** and **input features**

ML becomes just **optimizing weights** to best make a final prediction



SURF feature matching

# Challenges of general Machine Learning

## Challenges of general ML:

- Relevant data acquisition
- Data preprocessing
- Feature selection
- Model selection: simplicity versus complexity
- Result interpretation.

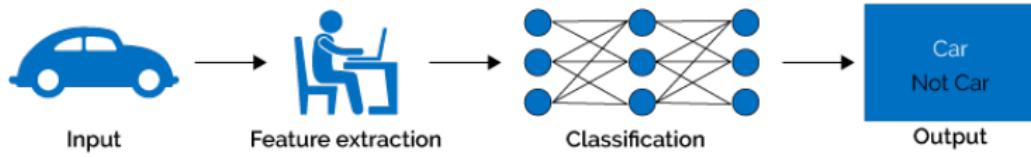
# What is Deep Learning (DL) ?

A machine learning subfield of learning **representations** of data. Exceptional effective at **learning patterns**.

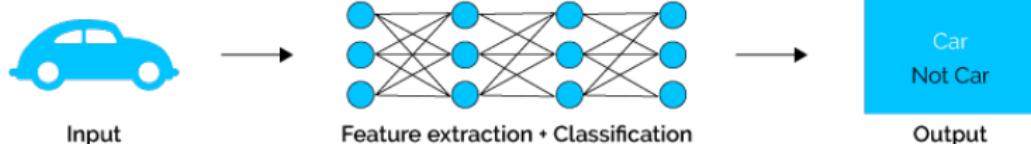
Deep learning algorithms attempt to learn (multiple levels of) representation by using a **hierarchy of multiple layers**

If you provide the system **tons of information**, it begins to understand it and respond in useful ways.

## Machine Learning

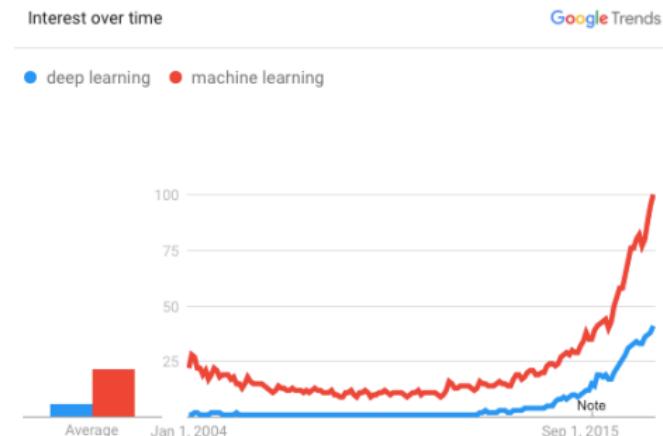


## Deep Learning



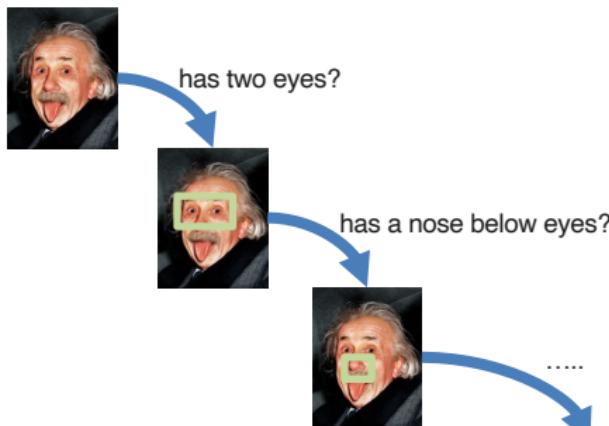
# Why is DL useful?

- Manually designed features are often **over-specified, incomplete** and take a **long time to design** and validate
- Learned Features are **easy to adapt, fast** to learn
- Deep learning provides a very **flexible**, (almost?) **universal**, learnable framework for representing world, visual and linguistic information.
- Can learn both unsupervised and supervised
- Effective **end-to-end** joint system learning
- Utilize large amounts of training data

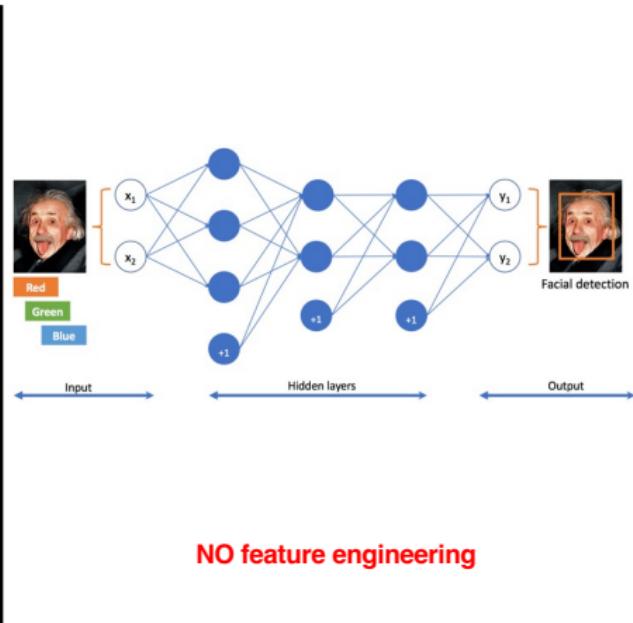


In ~2010 DL started outperforming other ML techniques first in speech and vision, then NLP

## Traditional approach

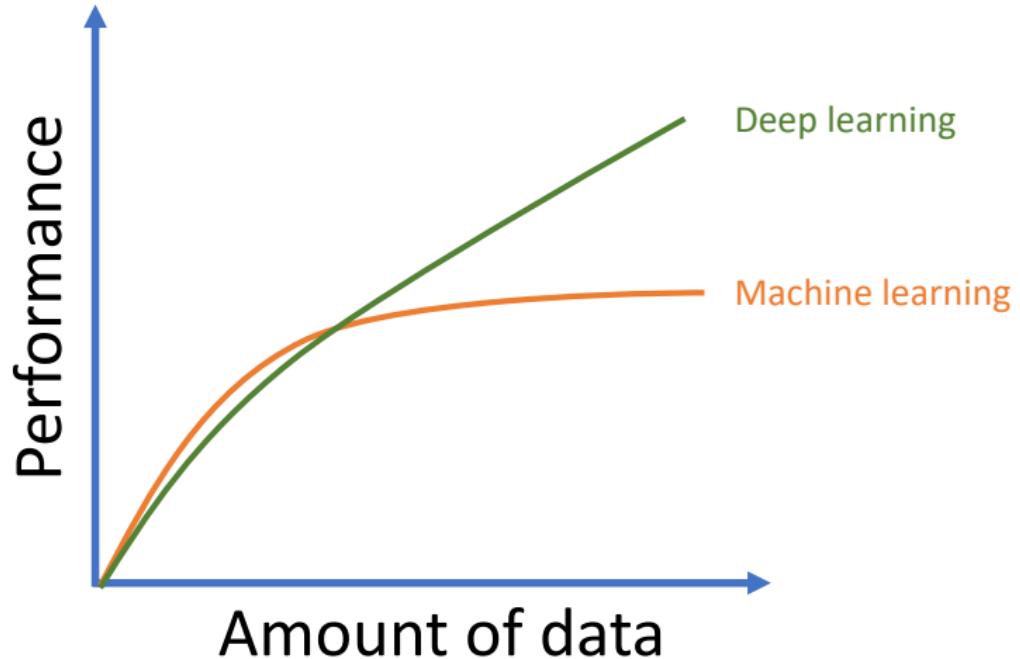


## Deep learning approach

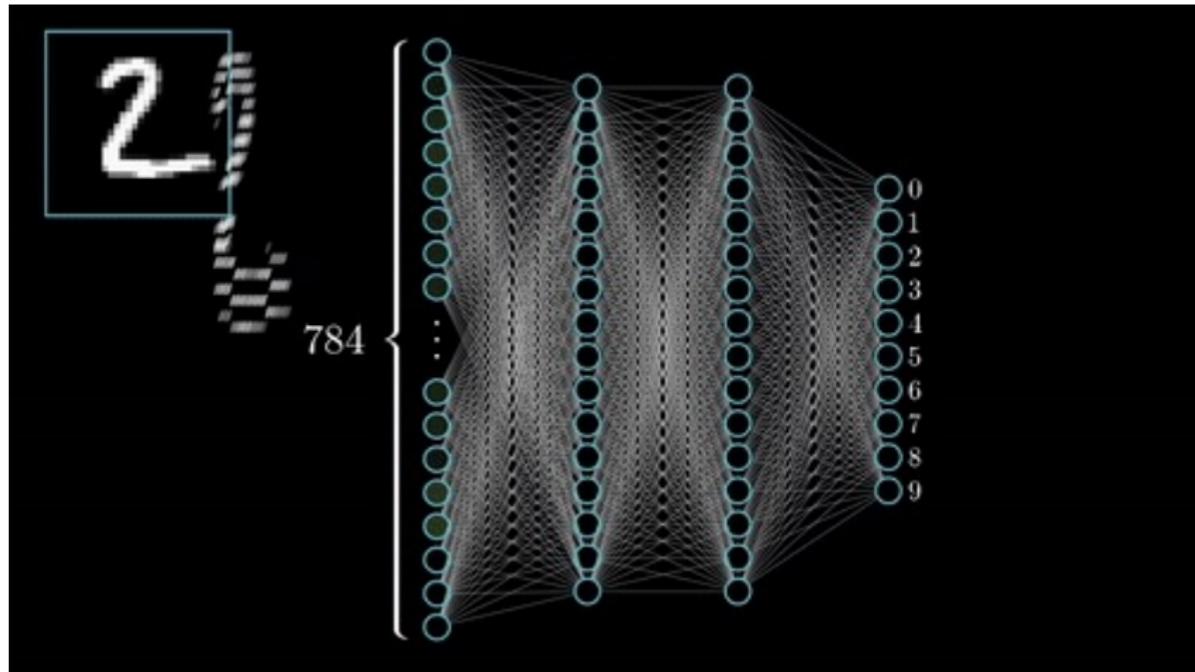


Ok, it's a face!

## Why is DL useful?

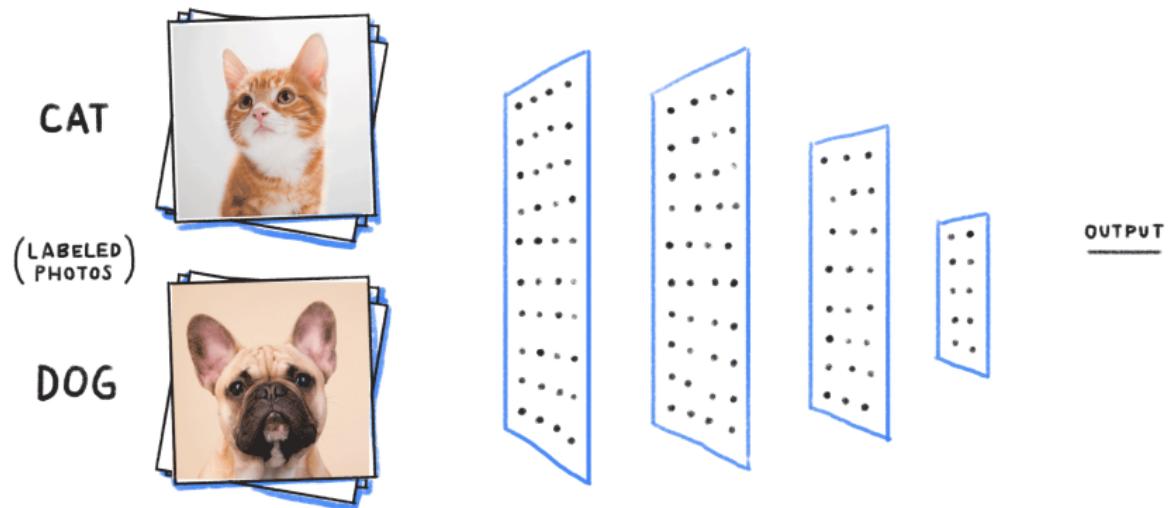


## Deep Learning as a classifier



Neural Network is trying to predict the image (data sample) that given to it. It predicts that the number is 2 here.

# Deep Learning as a classifier



Neural network is predicting cat/dog image

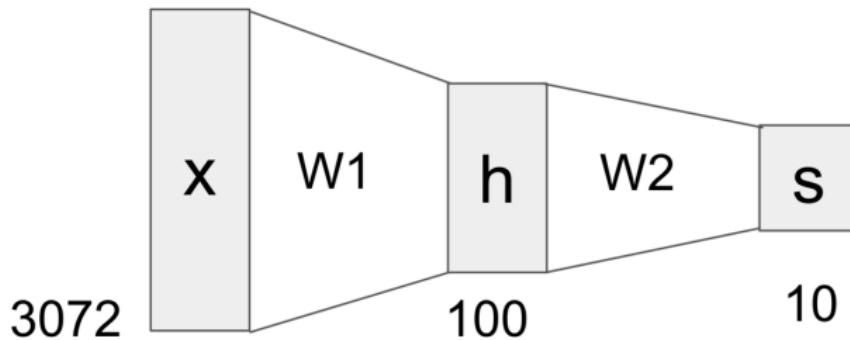
# Neural Network

Linear score function:

$$f = Wx$$

2-layer Neural Network

$$f = W_2 \max(0, W_1 x)$$



## Neural Network

Linear score function:

$$f = Wx$$

2-layer Neural Network

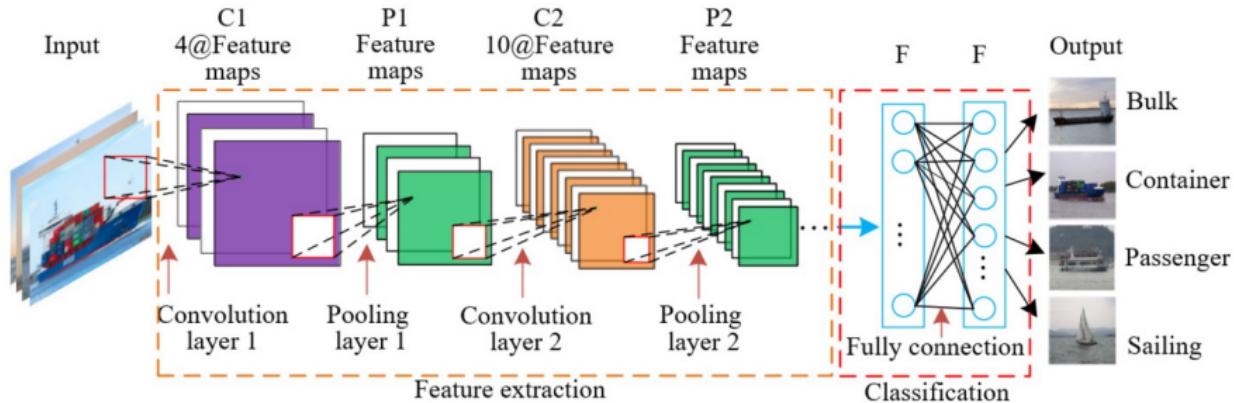
$$f = W_2 \max(0, W_1 x)$$

3-layer Neural Network

$$f = W_3 \max(0, W_2 \max(0, W_1 x))$$

... and so on.

# Deep Learning as a classifier



A typical neural network architecture for image classification task.

# Deep learning layers as Lego blocks



## Applications of AI

# AI applications

Healthcare

Automobile

Banking, Finance

Surveillance

Social Media

Entertainment

Education

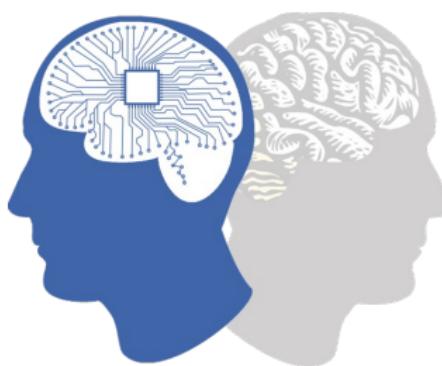
Space  
Exploration

Gaming

Robotics

Agriculture

E-commerce



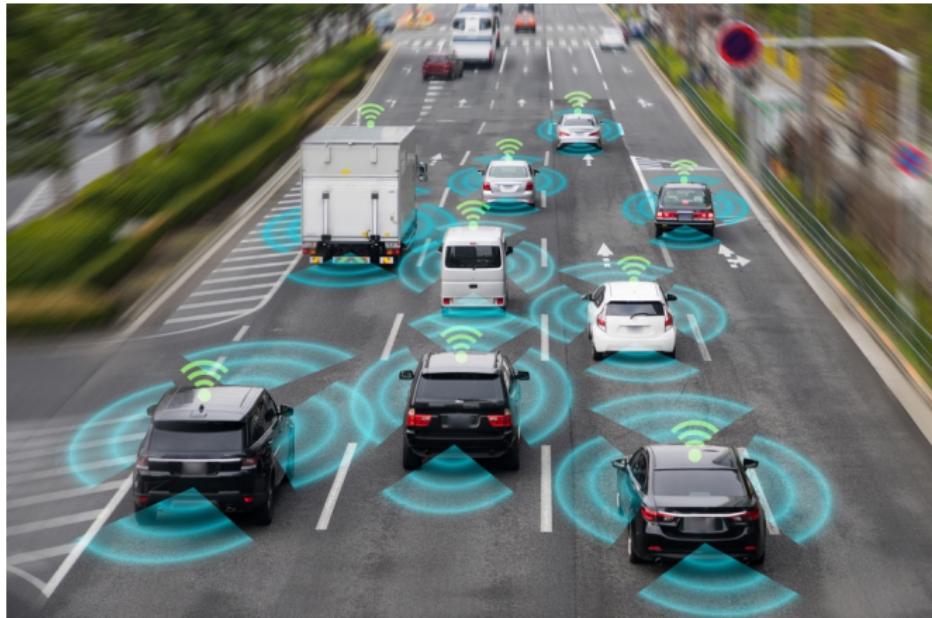
# Healthcare

- Wearable devices that can analyze sleep patterns, calories burned, heart rate.
- Medical image diagnosis.



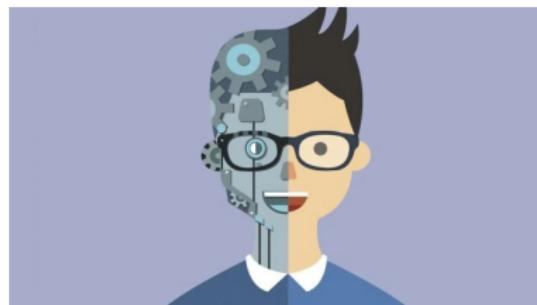
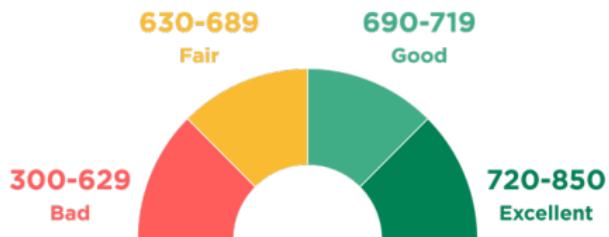
# Automobile

- Autonomous driving (a.k.a. self-driving) cars: avoid obstacles, communicate with near-by cars, etc., by applying Machine Vision and IoT technologies.



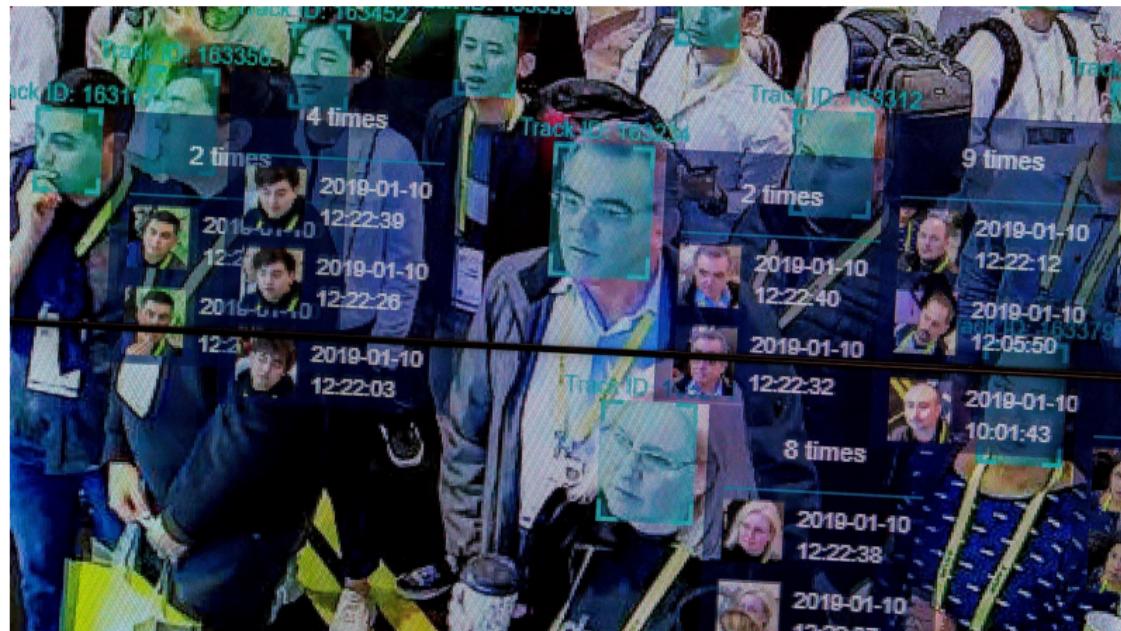
# Banking and Finance

- Credit scoring.
- Automatic customer support.



# Surveillance

- Facial recognition.
- Public safety.



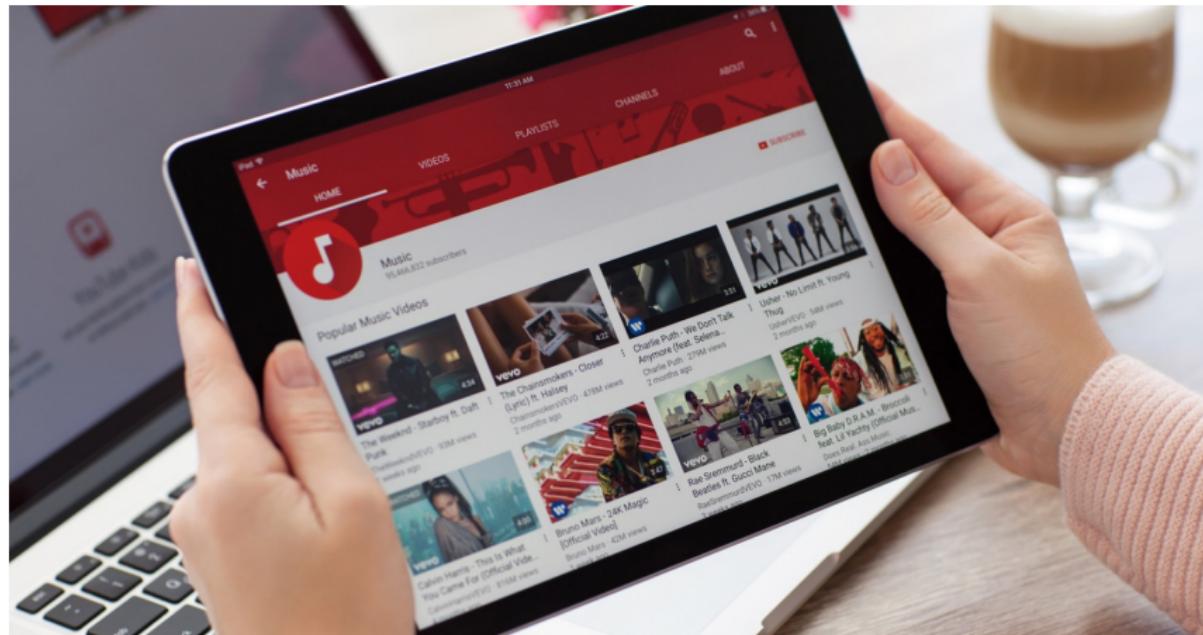
# Social media

- Social trends detection (social listening).
- User behavior understanding (for marketing/advertising purpose).



# Entertainment

- User data collection and analysis.
- Recommendation engine.



# Education

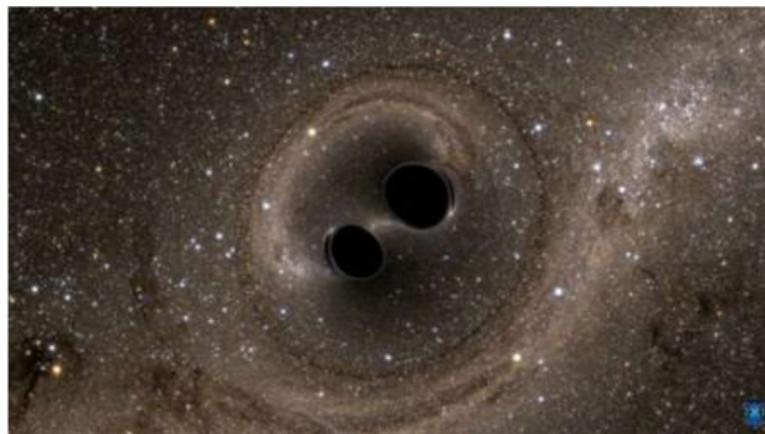
- AI-powered tutoring.
- Personalized education.



1 JANUARY 24, 2019

# Physicists use supercomputers and AI to create the most accurate model yet of black hole mergers

by Whitney Clavin



<https://phys.org/news/2019-01-physicists-supercomputers-ai-accurate-black.html>

# Gaming

- Reinforcement Learning



# Robotics

- Automate the works in factory.



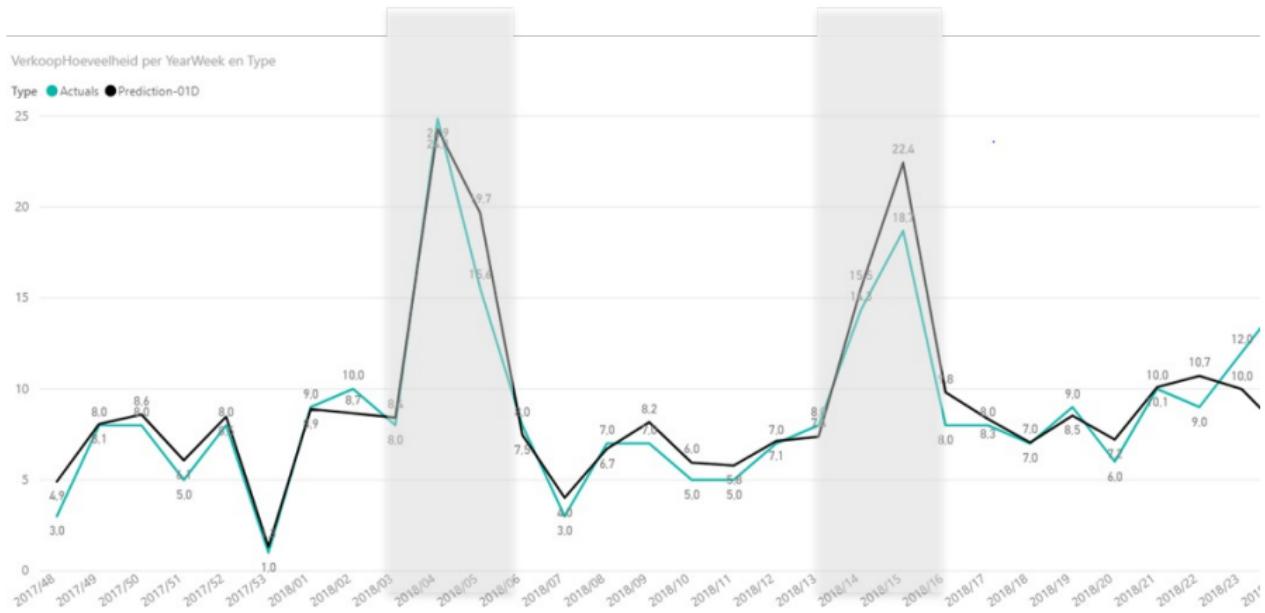
# Agriculture

- Use of drone for spraying insecticide.
- Detect weed formation in large farms.

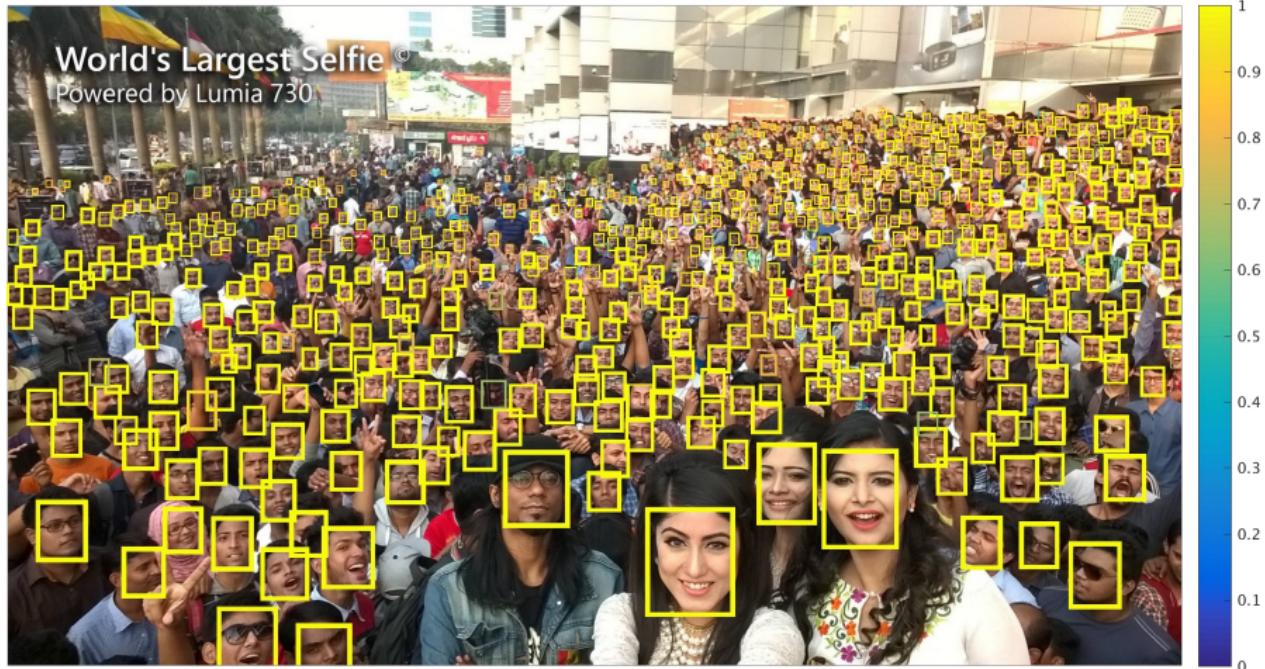


# E-commerce

- Logistics.
- Demand forecasting.
- Intelligent marketing



# Face detection



Hu, Peiyun and Ramanan, Deva, “*Finding Tiny Faces*”, CVPR’17

# Face detection



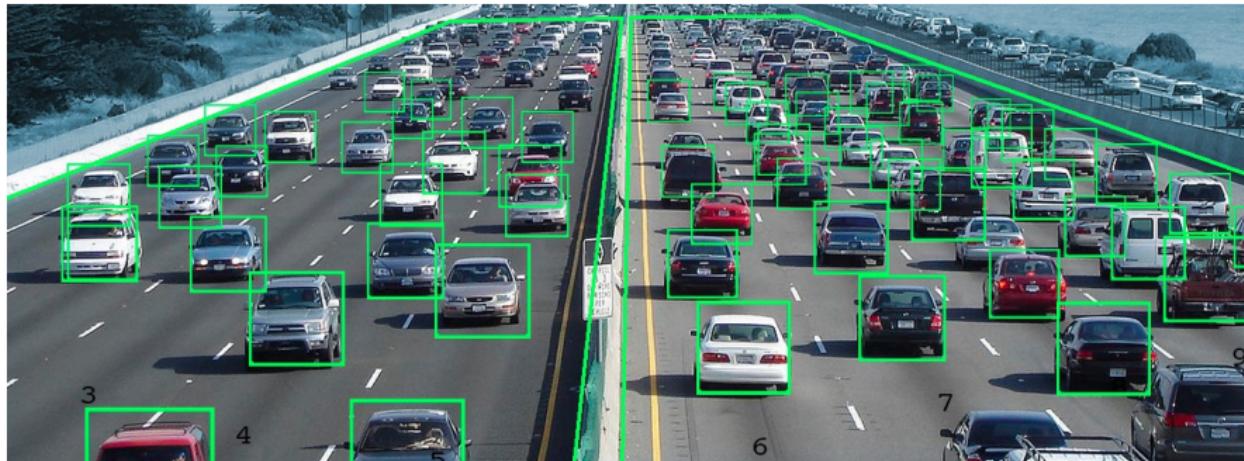
<https://devpost.com/software/face-mask-detection>

## Face generation



Tero Karras, Samuli Laine, Miika Aittala, Janne Hellsten, Jaakko Lehtinen and Timo Aila,  
*"Analyzing and Improving the Image Quality of StyleGAN"*, CVPR'20

# Traffic monitoring

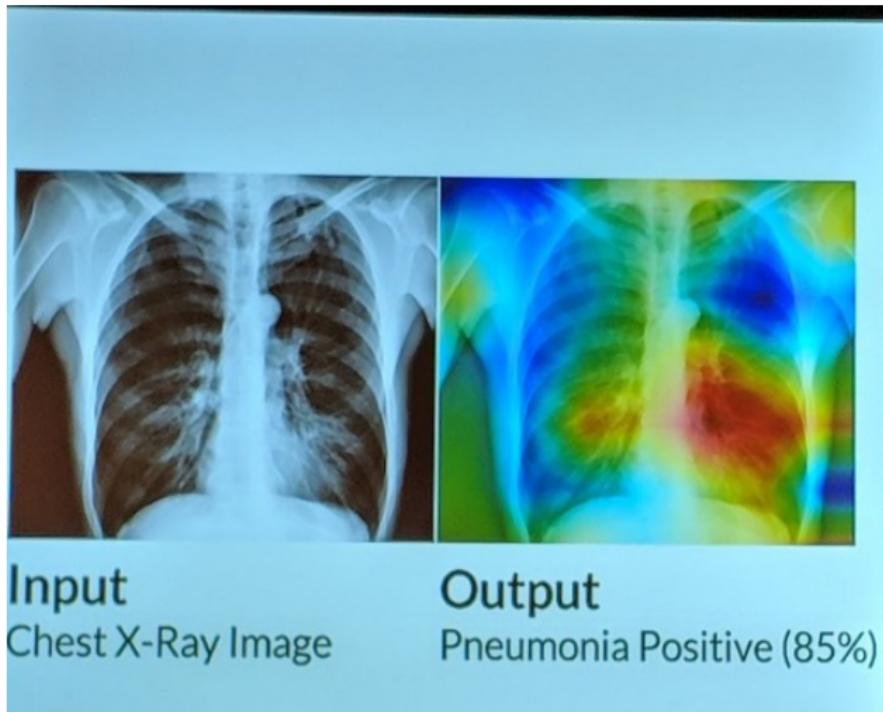


# Self-driving car



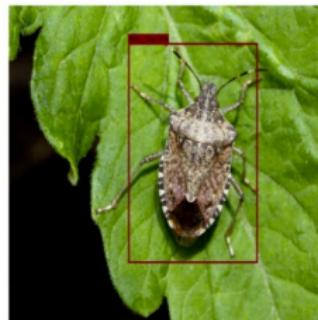
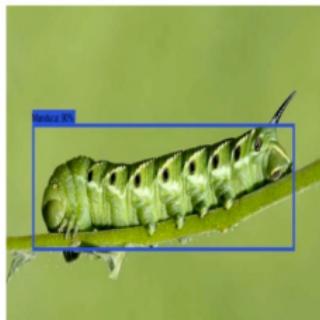
<https://www.inc.com/kevin-j-ryan/self-driving-cars-powered-by-people-playing-games-mighty-ai.html>

# Healthcare



<https://medium.com/@anushka.da3/application-of-computer-vision-in-health-care-9144899e2c54>

# Agriculture



Balakrishnan Ramalingam,,Rajesh Elara Mohan,Sathian Pookkuttath,Braulio Félix Gómez,Charan Satya Chandra Sairam Borusu,Tey Wee Teng and Yohkesh Krishnasamy Tamilselvam, "*Remote Insects Trap Monitoring System Using Deep Learning Framework and IoT*", MDPI journal.

# Retail



<https://aws.amazon.com/blogs/industries/seeing-dollar-signs-ways-to-leverage-computer-vision-in-retail-stores/>

## Artificial Intelligence: Theory vs. Reality

# Performance vs. Portability



Theory



Reality

# Performance vs. Power Consumption



Theory



Reality

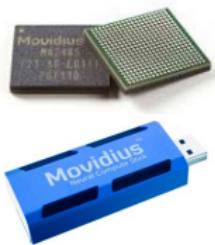
# Special Hardware for Deep Learning



Jetson TX2  
(NVIDIA)



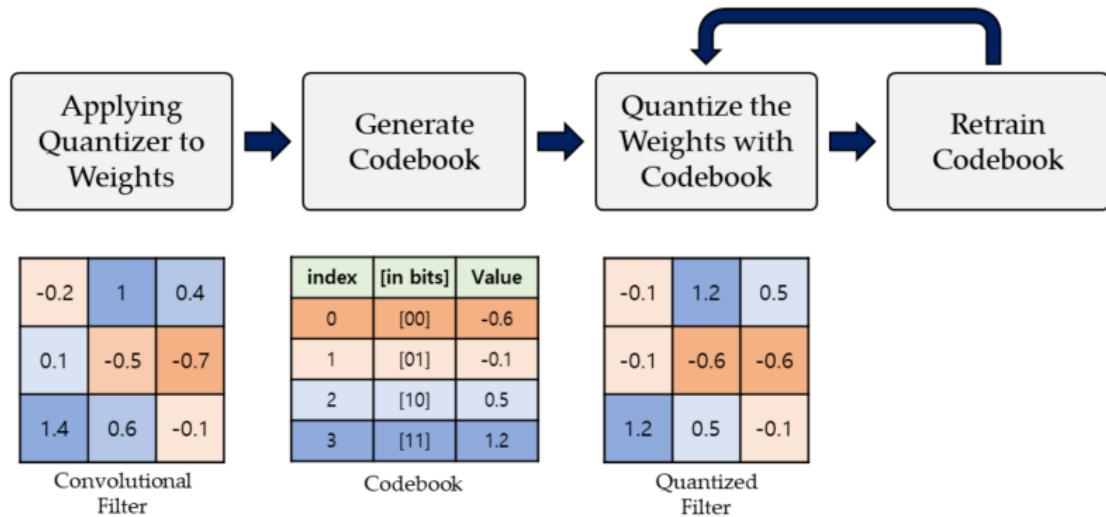
Google  
TPU



- Optimized for specific use case.
- Not plug-and-play, need good engineers to make it work.

**Still far from consumer...**

# Network quantization



# Network binarization

## (1) Binarizing Weight

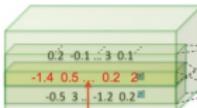
$$\begin{bmatrix} 0.1 & -0.3 & \dots & 0.2 & -0.5 \\ 0.2 & -0.4 & \dots & -0.2 & 0.1 \end{bmatrix} \rightarrow \mathbf{W}$$

$$\frac{1}{n} \|\mathbf{W}\|_{\ell_1} = \alpha$$

$$\begin{bmatrix} 1 & -1 & \dots & 1 & 1 \\ 1 & -1 & \dots & -1 & 1 \end{bmatrix} \rightarrow \mathbf{B}$$
  
$$\text{sign}(\mathbf{W})$$

## (2) Binarizing Input

Inefficient



$$\mathbf{X}_1$$

$$\mathbf{X}_2$$

$$\frac{1}{n} \|\mathbf{X}_1\|_{\ell_1} = \beta_1$$

$$\frac{1}{n} \|\mathbf{X}_2\|_{\ell_1} = \beta_2$$

$$\mathbf{K}$$

Redundant computations in overlapping areas



$$\text{sign}(\mathbf{X}_1) = \mathbf{H}_1$$

$$\text{sign}(\mathbf{X}_2) = \mathbf{H}_2$$

$$\text{sign}(\mathbf{I})$$

## (3) Binarizing Input

Efficient

$$\sum_{k=1}^c |I_{i,:k}| \rightarrow \mathbf{A}$$

$$= \mathbf{A}$$

$$\mathbf{A} * \underset{k}{\mathbf{K}} = \beta_1 \quad \beta_2$$

$$\begin{bmatrix} 1 & -1 & \dots & 1 & 1 \\ 1 & -1 & \dots & 1 & 1 \\ -1 & 1 & \dots & 1 & 1 \\ -1 & 1 & \dots & -1 & 1 \end{bmatrix} \rightarrow \mathbf{K}$$

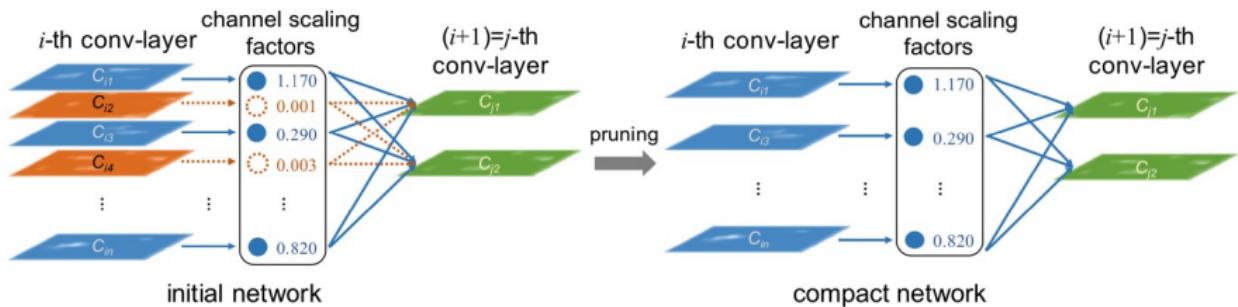
$$\text{sign}(\mathbf{I})$$

## (4) Convolution with XNOR-Bitcount

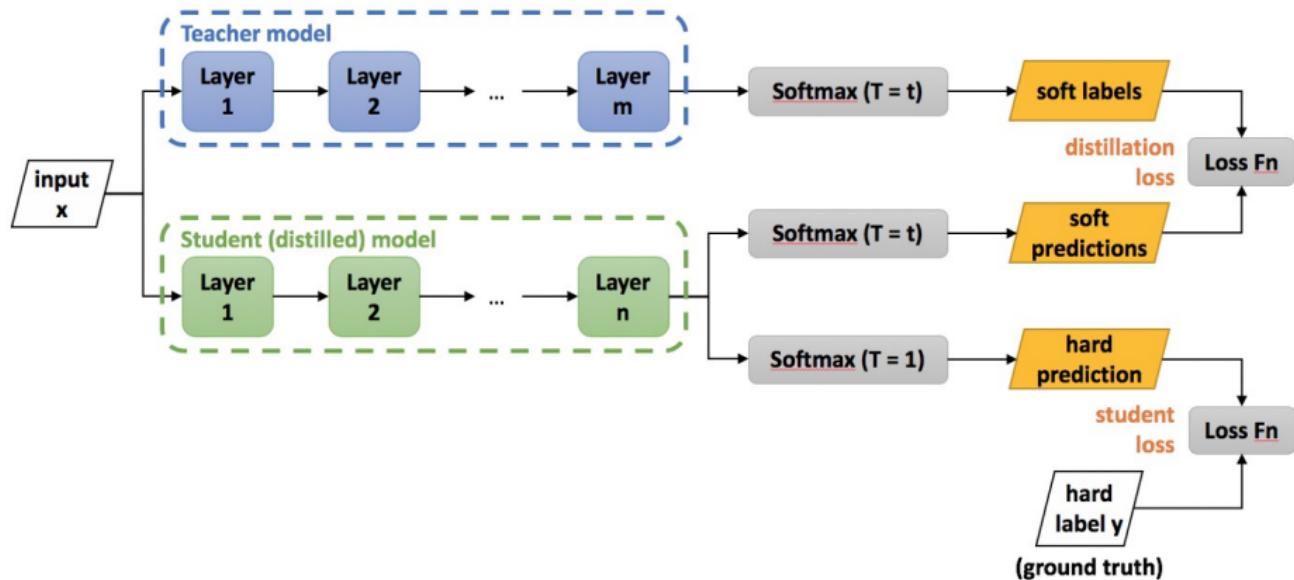
$$\begin{bmatrix} 0.2 & -0.1 & \dots & 3 & 0.1 \\ -1.4 & 0.5 & \dots & 0.2 & 2 \\ -0.5 & 3 & \dots & -1.2 & 0.2 \end{bmatrix} \otimes \begin{bmatrix} 0.1 & -0.3 & \dots & 0.2 & -0.5 \\ 0.2 & -0.4 & \dots & -0.2 & 0.1 \end{bmatrix} \approx$$

$$\left[ \begin{bmatrix} 1 & -1 & \dots & 1 & 1 \\ 1 & -1 & \dots & 1 & 1 \\ -1 & 1 & \dots & 1 & 1 \\ -1 & 1 & \dots & -1 & 1 \end{bmatrix} \otimes \begin{bmatrix} 1 & -1 & \dots & 1 & 1 \\ 1 & -1 & \dots & 1 & 1 \\ 1 & -1 & \dots & -1 & 1 \end{bmatrix} \right] \odot \mathbf{K} \odot \alpha$$
  
$$\text{sign}(\mathbf{I}) \otimes \text{sign}(\mathbf{W})$$

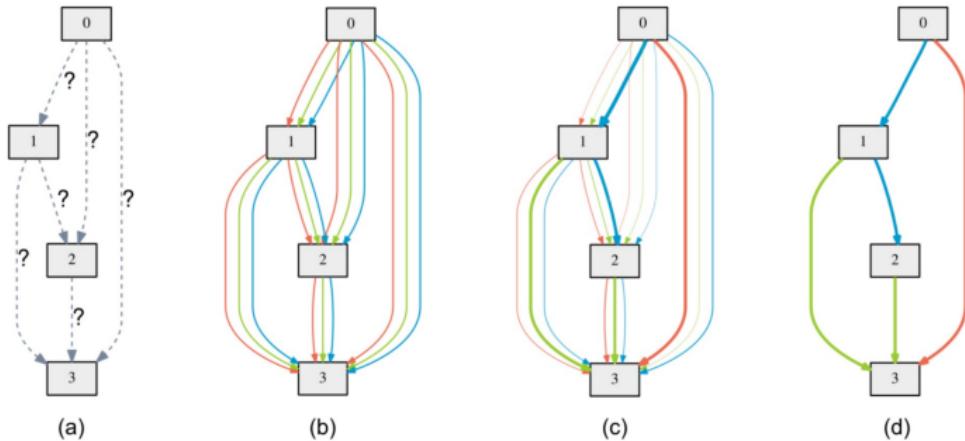
# Network pruning



# Knowledge Distillation



# Network Architecture Search (NAS)



## Summary

- Introduction to deep learning and neural networks.
- Feature engineering in machine learning and deep learning.
- Challenges in machine learning and deep learning.
- Applications of AI.
- AI: Theory vs. Reality.

## Q&A

Thank you