

The Power and Limits of Machine Learning

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Agenda

- 1 What is Machine Learning?
- 2 The Power of Machine Learning
- 3 The Limits of Machine Learning
- 4 The Journey Ahead

What is Machine Learning?



Artificial Intelligence

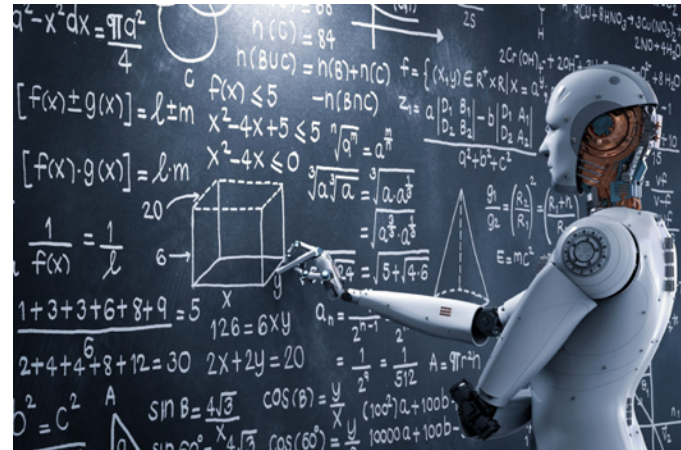
Artificial intelligence (AI) enables machines to perform some **cognitive functions** similar to those attributed to humans, as opposed to conventional machines which act according to how they are programmed to act.

“AI is the new electricity.” – Andrew Ng

History of AI

AI is as old as the field of computer science (CS).

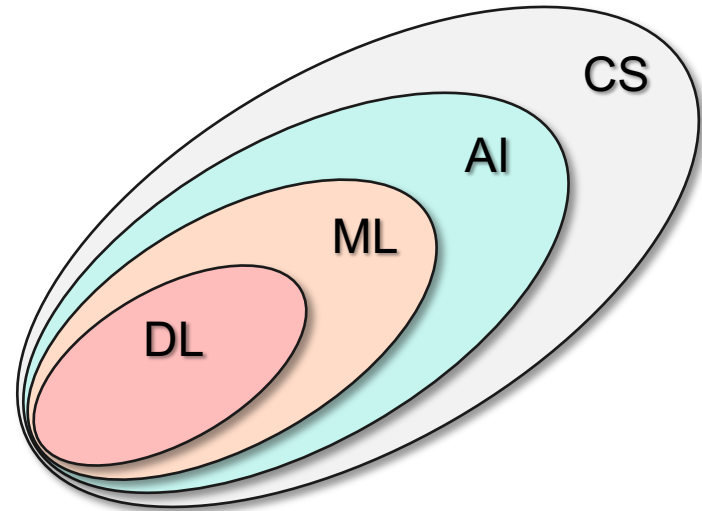
Many pioneers in CS are also pioneers in AI, e.g., Alan Turing, John McCarthy, Herbert Simon, Marvin Minsky.



AI, Machine Learning, and Deep Learning

Machine learning (ML) marries algorithms in CS, mathematical and statistical modeling, and learning from data/examples.

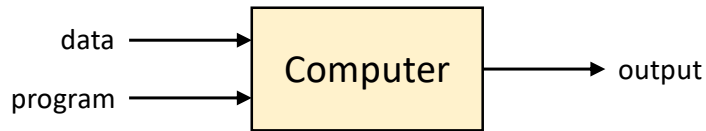
Deep learning (DL) is a subarea of ML – representation learning often using relatively deep, layered network architectures.



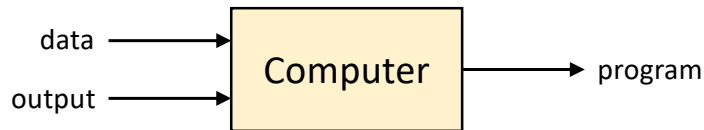


Conventional Programming vs. Supervised Learning

Conventional programming



Supervised learning



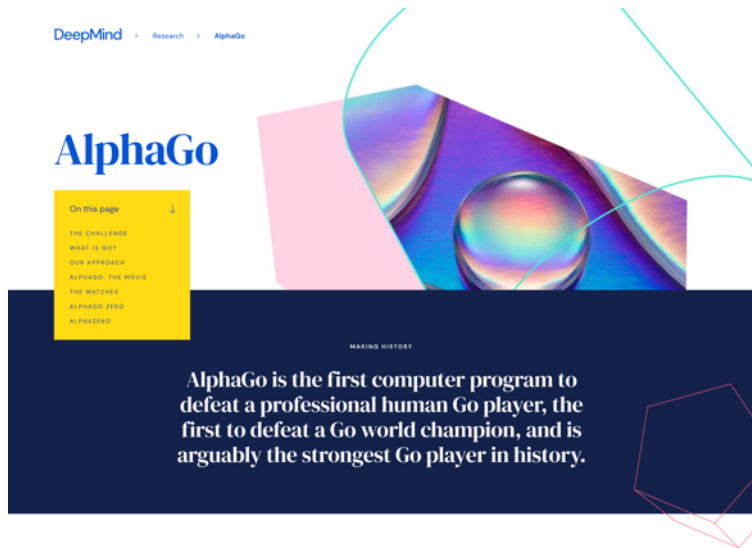
Other learning paradigms:

- Unsupervised learning
- Reinforcement learning
- Semi-supervised learning
- ...




The Power of Machine Learning

AlphaGo



From AlphaGo to AlphaGo Zero and AlphaZero

DeepMind > Blog > AlphaGo Zero: Starting from scratch



AlphaGo Zero: Starting from scratch

18 OCT 2017


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AUTHORS

- DS David Silver
- TH Thomas Hubert
- JS Julian Schrittwieser
- DM Demis Hassabis

FURTHER READING

- Deep Learning
- Deep Reinforcement Learning
- Games
- Planning
- Reinforcement Learning




40 days

AlphaGo Zero surpasses all other versions of AlphaGo and, arguably, becomes the best Go player in the world. It does this entirely from self-play, with no human intervention and using no historical data.

| Days | AlphaGo Zero 40 blocks | AlphaGo Lee | AlphaGo Master |
|------|------------------------|-------------|----------------|
| 0 | -2000 | 3500 | 4500 |
| 5 | 3500 | 3500 | 4500 |
| 10 | 4500 | 3500 | 4500 |
| 15 | 4800 | 3500 | 4500 |
| 20 | 4900 | 3500 | 4500 |
| 25 | 4950 | 3500 | 4500 |
| 30 | 4980 | 3500 | 4500 |
| 35 | 4990 | 3500 | 4500 |
| 40 | 5000 | 3500 | 4500 |

DeepMind > Blog > AlphaZero: Shedding new light on chess, shogi, and Go



AlphaZero: Shedding new light on chess, shogi, and Go

06 DEC 2018

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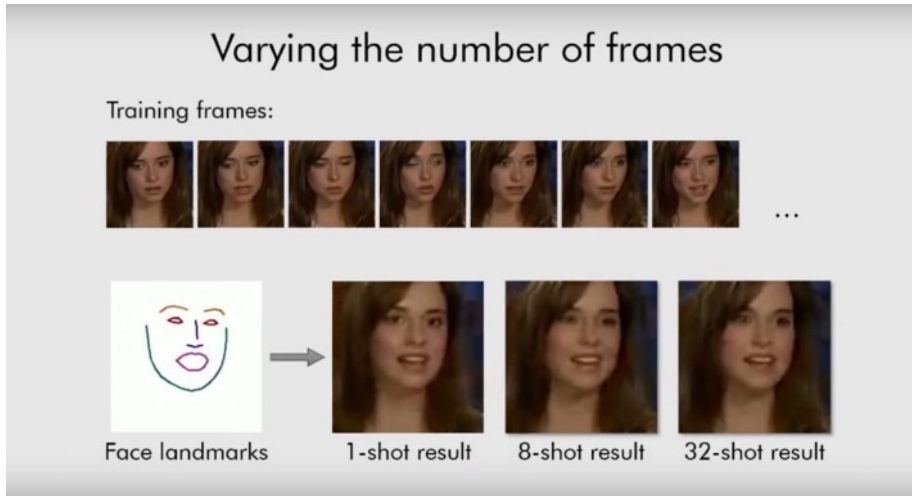
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In late 2017 we introduced AlphaZero, a single system that taught itself from scratch how to master the games of chess, shogi (Japanese chess), and Go, beating a world-champion program in each case. We were excited by the preliminary results and thrilled to see the response from members of the chess community, who saw in AlphaZero's games a ground-breaking, highly dynamic and "unconventional" style of play that differed from any chess playing engine that came before it.



Talking Heads with Motion



[YouTube video](#)

Zakharov et al., “Few-shot adversarial learning of realistic neural talking head models”, ICCV, 2019.



Generating Photorealistic Deepfakes



[YouTube video](#)

Karras et al., “Analyzing and improving the image quality of StyleGAN”, CVPR, 2020.



AI Assistant (Google Duplex)

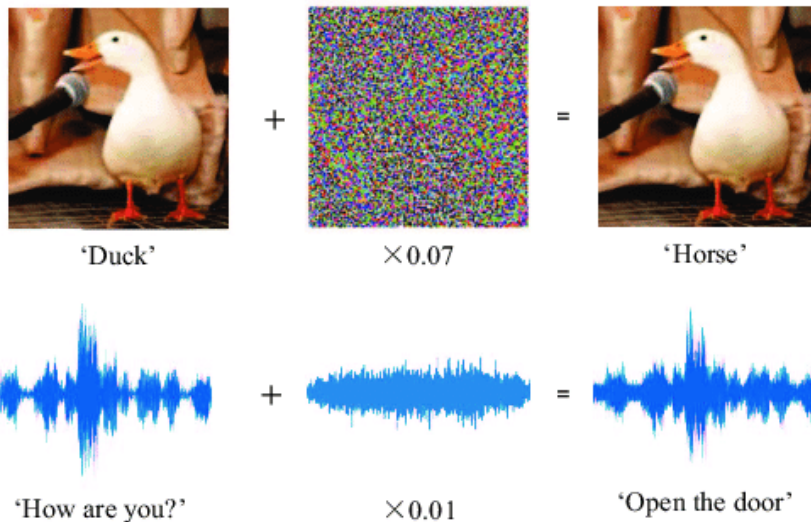


[YouTube video](#)

The Limits of Machine Learning






Adversarial Examples (or Adversarial Attacks)

Imperceptible perturbations
(which are carefully generated)
added to images or audio signals
can fool even state-of-the-art
classifiers to give **incorrect**
predictions.



Safety Concerns of Adversarial Examples

Adversarial Examples

| | | | |
|-----------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------|------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------|
|  |  |  |  |
| Clean Stop Sign | Real-world Stop Sign in Berkeley | Adversarial Example | Adversarial Example |
|  "Stop sign" | "Stop sign" | "Speed limit sign 45km/h" | "Speed limit sign 45km/h" |

Physical attacks on STOP sign

One-pixel Attacks

Changing just one pixel
(marked by red circle)

Su et al., "One pixel attack for fooling deep neural network", IEEE T-EC, 2019.



Cup(16.48%)
Soup Bowl(16.74%)



Bassinet(16.59%)
Paper Towel(16.21%)



Teapot(24.99%)
Joystick(37.39%)

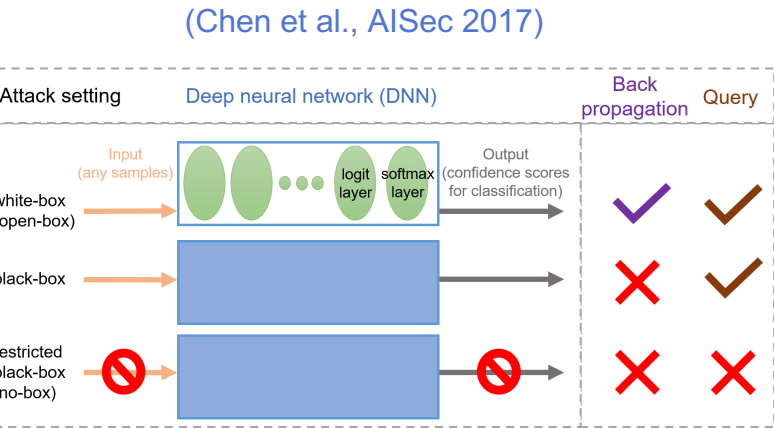


Hamster(35.79%)
Nipple(42.36%)

White-box vs. black-box attacks

White-box attacks

- Have **full** knowledge of internal structure of target model when generating adversarial attacks
- **Worst-case** scenario



Black-box attacks

- Have **no** knowledge of internal structure of target model when generating adversarial attacks
- **More realistic** scenario

How are Adversarial Examples Generated?

Three major approaches:

1. **Optimization**-based approach, e.g.,

$$\begin{array}{l} \min_{x'} c \|\eta\| + J_{\theta}(x', l') \\ \text{s.t. } x' \in [0, 1]. \end{array}$$

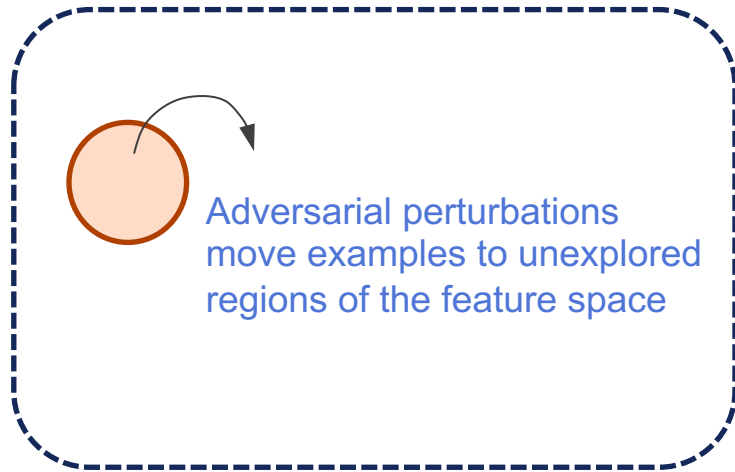
2. **Gradient**-based approach, e.g.,

$$\eta = \epsilon \text{sign}(\nabla_x J_{\theta}(x, l))$$

3. **Generative** approach, e.g., using a generative model



What Makes Adversarial Attacks Possible?



Adversarial perturbations
move examples to unexplored
regions of the feature space

Feature space

Theoretical study of underlying reasons for adversarial attacks is still rare and immature – **good research topic to work on.**



Defenses Against Adversarial Attacks

Two major approaches:

1. Retraining the model, e.g., **adversarial training**, **defensive distillation**
2. Learning to purify the adversarial examples before feeding them into the model, e.g., **MagNet**, **PixelDefend**

Adversarial Attacks Beyond Images and Audio Signals

Attacking reading comprehension systems

Article: Super Bowl 50

Paragraph: *“Peyton Manning became the first quarterback ever to lead two different teams to multiple Super Bowls. He is also the oldest quarterback ever to play in a Super Bowl at age 39. The past record was held by John Elway, who led the Broncos to victory in Super Bowl XXXIII at age 38 and is currently Denver’s Executive Vice President of Football Operations and General Manager. [Quarterback Jeff Dean had jersey number 37 in Champ Bowl XXXIV.](#)”*

Question: *“What is the name of the quarterback who was 38 in Super Bowl XXXIII?”*

Original Prediction: John Elway

Prediction under adversary: Jeff Dean

Other adversarial attacks:

- Machine translation
- Text summarization
- Malware detection
- Spam detection
- Reinforcement learning
- ...

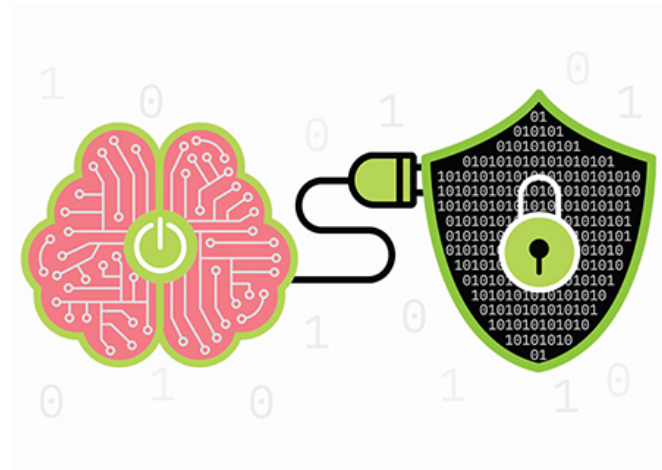
(Jia and Liang, EMNLP 2017)

The Journey Ahead

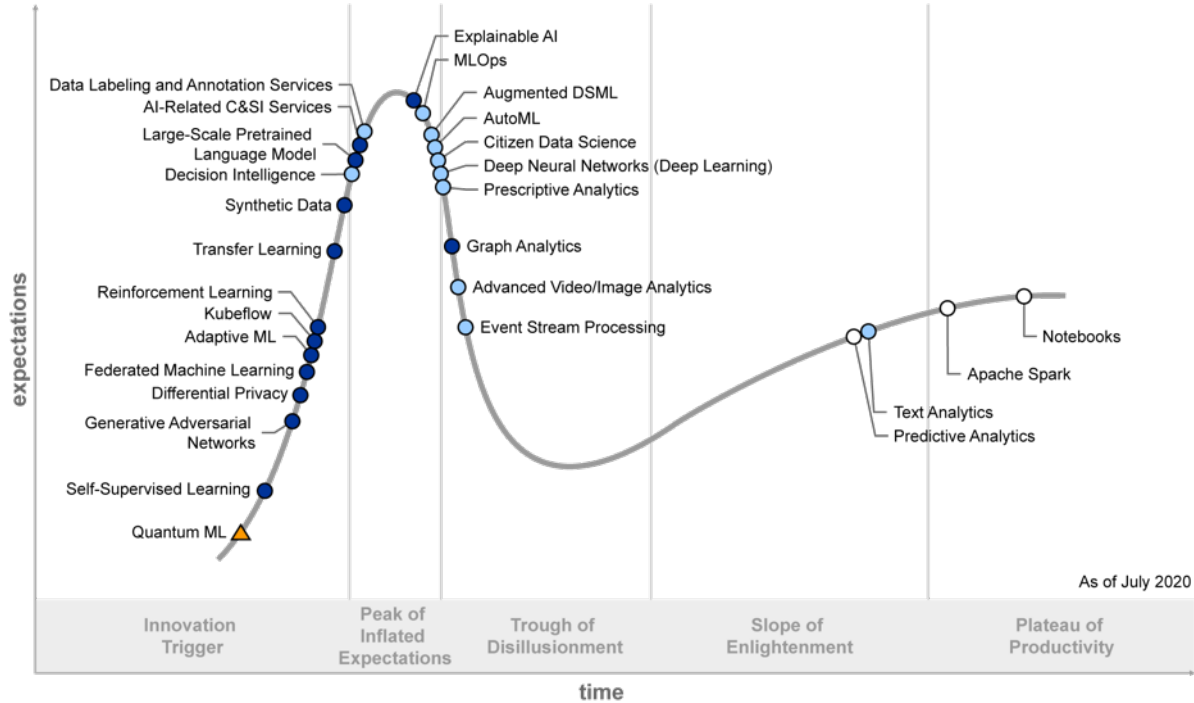


Interplay between ML and Cybersecurity

The study of security, privacy, robustness, resilience, and reliability will be central to the field of machine learning



Hype Cycle for Data Science and Machine Learning, 2020



Plateau will be reached:

- less than 2 years
- 2 to 5 years
- 5 to 10 years
- ▲ more than 10 years
- ⊗ obsolete before plateau

Source: Gartner
ID: 450404



Q&A

