Part VI: What's Next?

Outline

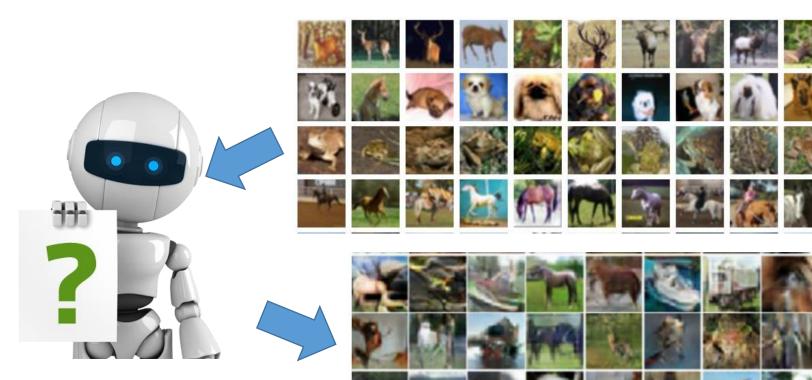
Deep Generative Models

 Richard Feynman: "What I cannot create, I do not understand."

Deep Reinforcement Learning

• The technique makes Alpha Go better than professional players.

Creation This is Unsupervised Leaning.



Draw something!

http://www.wikihow.com/Draw-a-Cat-Face

Creation

Now





V.S.



If machine can draw a cat ...



Deep Generative Models

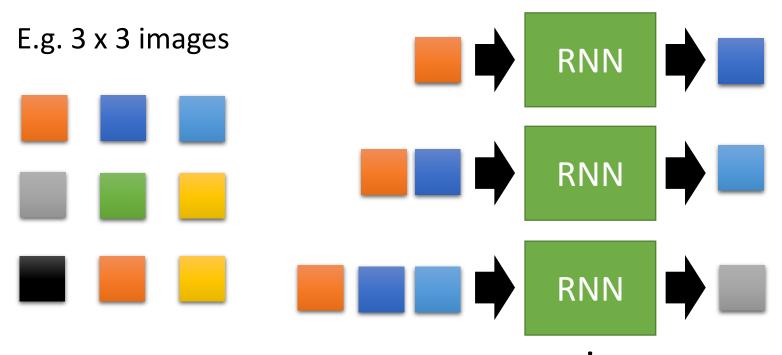
Component-by-component

Auto-encoder

Adversarial Generative Network (GAN)

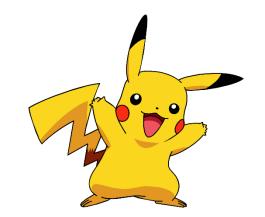
Component-by-component

• To create an image, generating a pixel each time



Can be trained with a large collection of images without any annotation

Pokémon Creation



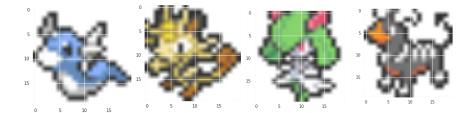
- Small images of 792 Pokémon's
 - Can machine learn to create new Pokémons?

Don't catch them! Create them!

 Source of image: http://bulbapedia.bulbagarden.net/wiki/List_of_Pok%C3%A 9mon_by_base_stats_(Generation_VI)

Original image is 40 x 40

Making them into 20 x 20

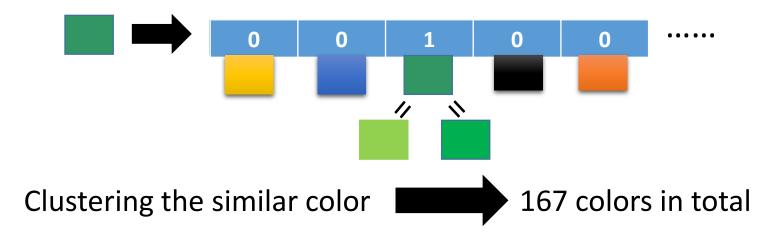


Pokémon Creation

Each pixel is represented by 3 numbers (corresponding to RGB)

R=50, G=150, B=100

Each pixel is represented by a 1-of-N encoding feature

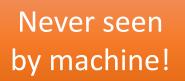


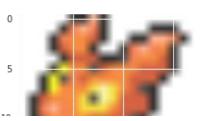
Pokémon Creation - Data

- Original image (40 x 40): <u>http://speech.ee.ntu.edu.tw/~tlkagk/courses/ML_2016/Pokemon_creation/image.rar</u>
- Pixels (20 x 20): <u>http://speech.ee.ntu.edu.tw/~tlkagk/courses/ML_2016/Pokemon_creation/pixe</u>
 <u>l_color.txt</u>
 - Each line corresponds to an image, and each number corresponds to a pixel
 - http://speech.ee.ntu.edu.tw/~tlkagk/courses/ML_2016/Pokemon_cre ation/colormap.txt

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Real Pokémon

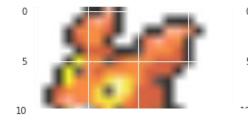








Cover 50%





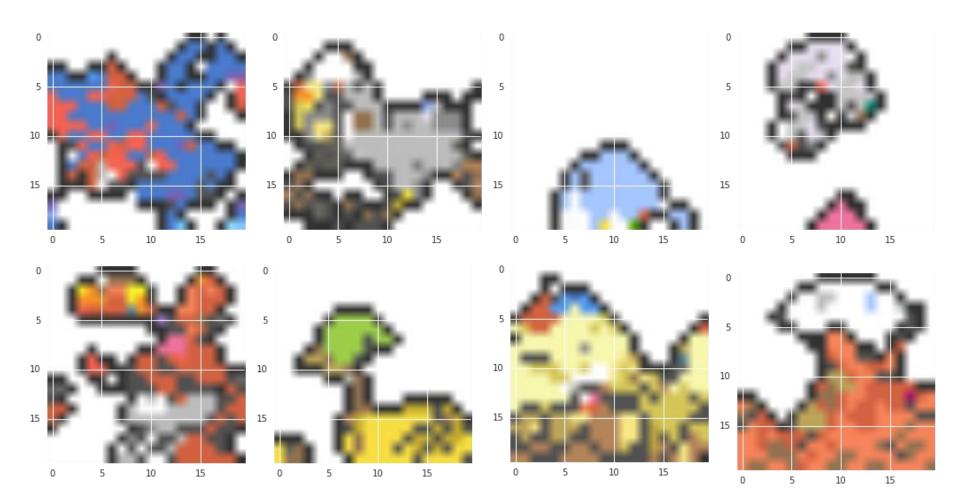




Cover 75%

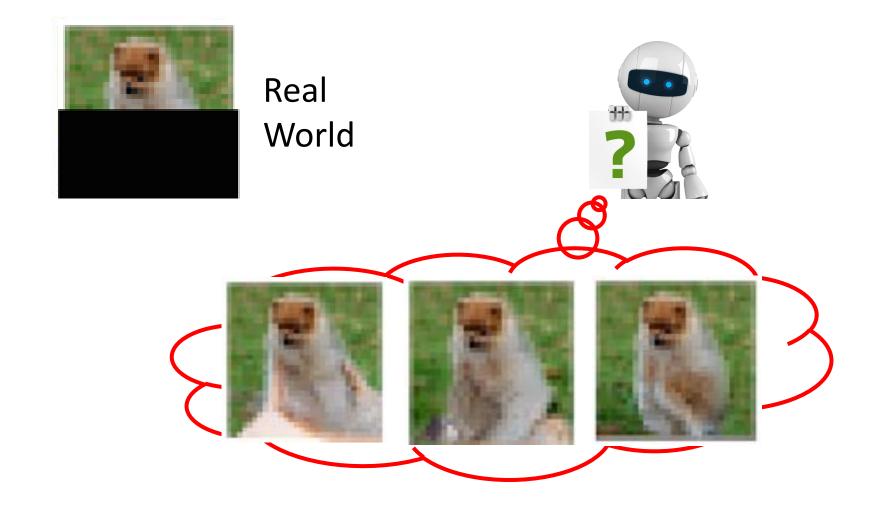
Pokémon Creation

Drawing from scratch Need some randomness

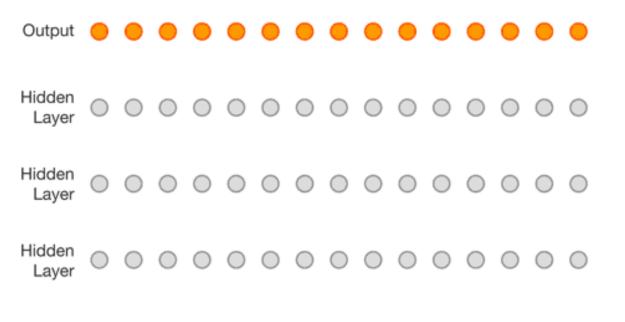


PixelRNN

Ref: Aaron van den Oord, Nal Kalchbrenner, Koray Kavukcuoglu, Pixel Recurrent Neural Networks, arXiv preprint, 2016



More than images



Input O O O O O O O O O O O O O O O O O O

Audio: Aaron van den Oord, Sander Dieleman, Heiga Zen, Karen Simonyan, Oriol Vinyals, Alex Graves, Nal Kalchbrenner, Andrew Senior, Koray Kavukcuoglu, WaveNet: A Generative Model for Raw Audio, arXiv preprint, 2016

Video: Nal Kalchbrenner, Aaron van den Oord, Karen Simonyan, Ivo Danihelka, Oriol Vinyals, Alex Graves, Koray Kavukcuoglu, Video Pixel Networks, arXiv preprint, 2016

Deep Generative Models

Criticism: do not consider the generation process globally

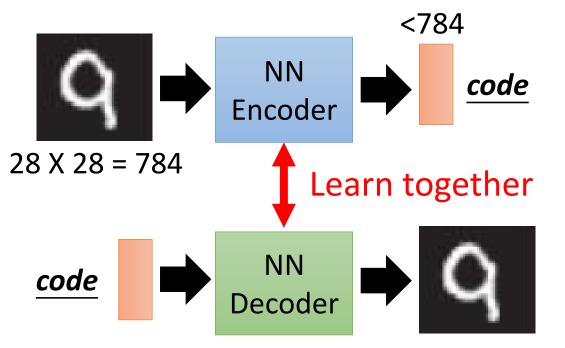
Component-by-component

Auto-encoder

Dimension reduction

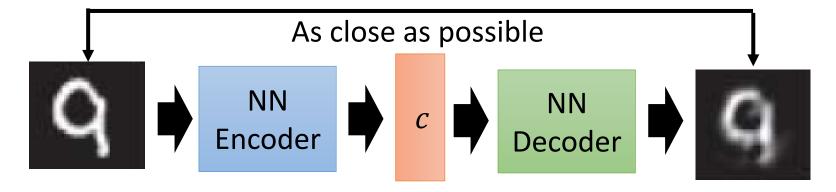
Adversarial Generative Network (GAN)

Auto-encoder



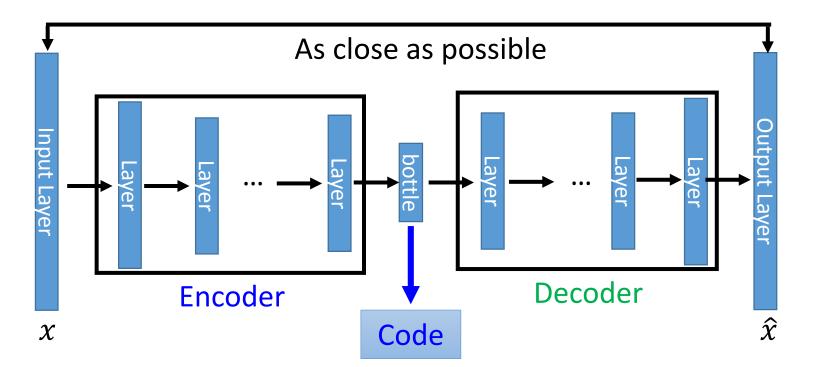
Representation of the input object

Can reconstruct the original object

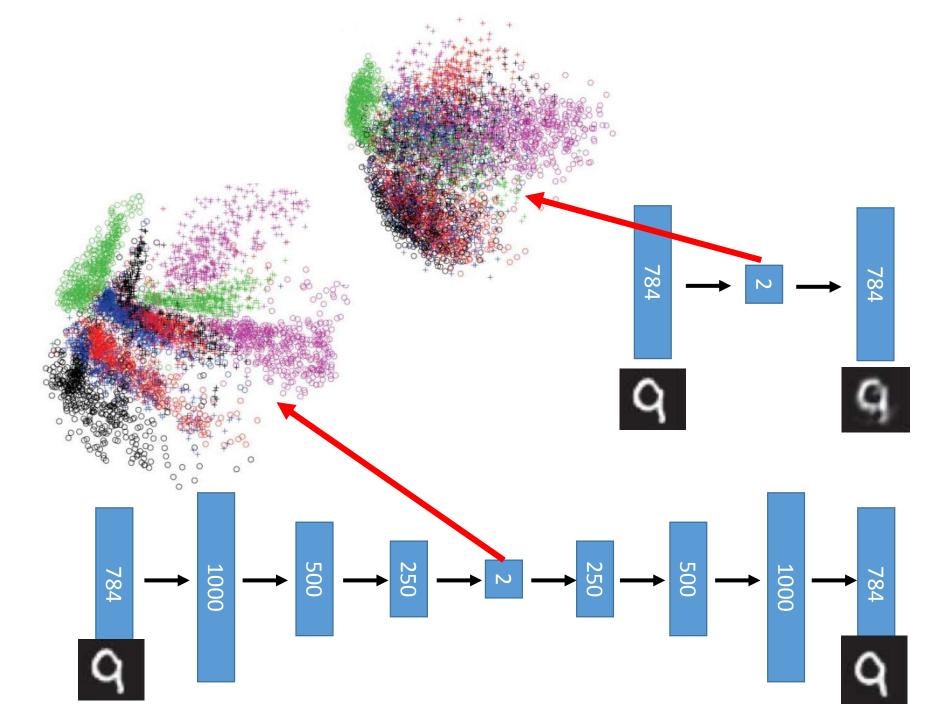


Deep Auto-encoder

• NN encoder + NN decoder = a deep network

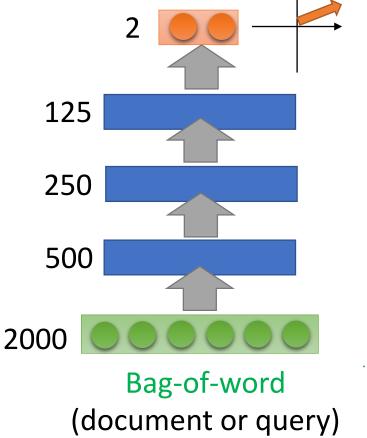


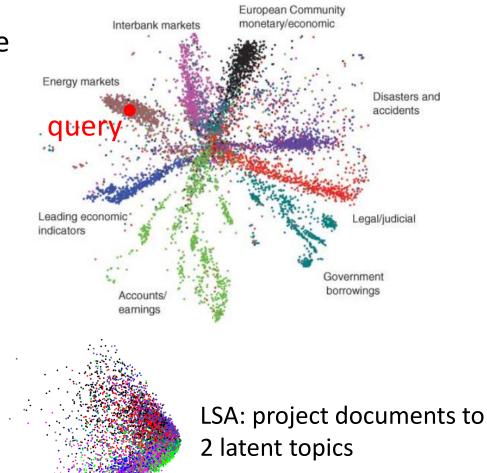
Reference: Hinton, Geoffrey E., and Ruslan R. Salakhutdinov. "Reducing the dimensionality of data with neural networks." *Science* 313.5786 (2006): 504-507



Auto-encoder – Text Retrieval

The documents talking about the same thing will have close code.



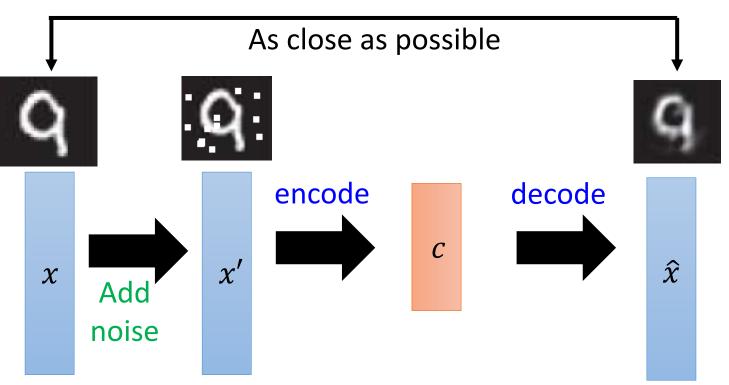


Auto-encoder

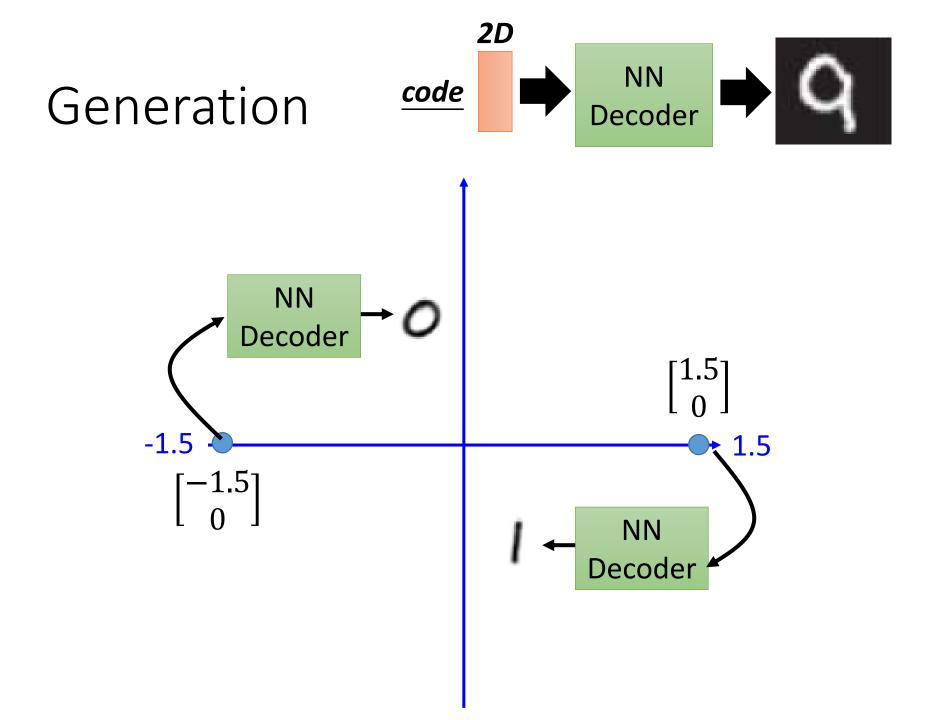
More: Contractive auto-encoder

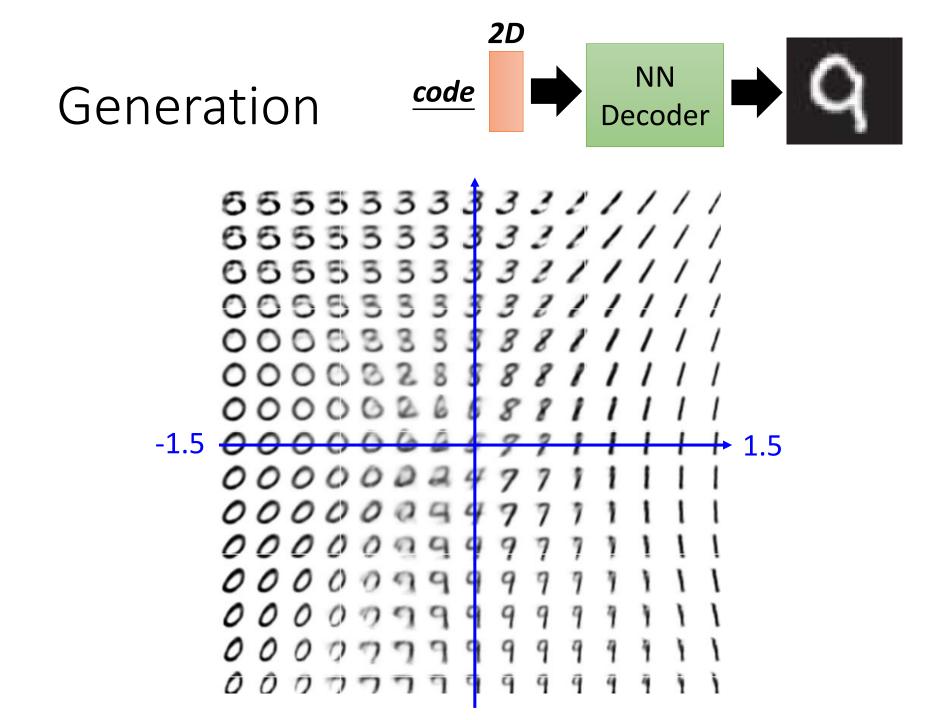
Ref: Rifai, Salah, et al. "Contractive auto-encoders: Explicit invariance during feature extraction." *Proceedings of the 28th International Conference on Machine Learning (ICML-11)*. 2011.

De-noising auto-encoder

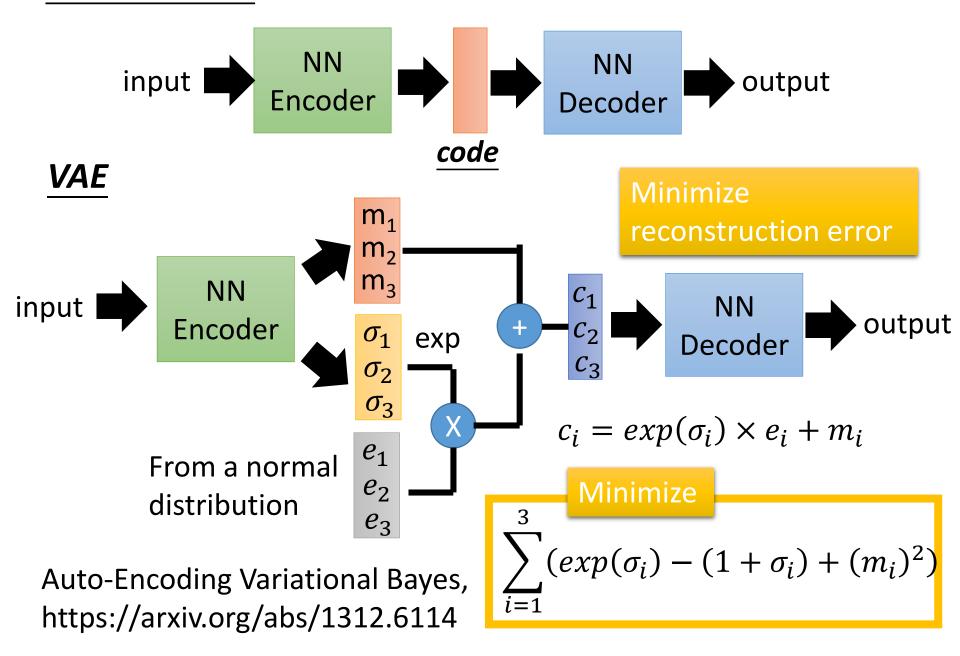


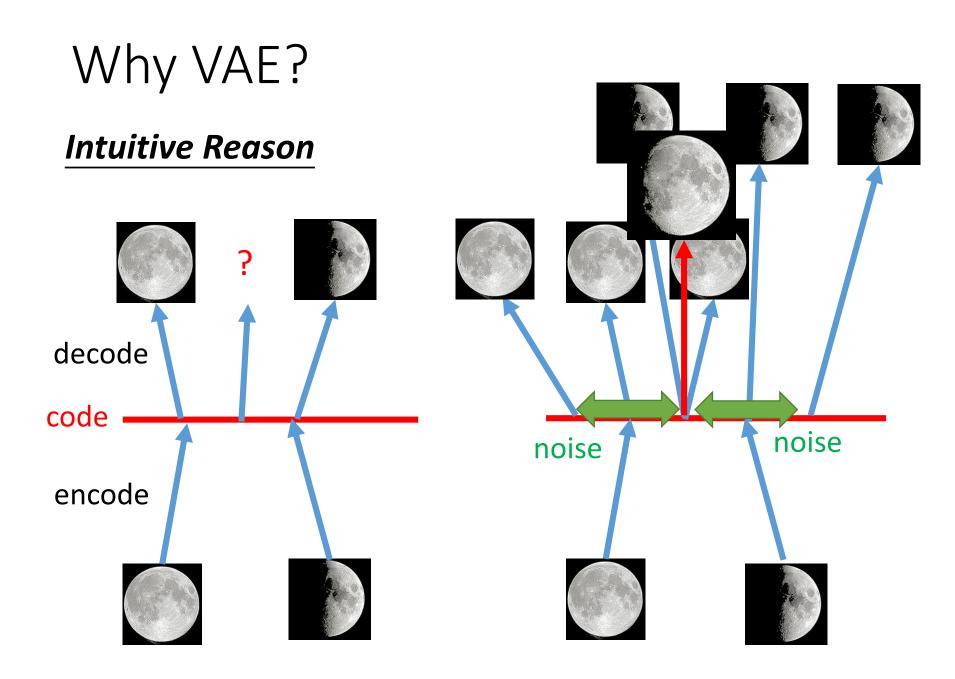
Vincent, Pascal, et al. "Extracting and composing robust features with denoising autoencoders." *ICML*, 2008.



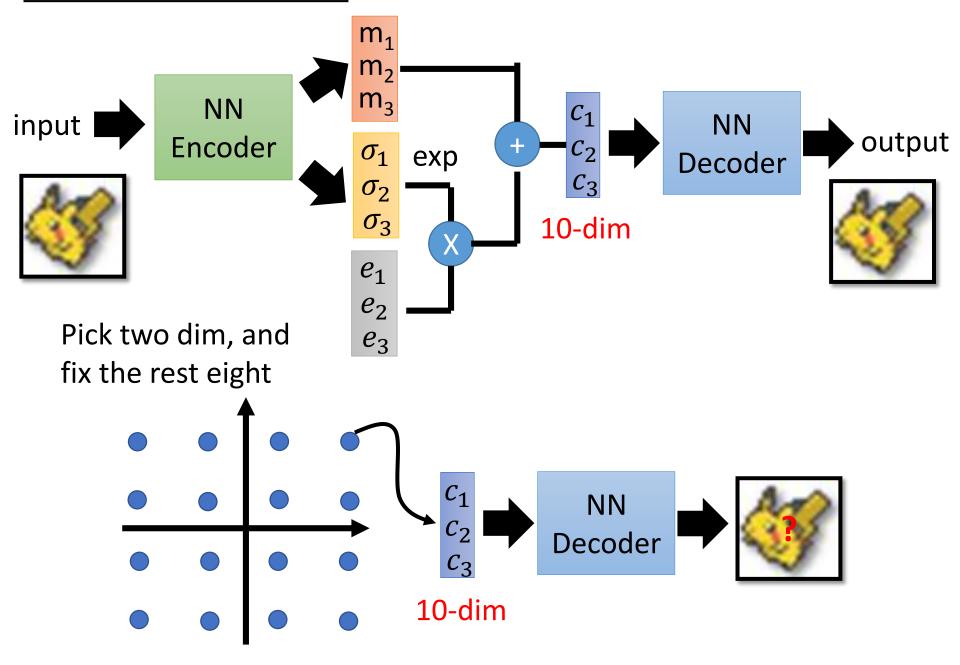


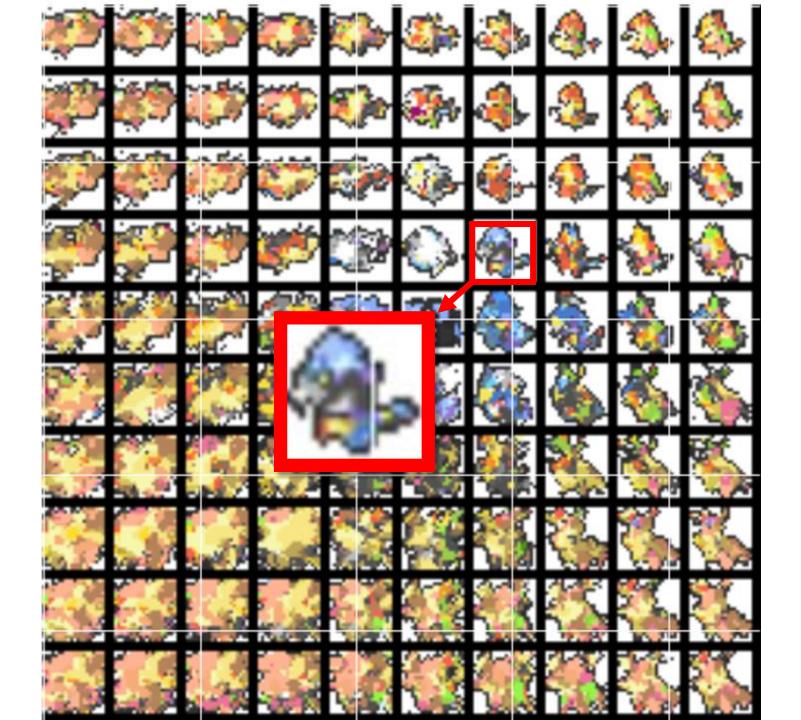
Auto-encoder

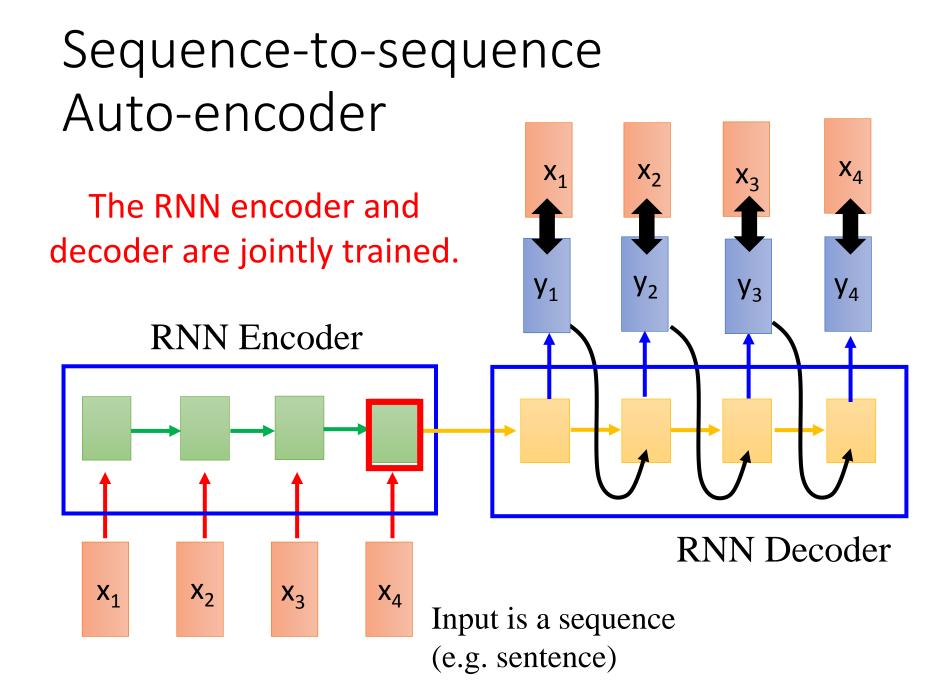


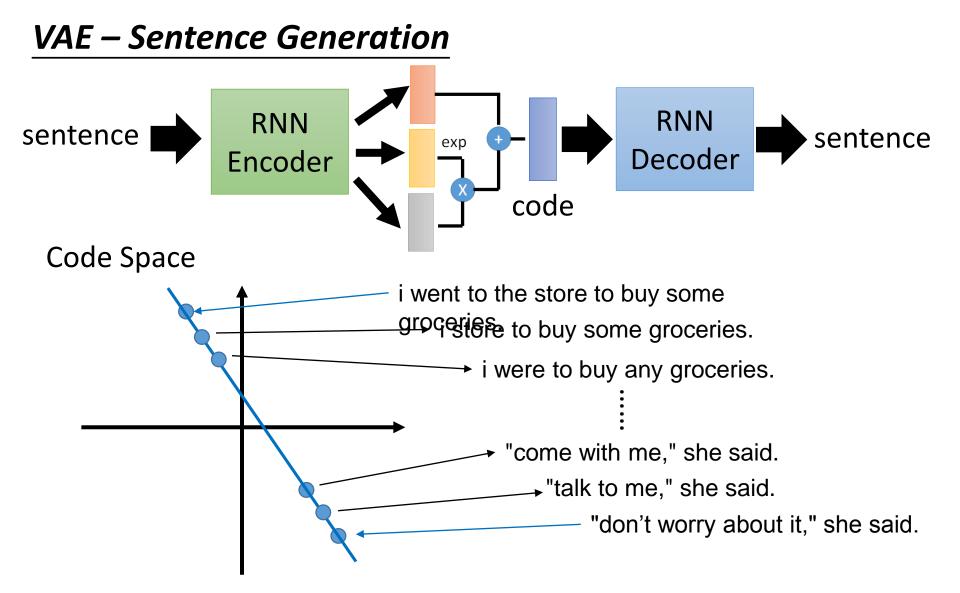


Pokémon Creation









Ref: <u>http://www.wired.co.uk/article/google-artificial-intelligence-poetry</u> Samuel R. Bowman, Luke Vilnis, Oriol Vinyals, Andrew M. Dai, Rafal Jozefowicz, Samy Bengio, Generating Sentences from a Continuous Space, arXiv prepring, 2015

To learn more ...

- Carl Doersch, Tutorial on Variational Autoencoders
- Diederik P. Kingma, Danilo J. Rezende, Shakir Mohamed, Max Welling, "Semi-supervised learning with deep generative models." *NIPS*, 2014.
- Sohn, Kihyuk, Honglak Lee, and Xinchen Yan, "Learning Structured Output Representation using Deep Conditional Generative Models." *NIPS*, 2015.
- Xinchen Yan, Jimei Yang, Kihyuk Sohn, Honglak Lee, "Attribute2Image: Conditional Image Generation from Visual Attributes", ECCV, 2016
- Cool demo:
 - http://vdumoulin.github.io/morphing_faces/
 - http://fvae.ail.tokyo/

Deep Generative Models

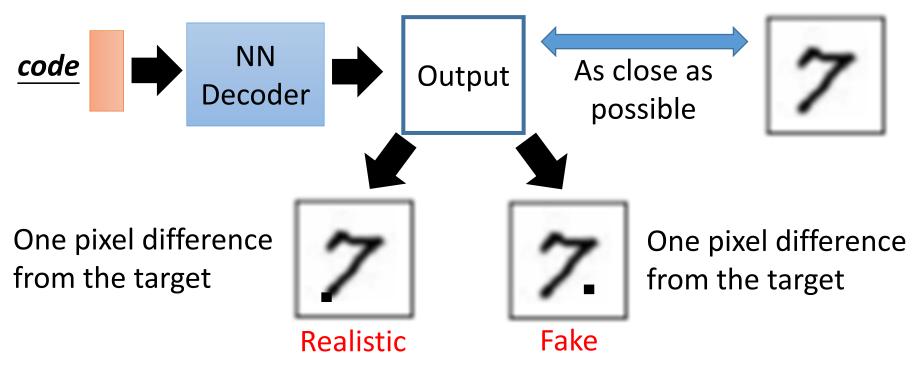
Component-by-component

Auto-encoder

Adversarial Generative Network (GAN)

Problems of VAE

• It does not really try to simulate real images



VAE may just memorize the existing images, instead of generating new images

Yann LeCun's comment

What are some recent and potentially upcoming breakthroughs in deep learning?



Yann LeCun, Director of Al Research at Facebook and Professor at NYU Written Jul 29 · Upvoted by Joaquin Quiñonero Candela, <u>Director Applied Machine</u> Learning at Facebook and Nikhil Garg, I lead a team of Quora engineers working on ML/NLP problems



• • • • • •

The most important one, in my opinion, is adversarial training (also called GAN for Generative Adversarial Networks). This is an idea that was originally proposed by Ian Goodfellow when he was a student with Yoshua Bengio at the University of Montreal (he since moved to Google Brain and recently to OpenAI).

This, and the variations that are now being proposed is the most interesting idea in the last 10 years in ML, in my opinion.

https://www.quora.com/What-are-some-recent-and-potentially-upcoming-breakthroughsin-deep-learning

Yann LeCun's comment

What are some recent and potentially upcoming breakthroughs in unsupervised learning?



Yann LeCun, Director of Al Research at Facebook and Professor at NYU Written Jul 29 · Upvoted by Joaquin Quiñonero Candela, Director Applied Machine Learning at Facebook and Huang Xiao



Adversarial training is the coolest thing since sliced bread.

I've listed a bunch of relevant papers in a previous answer.

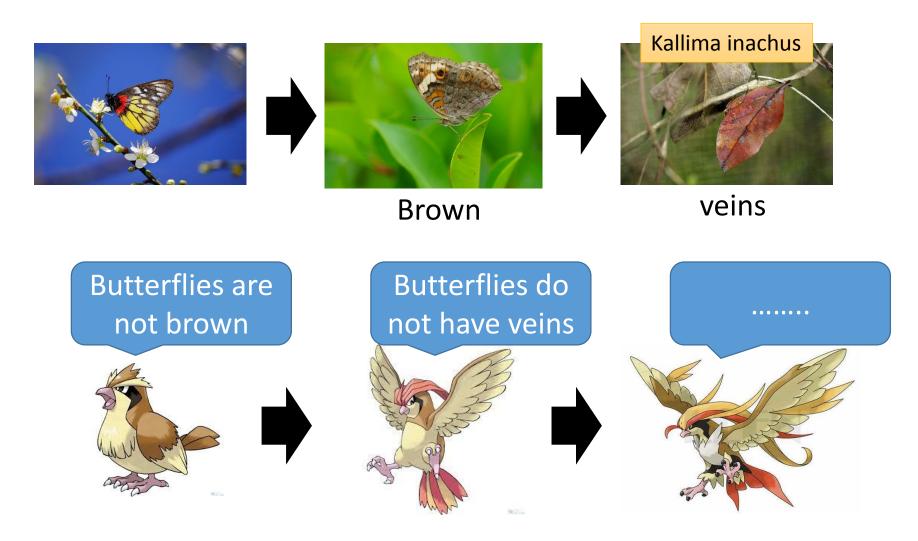
Expect more impressive results with this technique in the coming years.

What's missing at the moment is a good understanding of it so we can make it work reliably. It's very finicky. Sort of like ConvNet were in the 1990s, when I had the reputation of being the only person who could make them work (which wasn't true).

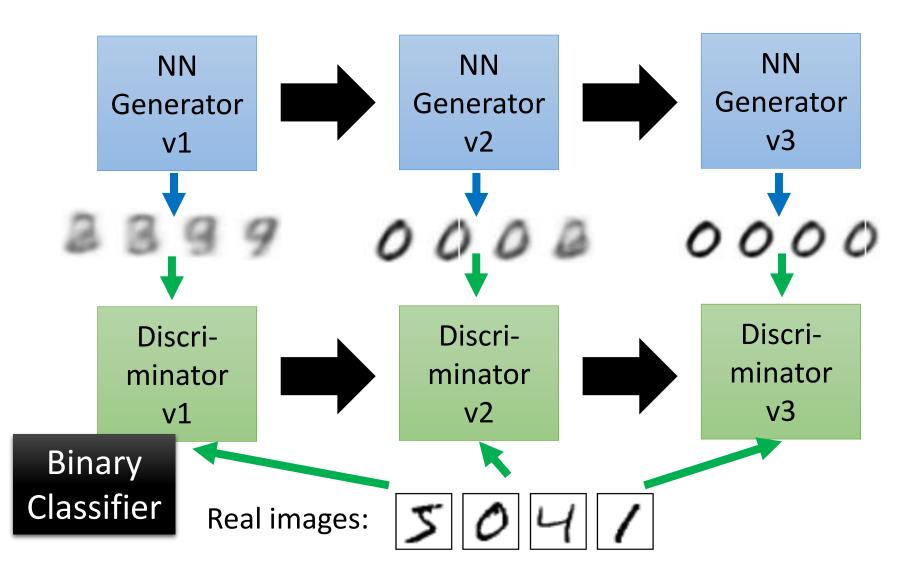
https://www.quora.com/What-are-some-recent-andpotentially-upcoming-breakthroughs-in-unsupervised-learning

Evolution

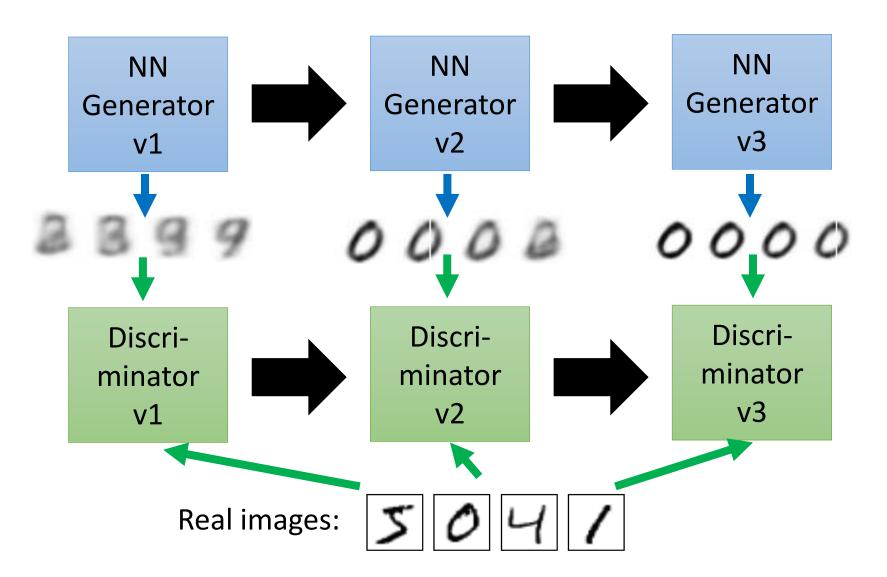
http://peellden.pixnet.net/blog/post/40406899-2013-%E7%AC%AC%E5%9B%9B%E5%AD%A3%EF%BC%8C %E5%86%AC%E8%9D%B6%E5%AF%82%E5%AF%A5



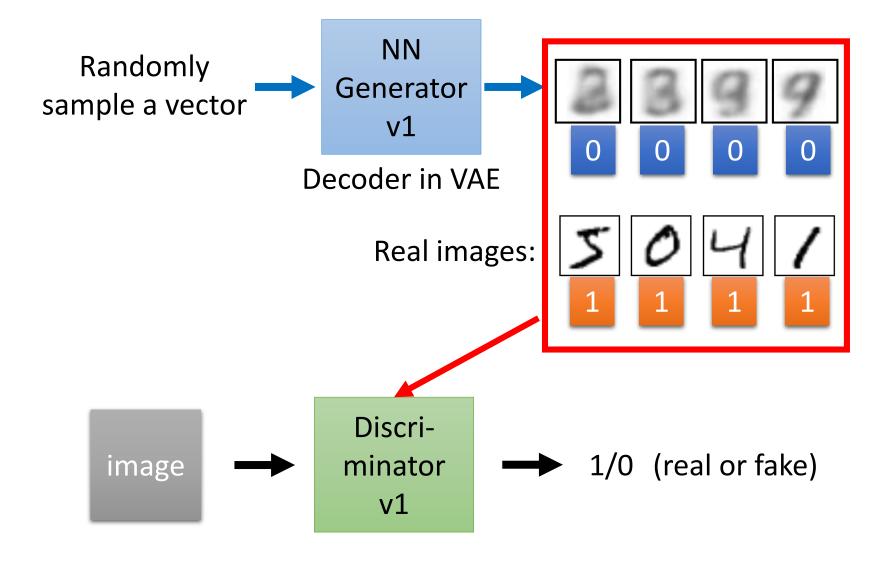
The evolution of generation



The evolution of generation



GAN - Discriminator



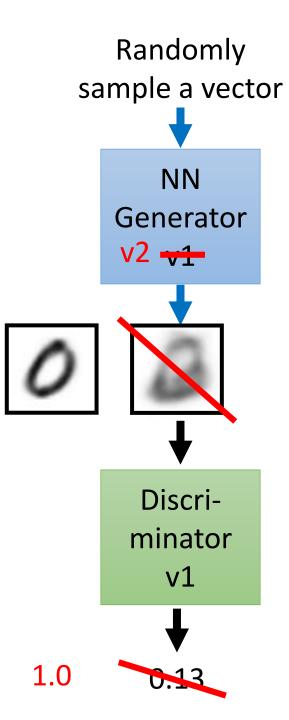
GAN - Generator

Updating the parameters of generator

The output be classified as "real" (as close to 1 as possible)

Generator + Discriminator = a network

Using gradient descent to update the parameters in the generator, but fix the discriminator



Cifar-10

• Which one is machine-generated?



Ref: https://openai.com/blog/generative-models/

Cartoon

• Ref: https://github.com/mattya/chainer-DCGAN



Cartoon

Ref: http://qiita.com/mattya/items/e5bfe5e04b9d2f0bbd47



元画像 -赤髪+金髪 -赤目+青目 +制服+セーラー +笑顔+口開き +青背景



ー番左のキャラクターが元画像で、 右に行くほど長髪化ベクトルを強く足している

To learn more ...

- "Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks"
- "Improved Techniques for Training GANs"
- "Autoencoding beyond pixels using a learned similarity metric"
- "Deep Generative Image Models using a Laplacian Pyramid of Adversarial Network"
- "Super Resolution using GANs"
- "Generative Adversarial Text to Image Synthesis"

Outline

Deep Generative Models

 Richard Feynman: "What I cannot create, I do not understand."

Deep Reinforcement Learning

• The technique makes Alpha Go better than professional players.

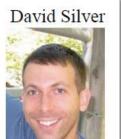
Deep Reinforcement Learning



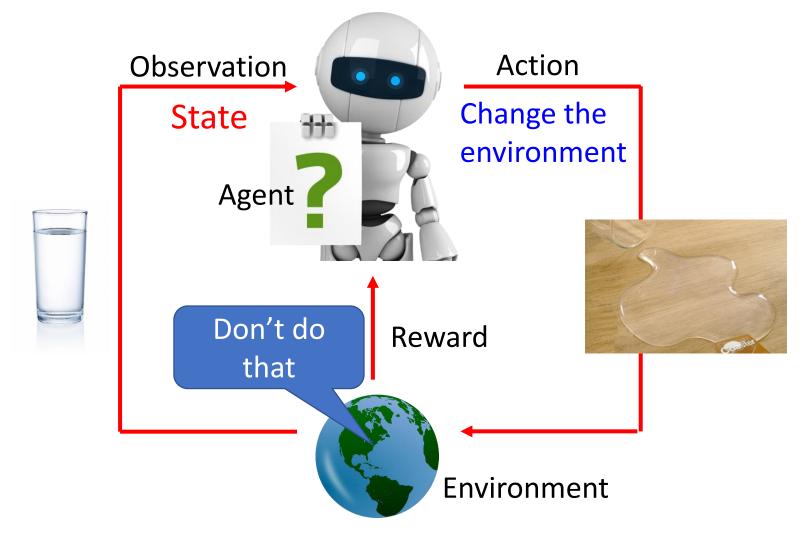
At last – a computer program that can beat a champion Go player PAGE 484

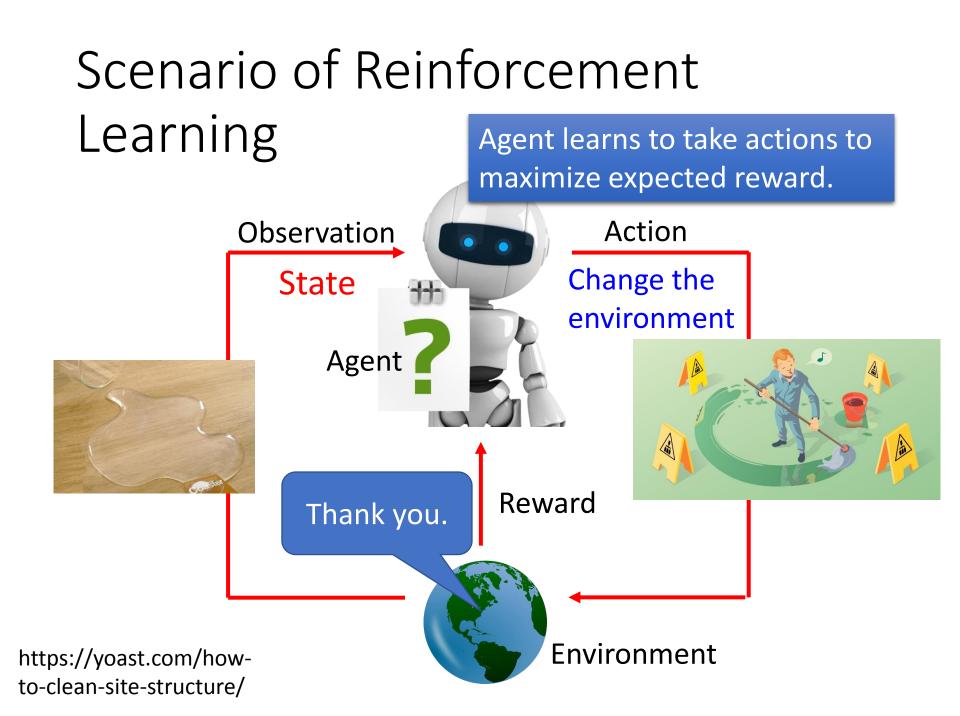
ALL SYSTEMS

Deep Reinforcement Learning: AI = RL + DL

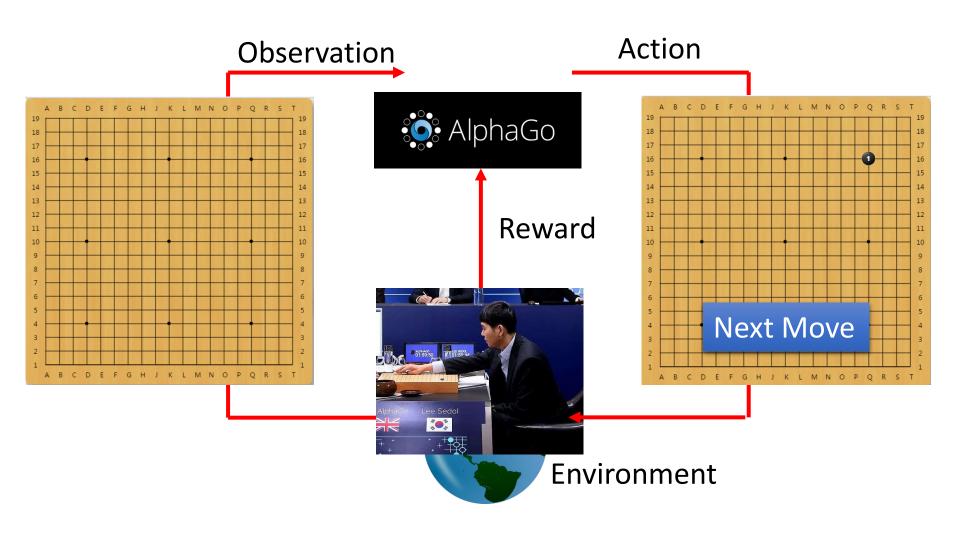


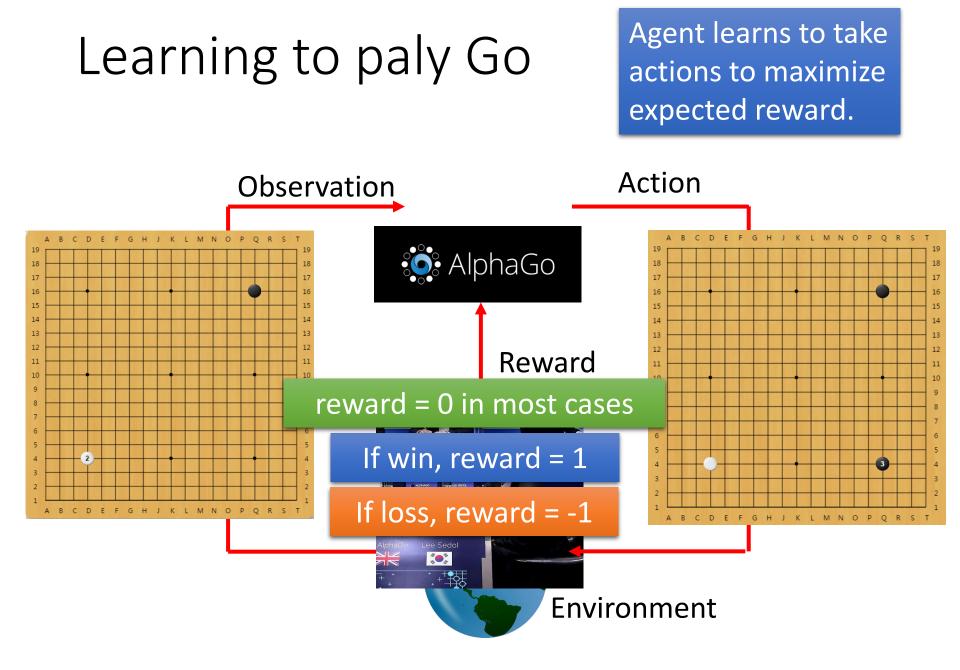
Scenario of Reinforcement Learning





Learning to paly Go





Learning to paly Go - Supervised v.s. Reinforcement

• Supervised: Learning from teacher

Cannot be better than its teacher



Next move: "5-5"



Next move: "3-3"

• Reinforcement Learning

Learning from experience

First move 📄 many moves



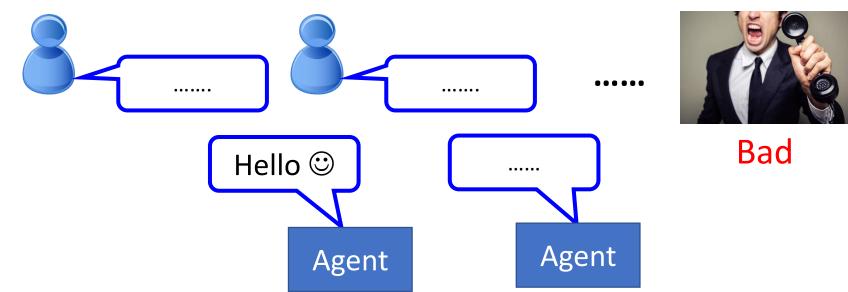
(Two agents play with each other.)

Alpha Go is supervised learning + reinforcement learning.

Learning a chat-bot - Supervised v.s. Reinforcement

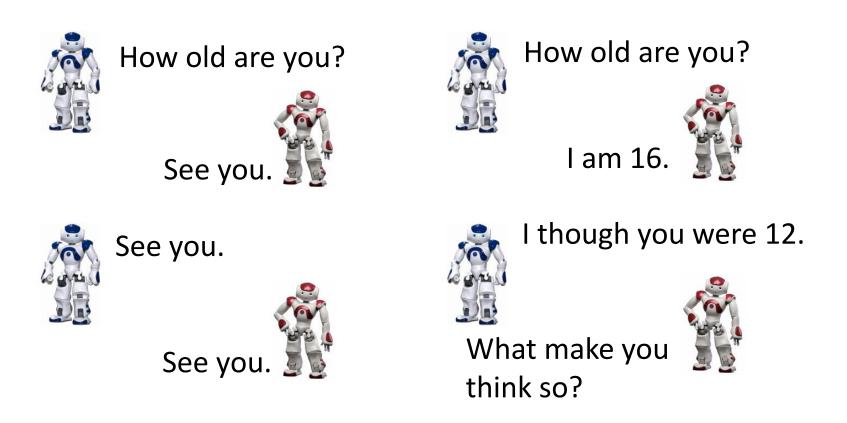
Supervised

- "Hello" Say "Hi" "Bye bye" Say "Good bye"
- Reinforcement



Learning a chat-bot - Reinforcement Learning

 Let two agents talk to each other (sometimes generate good dialogue, sometimes bad)



Learning a chat-bot - Reinforcement Learning

- By this approach, we can generate a lot of dialogues.
- Use some pre-defined rules to evaluate the goodness of a dialogue

Machine learns from the evaluation

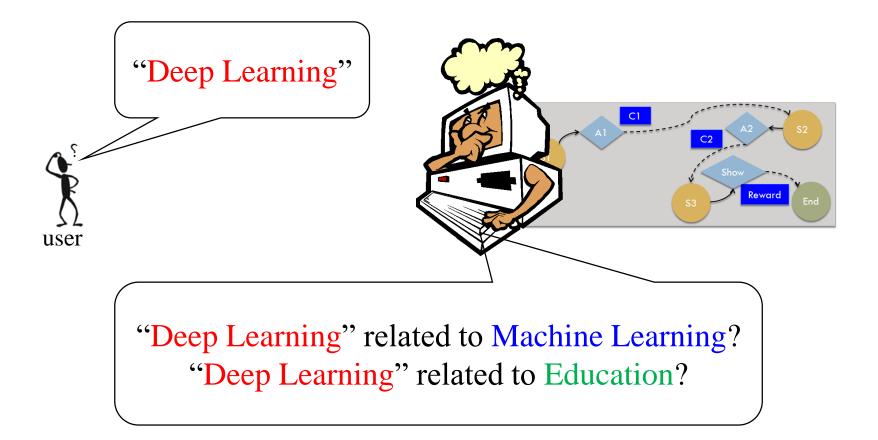


Deep Reinforcement Learning for Dialogue Generation https://arxiv.org/pdf/1606.01541v3.pdf

Application: Interactive Retrieval

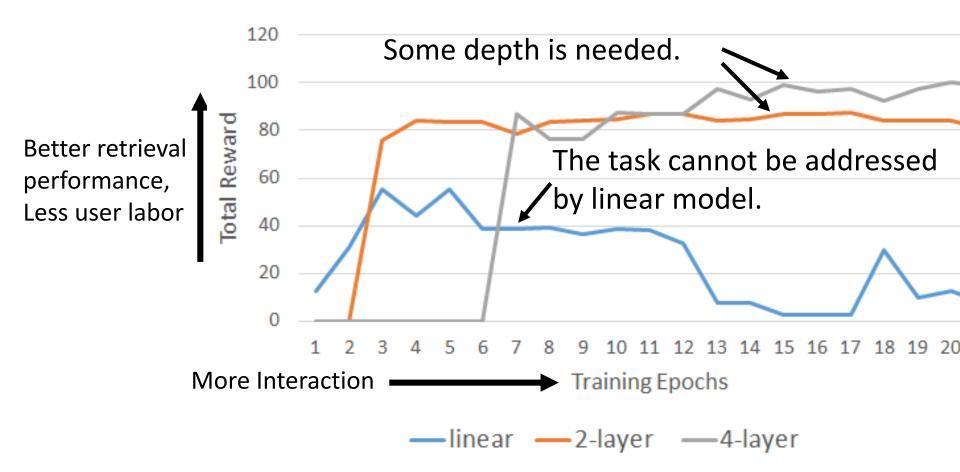
• Interactive retrieval is helpful. [Wu 8

[Wu & Lee, INTERSPEECH 16]



Deep Reinforcement Learning

Different network depth



More applications

- Flying Helicopter
 - https://www.youtube.com/watch?v=0JL04JJjocc
- Driving
 - https://www.youtube.com/watch?v=0xo1Ldx3L5Q
- Google Cuts Its Giant Electricity Bill With DeepMind-Powered AI
 - http://www.bloomberg.com/news/articles/2016-07-19/google-cuts-itsgiant-electricity-bill-with-deepmind-powered-ai
- Text generation
 - Hongyu Guo, "Generating Text with Deep Reinforcement Learning", NIPS, 2015
 - Marc'Aurelio Ranzato, Sumit Chopra, Michael Auli, Wojciech Zaremba, "Sequence Level Training with Recurrent Neural Networks", ICLR, 2016

To learn deep reinforcement learning

- Textbook: Reinforcement Learning: An Introduction
 - https://webdocs.cs.ualberta.ca/~sutton/book/thebook.html
- Lectures of David Silver
 - http://www0.cs.ucl.ac.uk/staff/D.Silver/web/Teaching.ht ml (10 lectures, 1:30 each)
 - http://videolectures.net/rldm2015_silver_reinforcement
 t_learning/ (Deep Reinforcement Learning)
- Lectures of John Schulman
 - https://youtu.be/aUrX-rP_ss4