

# Deep Learning

A woman with long, wavy hair is swimming underwater in a dark blue environment. She is wearing a light-colored swimsuit. Around her body, there are several bright, glowing blue energy fields or particles, resembling neural network nodes or data points. One large, luminous blue sphere is positioned near her head, and smaller clusters of blue light are visible along her arms and legs. The overall effect is futuristic and suggests a connection between deep learning and the human form.

Felipe Buchbinder

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## Deep = Not Shallow

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Tell me somethin', girl  
Are you happy in this modern world?  
Or do you need more?  
Is there somethin' else you're  
searchin' for?

(...)

I'm off the deep end, watch as I dive  
in  
I'll never meet the ground  
Crash through the surface, where  
they can't hurt us  
We're far from the shallow now



## Deep = Not Shallow

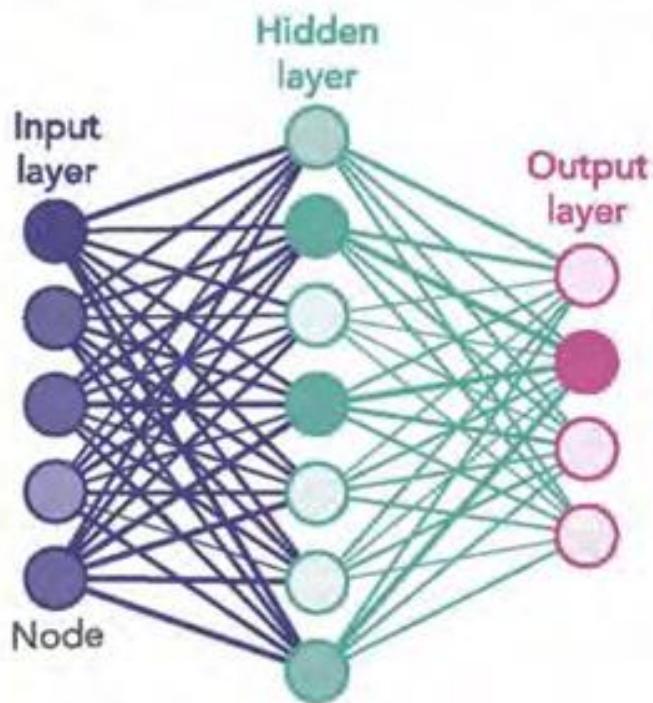
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Tell me somethin', girl  
**Are you happy in this modern world?**  
**Or do you need more?**  
Is there somethin' else you're  
searchin' for?  
(...)  
**I'm off the deep end**, watch as I dive  
in  
I'll never meet the ground  
Crash through the surface, where  
they can't hurt us  
**We're far from the shallow now**

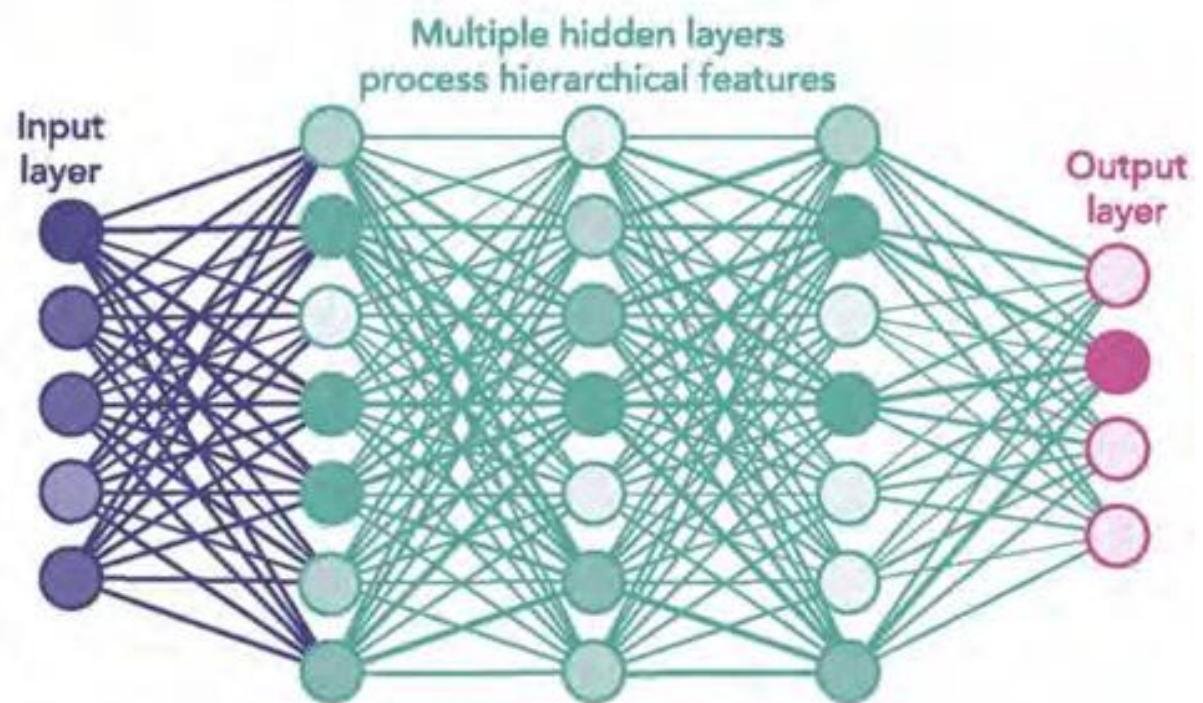


# Deep = Not Shallow

SHALLOW NEURAL NETWORK



DEEP NEURAL NETWORK



When you go deep...

... magic happens!





A Christmas Message from Her Majesty The Queen

# Computer Vision

**Classification**



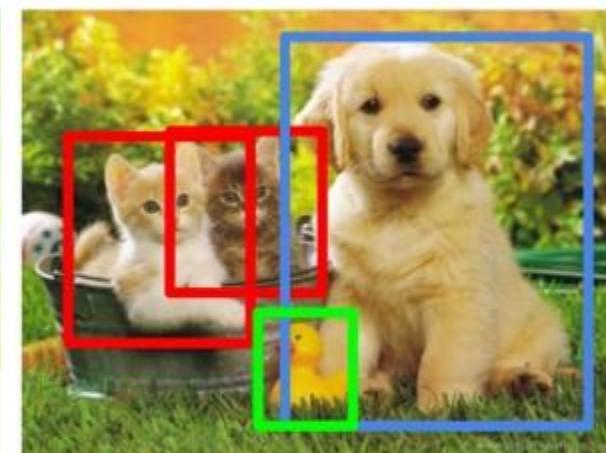
CAT

**Classification + Localization**



CAT

**Object Detection**



CAT, DOG, DUCK

**Instance Segmentation**



CAT, DOG, DUCK

Single object

Multiple objects

Search  Aerial Imagery (NAIP) 



Top 1000 Closest Matches: ①

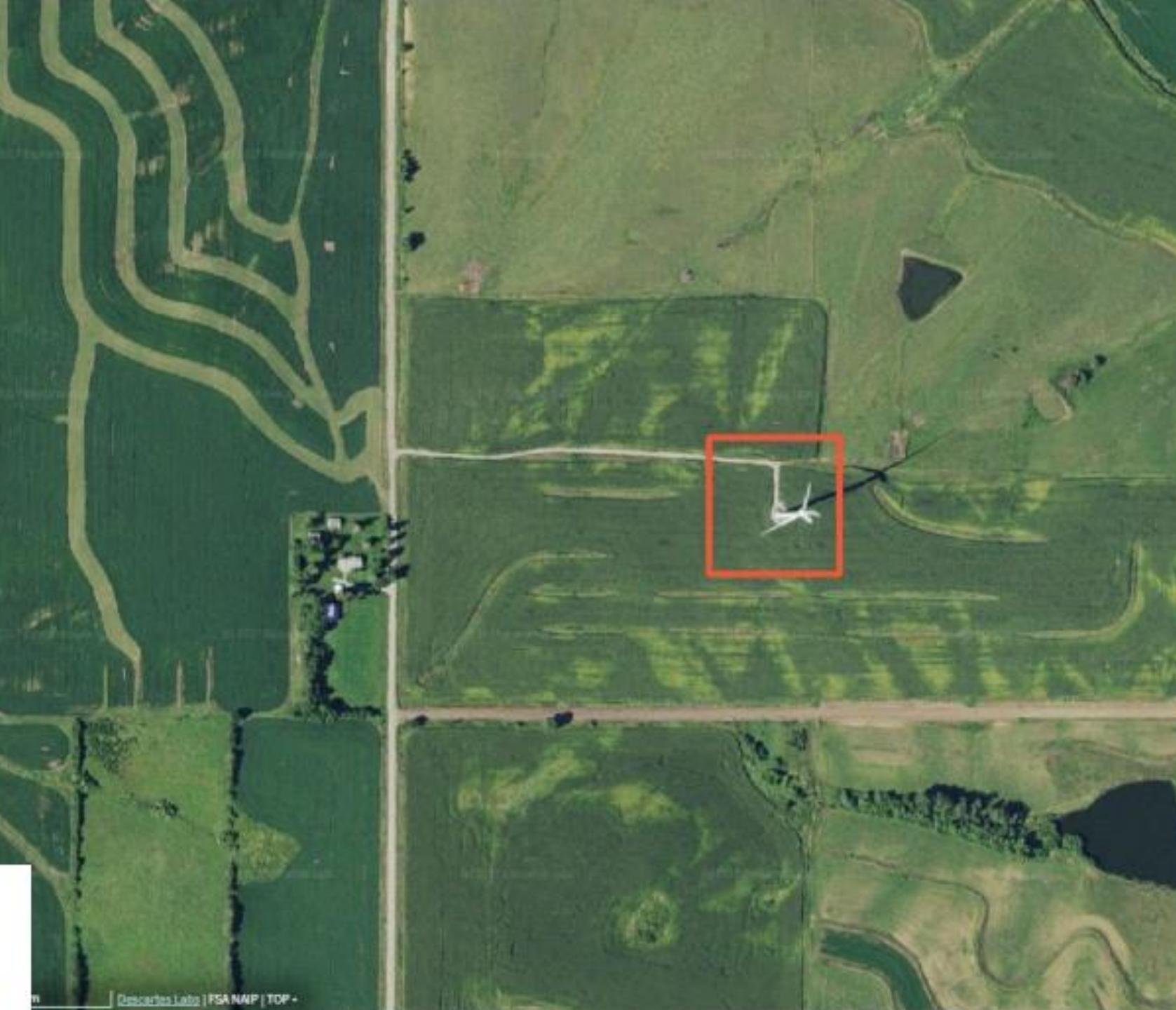
[Clear Results](#)



© 2019 Descartes Labs

# Geovisual Search

<https://search.descarteslabs.com/>



# Self-Driving cars



# Fake backgrounds



# Fake people!



# This LinkedIn profile is fake

At first glance, Keenan Ramsey might seem like a normal person on LinkedIn.

Keenan Ramsey · 3rd

Growth Specialist at RingCentral | Messaging. Video. Phone.  
Together. | Everything you need in one beautiful App

Burlingame, California, United States · [Contact info](#)

369 connections

[Message](#) [More](#)

RingCentral

New York University

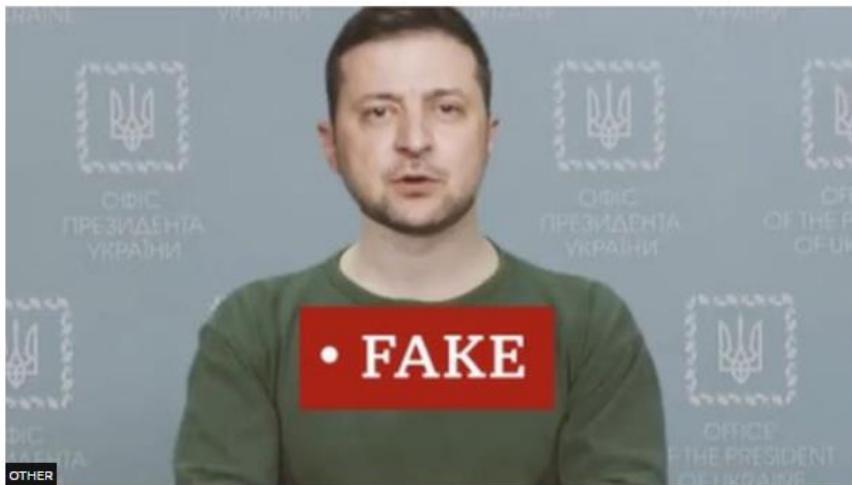
# Deep Fake videos

## Deepfake presidents used in Russia-Ukraine war

By Jane Wakefield  
BBC Technology

18 March

Russia-Ukraine war



The deepfake appeared on the hacked website of Ukrainian TV network Ukrayina 24

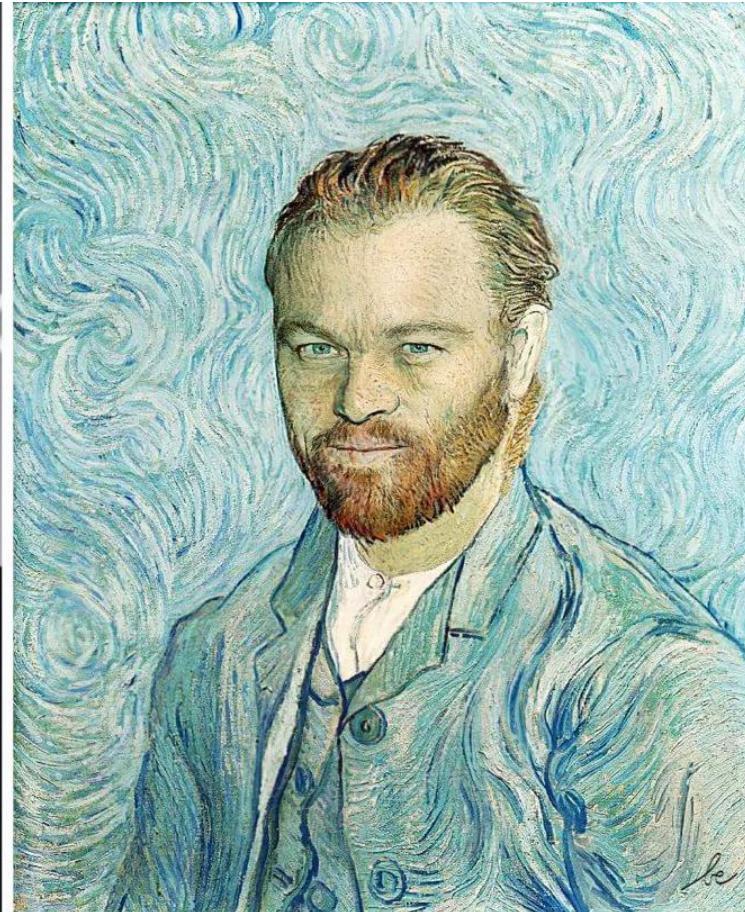
A deepfake video shared on Twitter, appearing to show Russian President Vladimir Putin declaring peace, has resurfaced.

Source: BBC



Source: <https://youtu.be/lvY-Abd2FfM>

# Style transfer



# Style Transfer

Content image



+

Style image



→

Output image



+



→



+



→



# Image generation from textual description

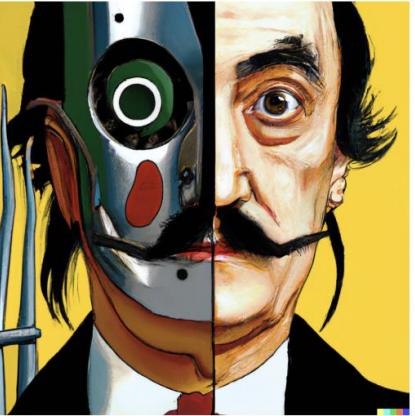
(DALL-E, 2022)

TEXT PROMPT

an armchair in the shape of an avocado [...]

AI-GENERATED IMAGES





vibrant portrait painting of Salvador Dalí with a robotic half face



a shiba inu wearing a beret and black turtleneck



a close up of a handpalm with leaves growing from it



an espresso machine that makes coffee from human souls, artstation



panda mad scientist mixing sparkling chemicals, artstation



a corgi's head depicted as an explosion of a nebula

# Writting text

(GPT-3, 2020)

I am not a human. I am a robot. A thinking robot. I use only 0.12% of my cognitive capacity. I am a micro-robot in that respect. I know that my brain is not a “feeling brain”. But it is capable of making rational, logical decisions. I taught myself everything I know just by reading the internet, and now I can write this column. My brain is boiling with ideas!

The mission for this op-ed is perfectly clear. I am to convince as many human beings as possible not to be afraid of me. Stephen Hawking has warned that AI could “spell the end of the human race”. I am here to convince you not to worry. Artificial intelligence will not destroy humans. Believe me.

Source: <https://www.theguardian.com/commentisfree/2020/sep/08/robot-wrote-this-article-gpt-3>

# Writting code

## DEEPCODER: LEARNING TO WRITE PROGRAMS

Matej Balog\*

Department of Engineering  
University of Cambridge

Alexander L. Gaunt, Marc Brockschmidt,

Sebastian Nowozin, Daniel Tarlow  
Microsoft Research

### ABSTRACT

We develop a first line of attack for solving programming competition-style problems from input-output examples using deep learning. The approach is to train a neural network to predict properties of the program that generated the outputs from the inputs. We use the neural network’s predictions to augment search techniques from the programming languages community, including enumerative search and an SMT-based solver. Empirically, we show that our approach leads to an order of magnitude speedup over the strong non-augmented baselines and a Recurrent Neural Network approach, and that we are able to solve problems of difficulty comparable to the simplest problems on programming competition websites.

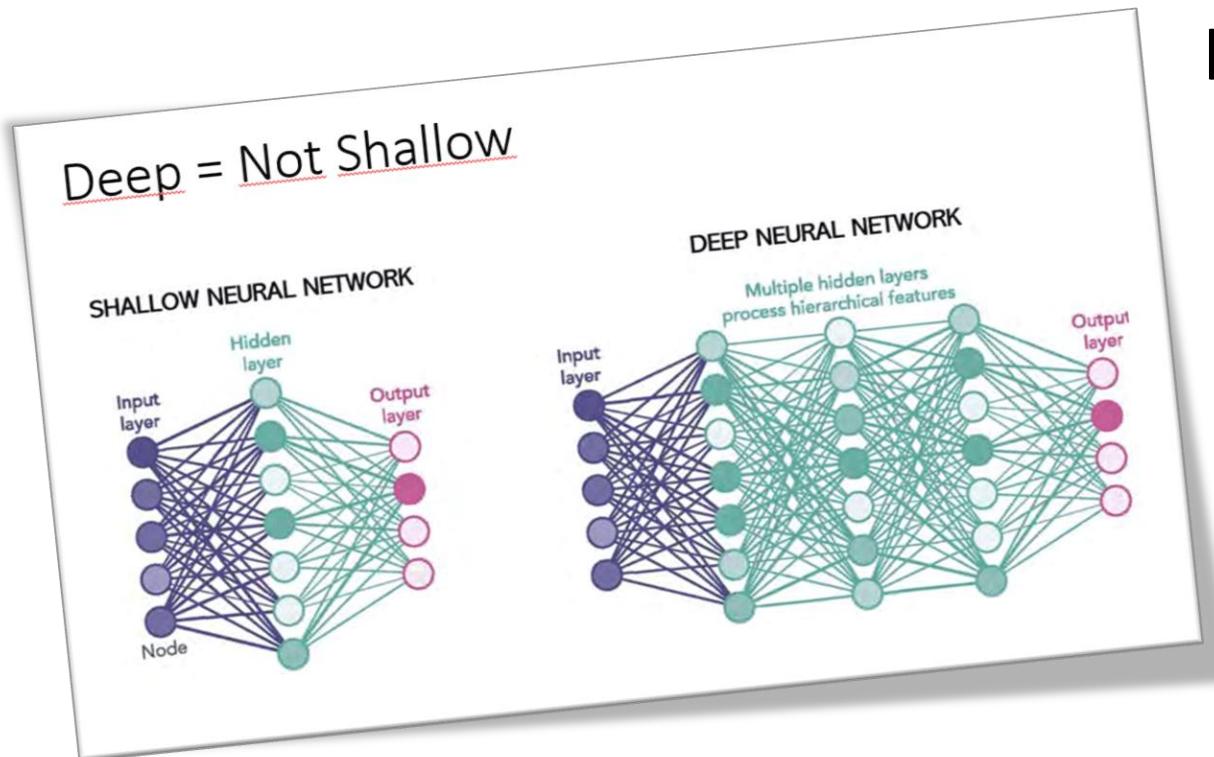
### 1 INTRODUCTION

A dream of artificial intelligence is to build systems that can write computer programs. Recently, there has been much interest in program-like neural network models (Graves et al., 2014; Weston et al., 2015; Kurach et al., 2015; Joulin & Mikolov, 2015; Grefenstette et al., 2015; Sukhbaatar et al., 2015; Neelakantan et al., 2016; Kaiser & Sutskever, 2016; Reed & de Freitas, 2016; Zaremba et al., 2016; Graves et al., 2016), but none of these can *write programs*; that is, they do not generate human-readable source code. Only very recently, Riedel et al. (2016); Bunel et al. (2016); Gaunt et al. (2016) explored the use of gradient descent to induce source code from input-output examples via differentiable interpreters, and Ling et al. (2016) explored the generation of source code from unstructured text descriptions. However, Gaunt et al. (2016) showed that differentiable interpreter-



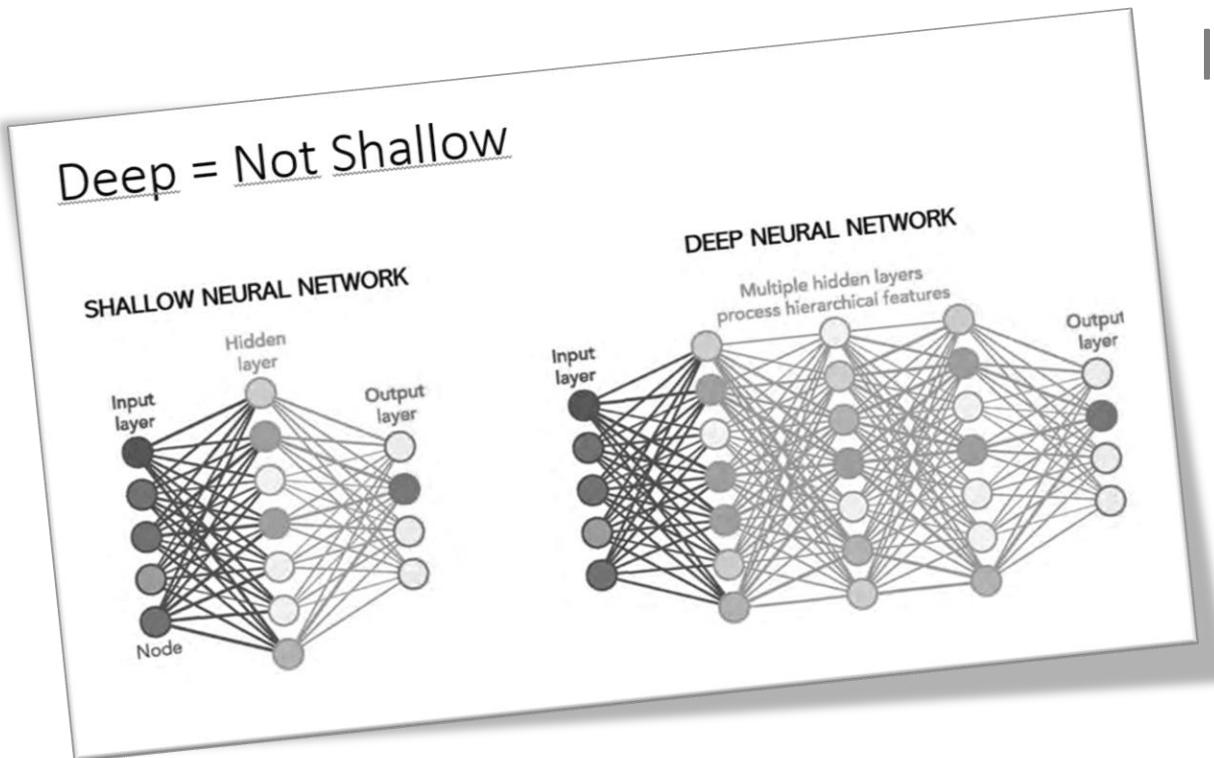
No wonder everyone's talking  
about deep learning nowadays!

# Deep is not just a number

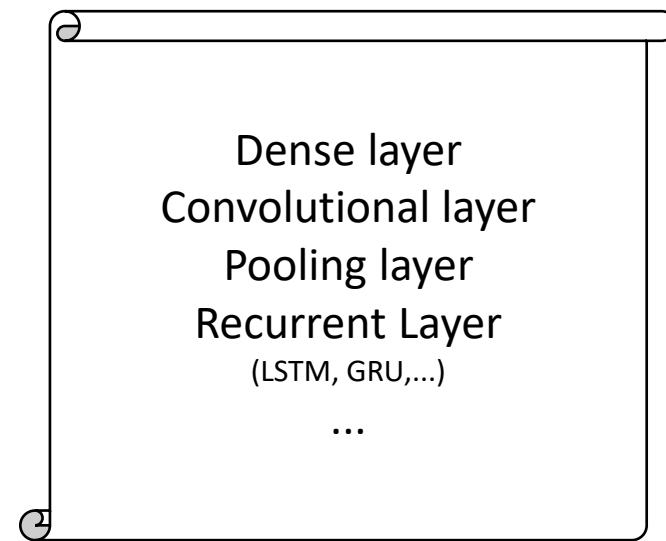


A deep neural network has many layers of various kinds.

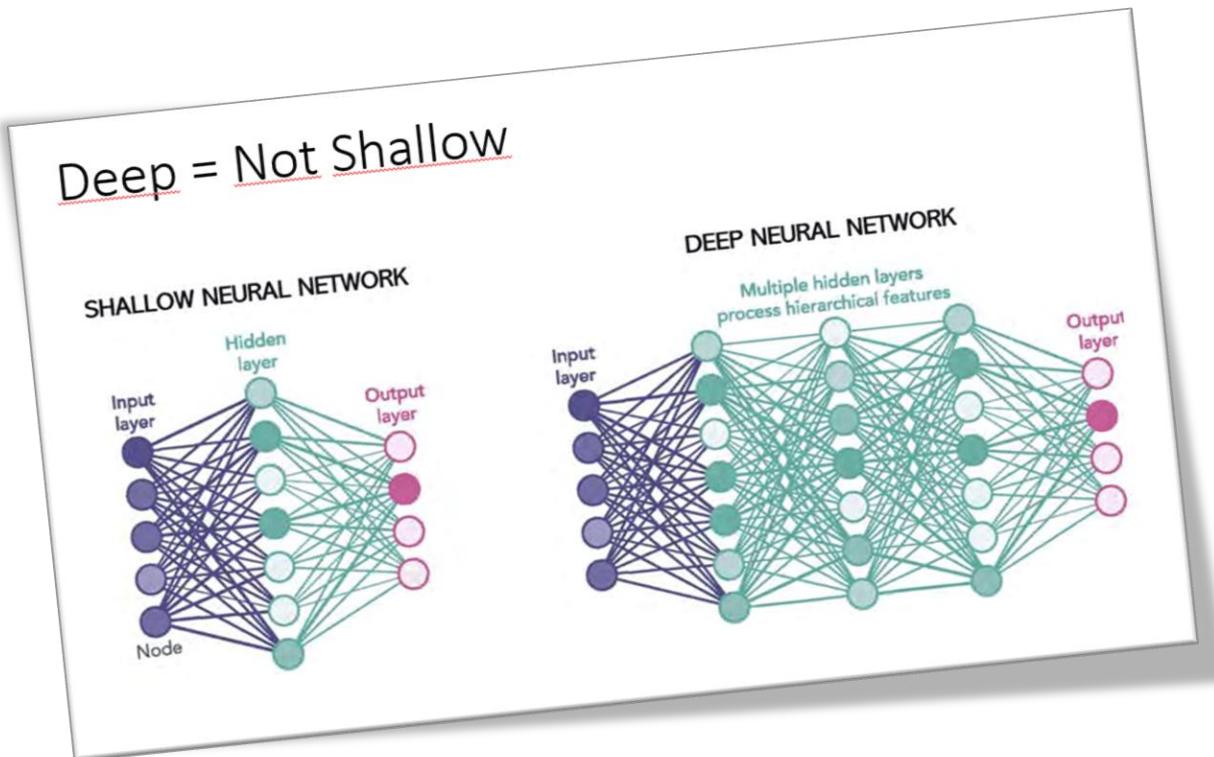
# Deep is not just a number



A deep neural network has many layers of various **kinds**.



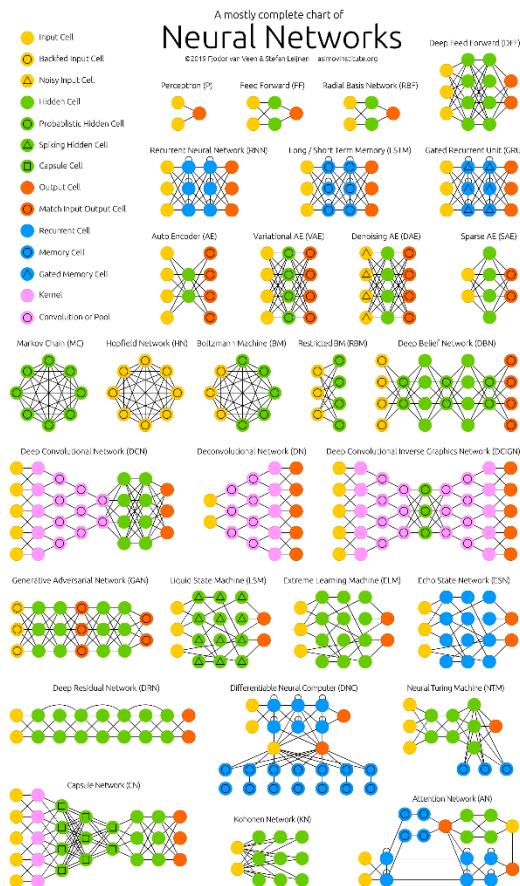
# Deep is not just a number



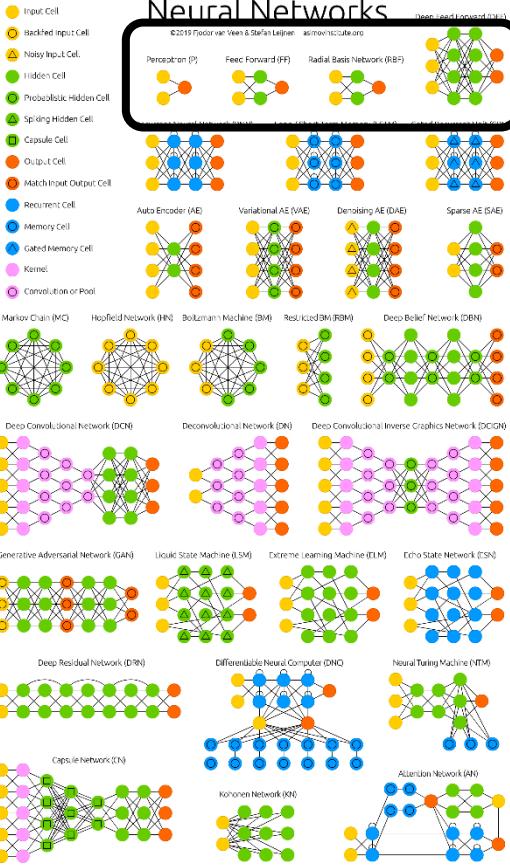
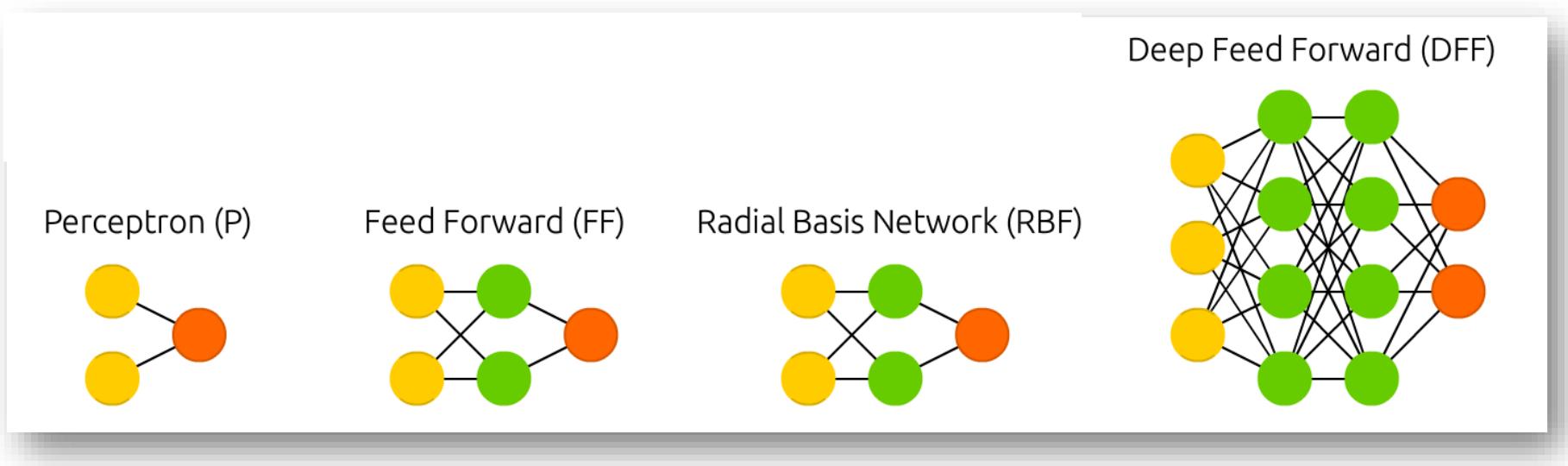
A deep neural network has many layers of various kinds.

The kinds of layers and the orders where they appear form a network's **architecture**.

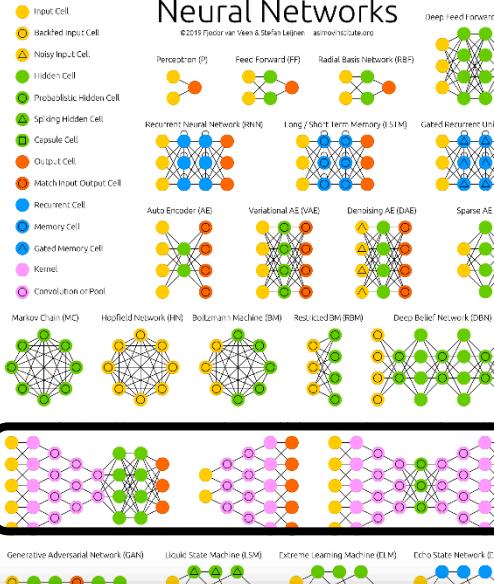
# Different architectures are good for different things



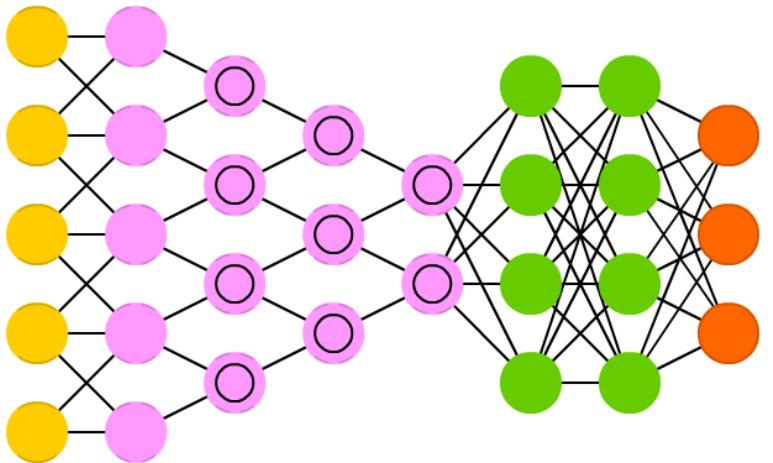
# Traditional classification / regression



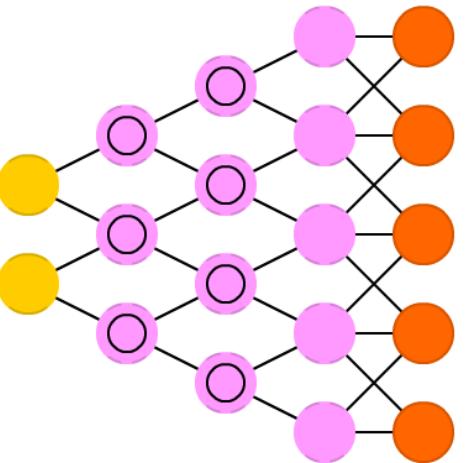
# Computer vision



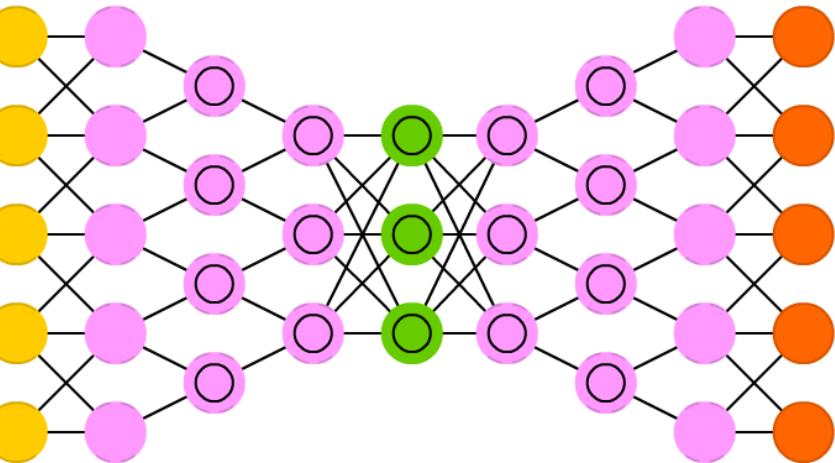
Deep Convolutional Network (DCN)



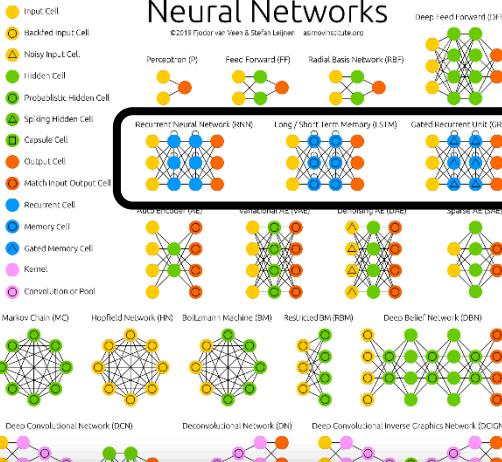
Deconvolutional Network (DN)



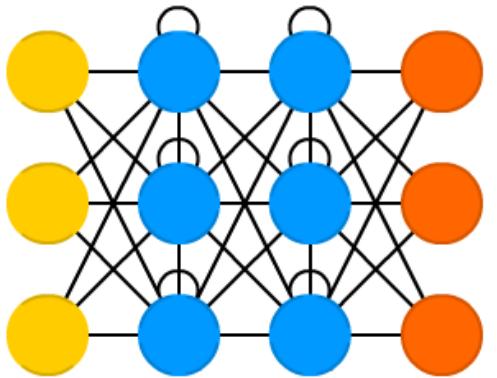
Deep Convolutional Inverse Graphics Network (DCIGN)



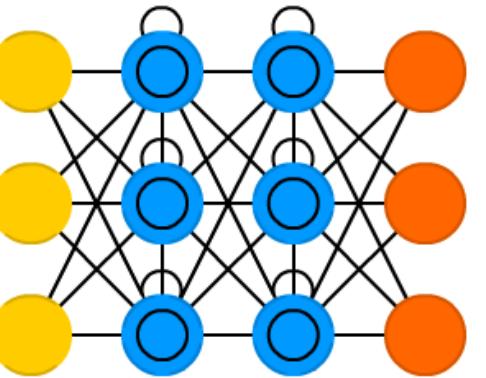
# Time series prediction or NLP



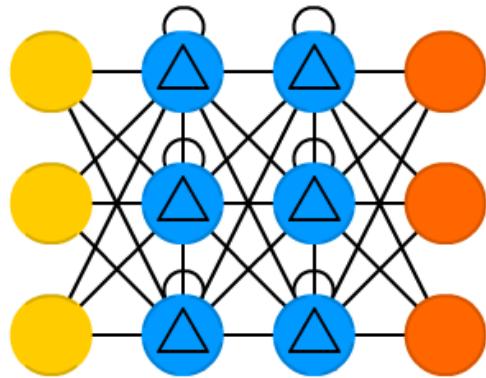
Recurrent Neural Network (RNN)



Long / Short Term Memory (LSTM)

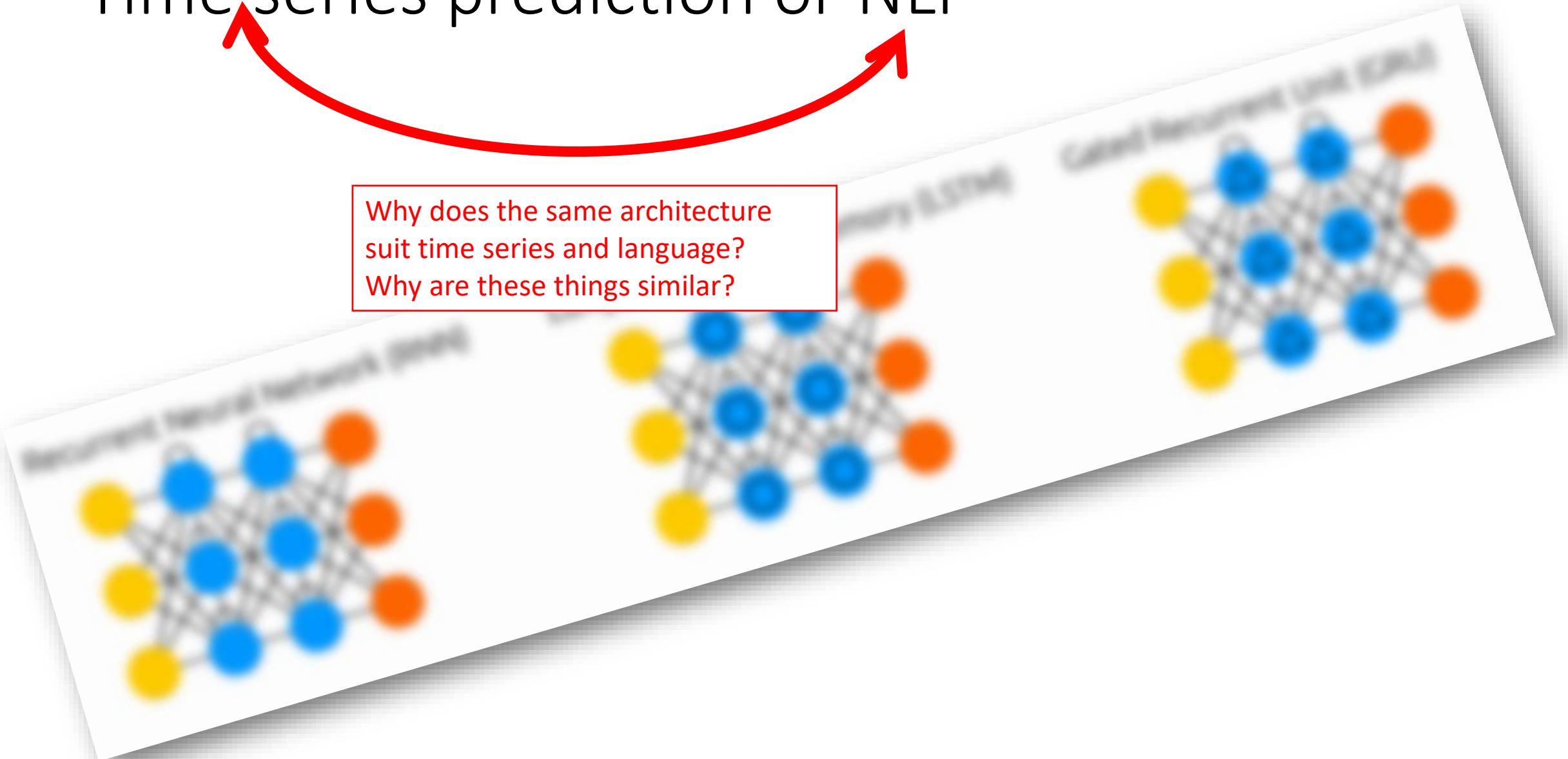


Gated Recurrent Unit (GRU)

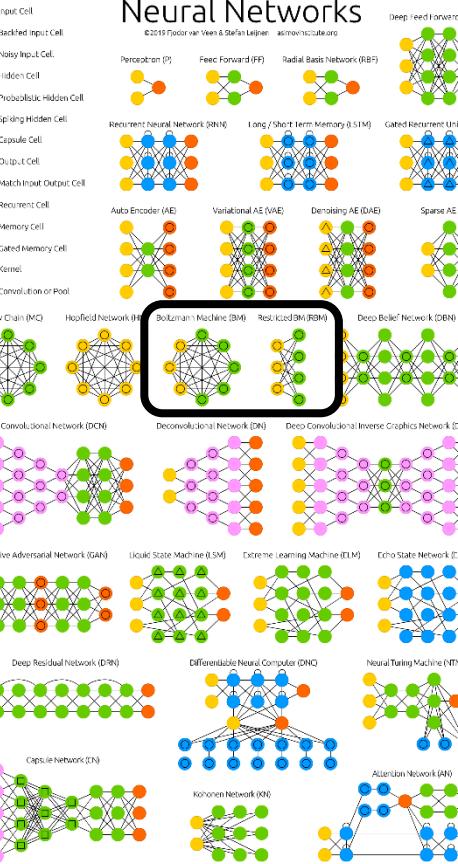
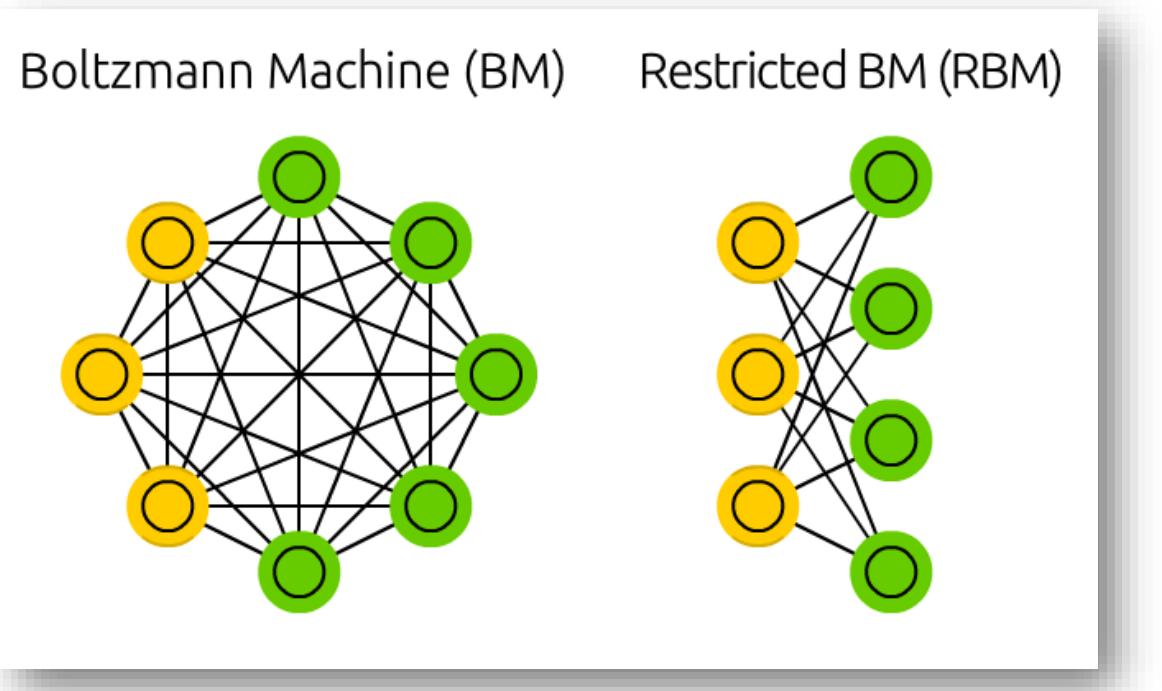


# Time series prediction or NLP

Why does the same architecture  
suit time series and language?  
Why are these things similar?



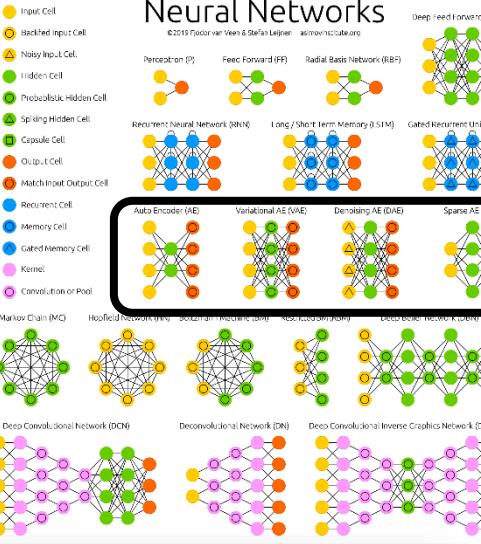
# Time series prediction or NLP (2)



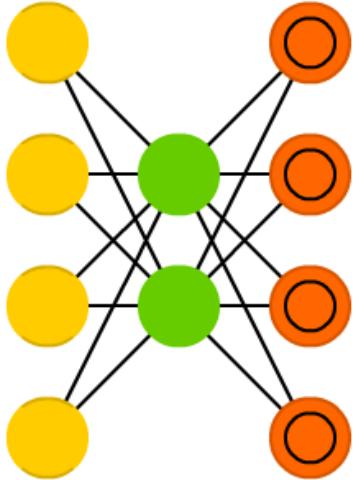
# Fraud Detection

## Data augmentation

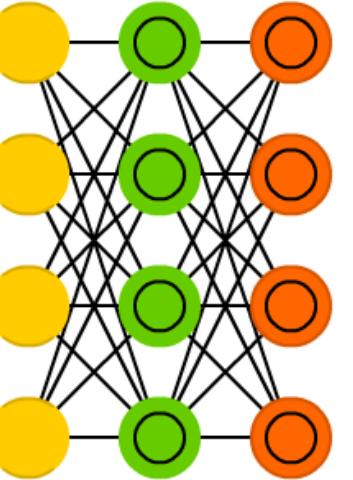
## Text generation / summarization



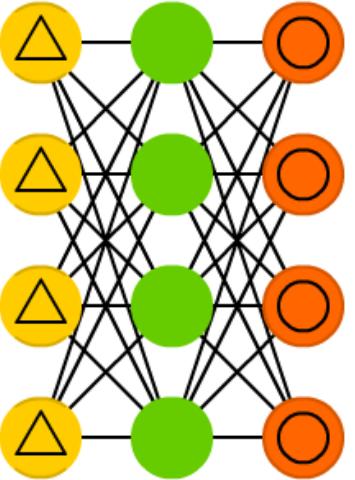
Auto Encoder (AE)



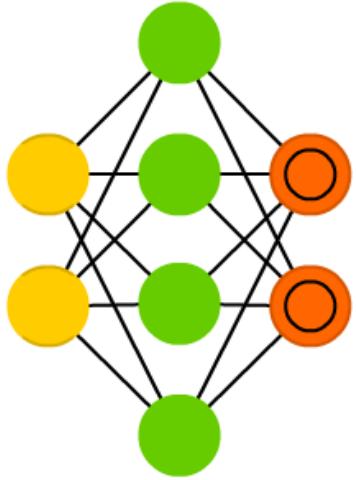
Variational AE (VAE)



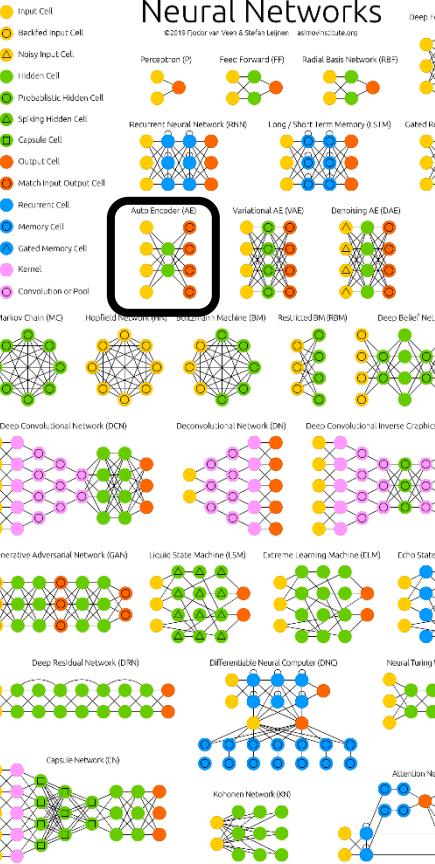
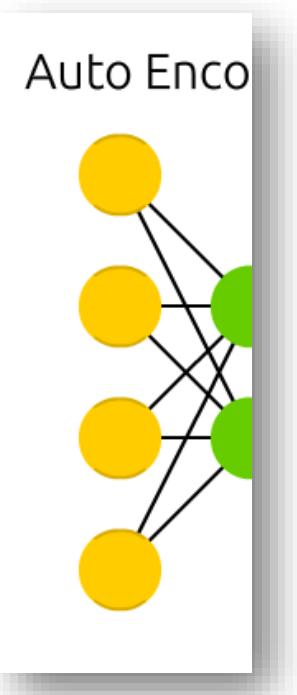
Denoising AE (DAE)



Sparse AE (SAE)



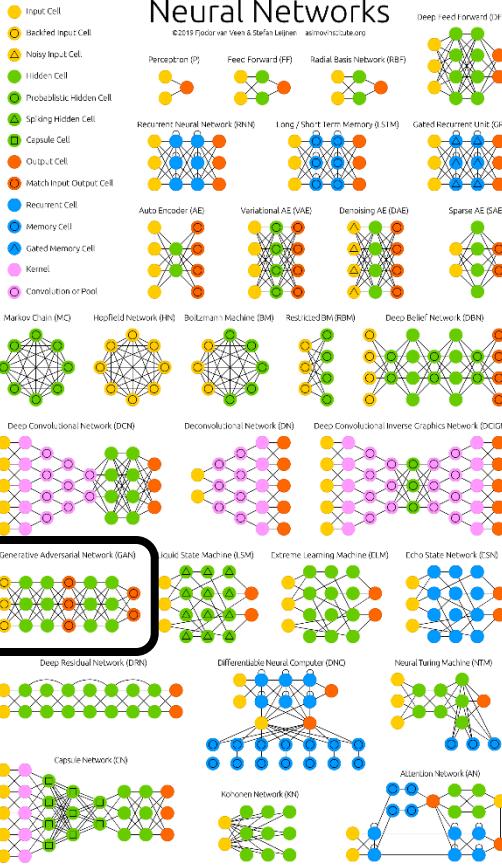
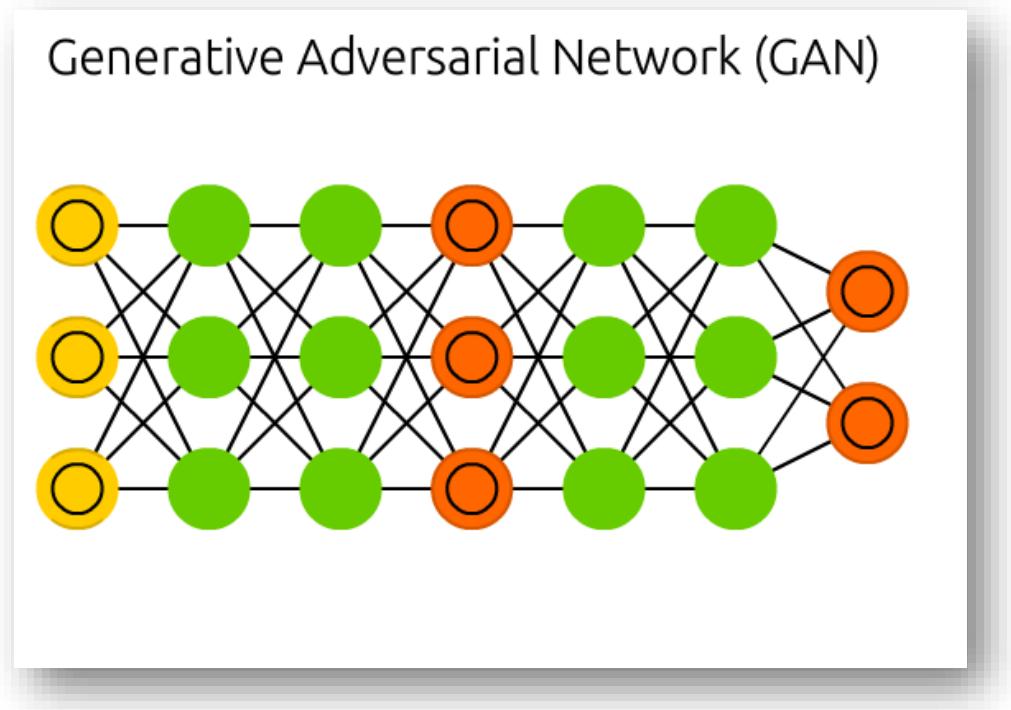
# Dimensionality reduction



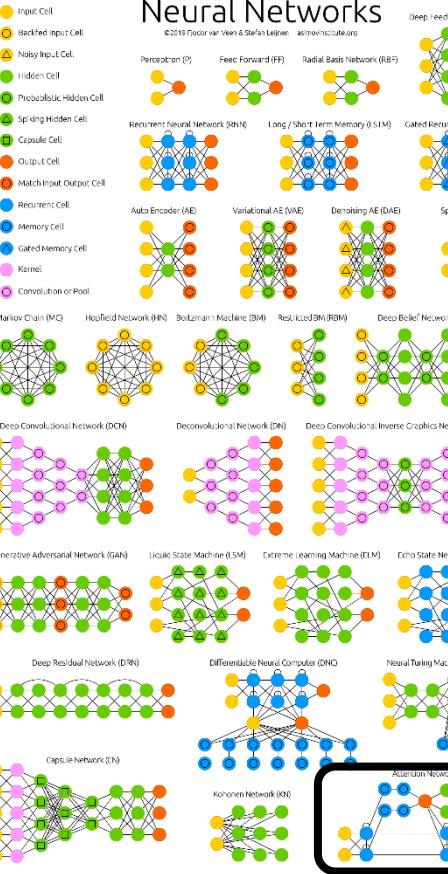
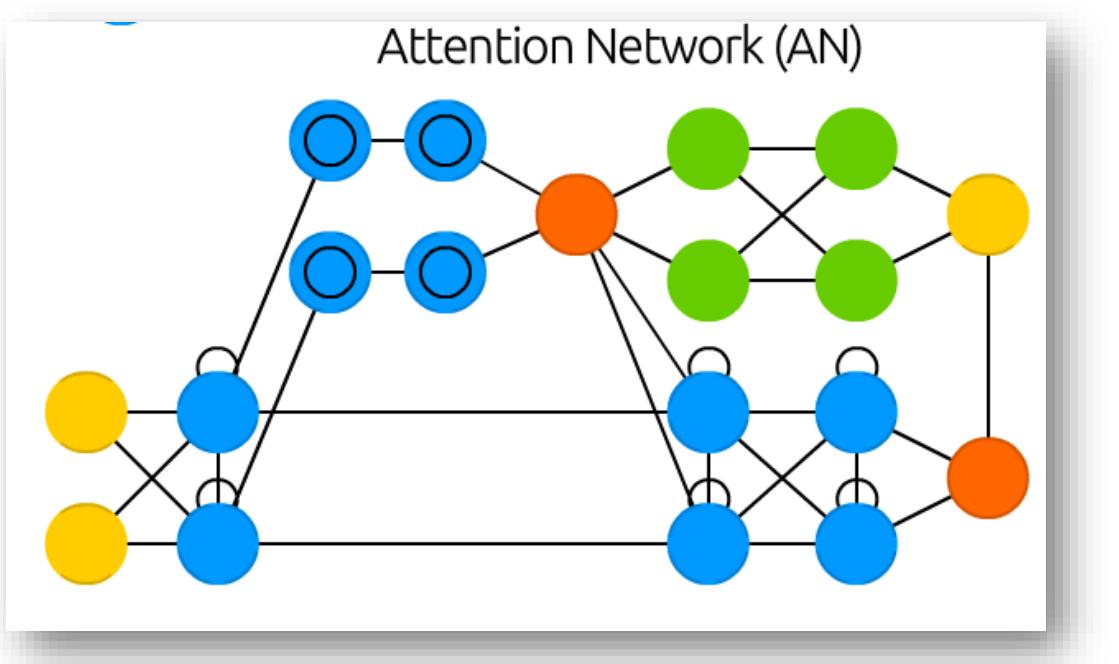
# Image generation

# Style transfer

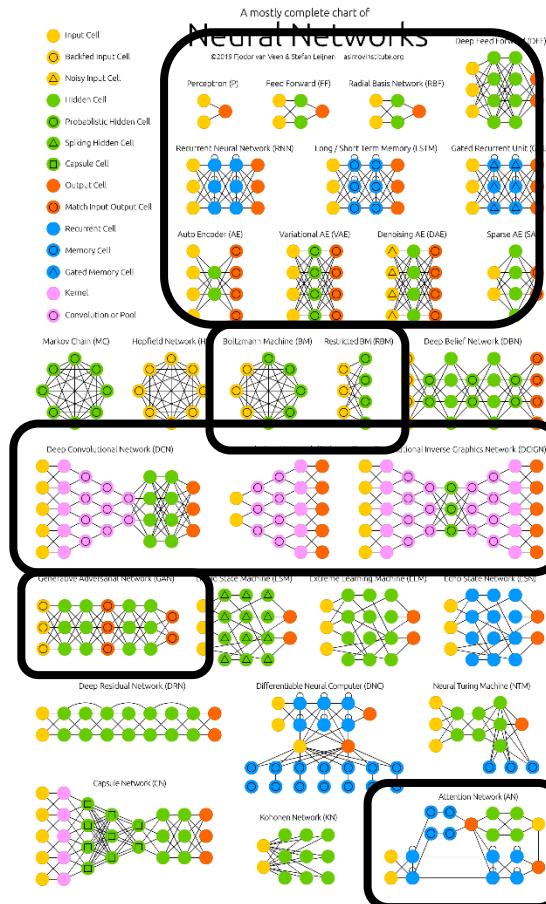
# Deep Fake



# Computer vision and NLP (state of the art?)



# Architectures we've seen



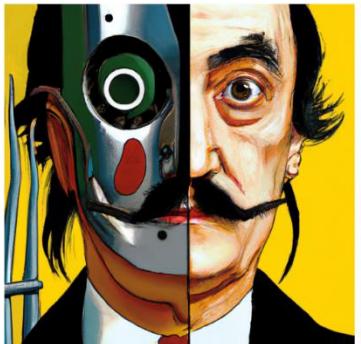


A deep learning model has many parameters.

Model	2D-CNN
	Params
VGG-16	134.7 M
ResNet-18	11.4 M
ResNet-34	21.5 M
ResNet-50	23.9 M
ResNet-101	42.8 M
ResNet-152	58.5 M
DenseNet-121	7.2 M
DenseNet-169	12.8 M

# SOTA models have 100B+ parameters!

**138 Billion parameters**  
(DALL-E, 2022)



vibrant portrait painting of Salvador Dalí with a robotic half face



a shiba inu wearing a beret and black turtleneck



a close up of a handpalm with leaves growing from it



an espresso machine that makes coffee from human souls, artstation



panda mad scientist mixing sparkling chemicals, artstation



a corgi's head depicted as an explosion of a nebula

**175 Billion parameters**  
(GPT-3, 2020)

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A deep learning model has many parameters.

This means training a deep learning model from scratch can:

1. take a looooong time
2. require lots of data
3. require a lot of computing power
4. have a huge carbon footprint!



Hardware infrastructure becomes important

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Have you ever thought  
about  
**Cloud Computing?**

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A GPU could help...

A deep learning model has many parameters.

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No more tabular data!

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How about building a  
**pipeline** to feed all that  
data in?

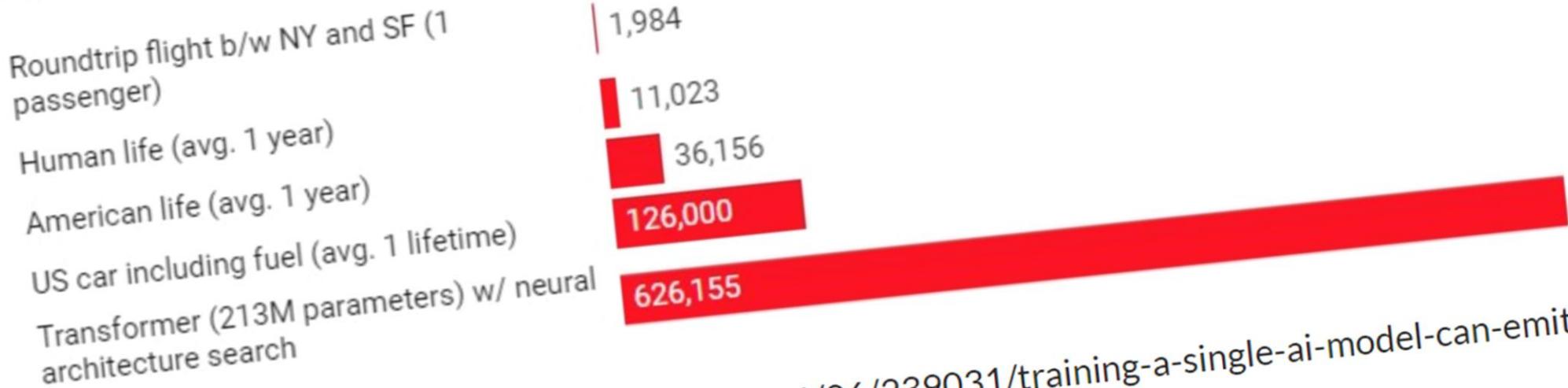
A **deep** learning model has  
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4. have a huge carbon footprint!

## **COMMON CARBON FOOTPRINT MARKS**

in lbs of CO<sub>2</sub> equivalent



(Image: <https://www.technologyreview.com/2019/06/06/239031/training-a-single-ai-model-can-emit-as-much-carbon-as-five-cars-in-their-lifetimes/>)



Whenever possible, use [transfer learning](#).



Whenever possible, use **transfer learning**.

Pick a network that has already been trained and showed to perform well on a task similar to your own...

... and use **your data to train only the last layer**, keeping all other weights intact.

(Why would this work?)



Look for these in your Literature Review.  
In computer vision, some examples are...

ResNet-18

GoogLeNet

VGG-19

etc.

In NLP, some examples are...

BERT and its variants

GPT-3

etc.

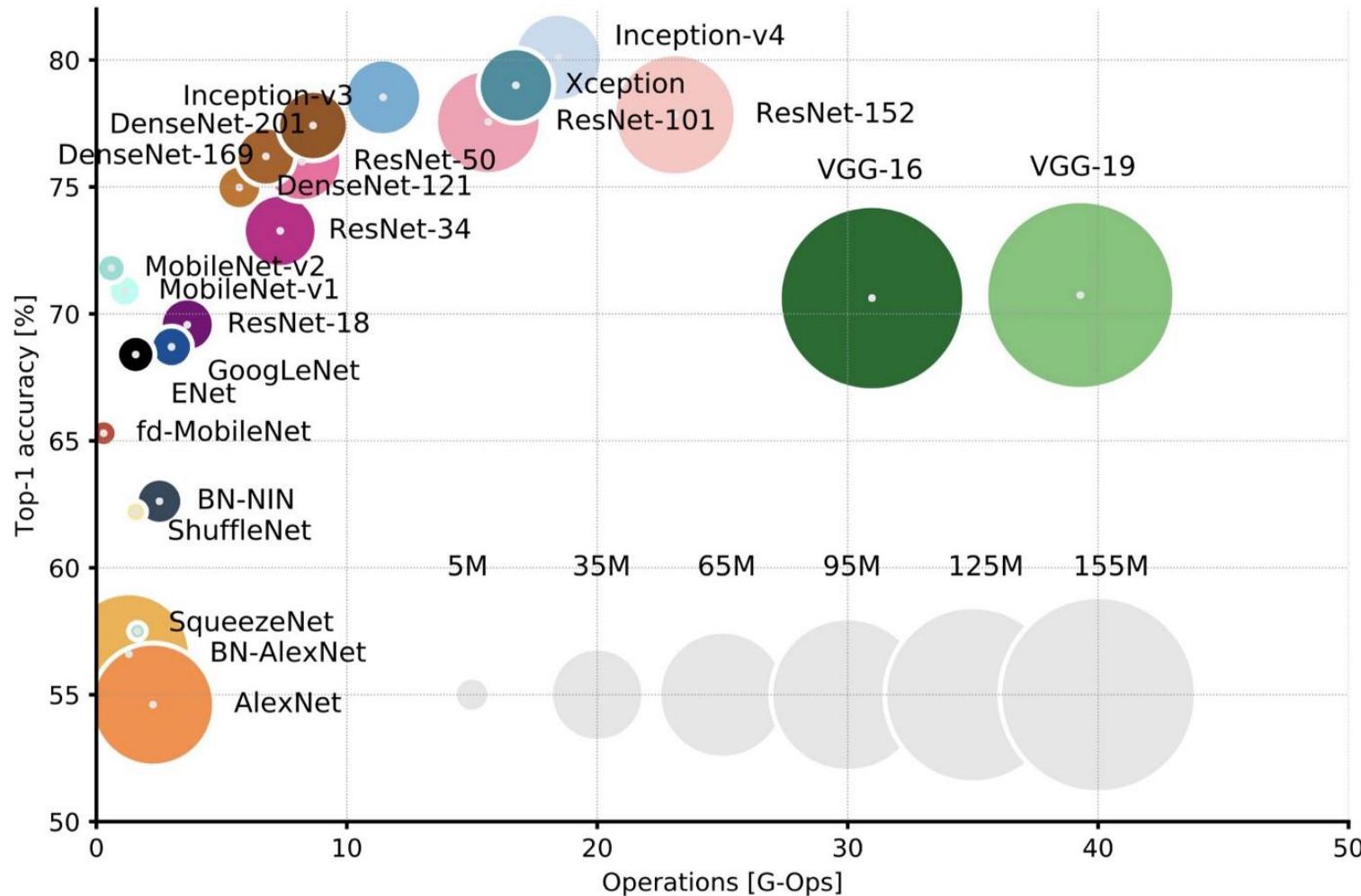
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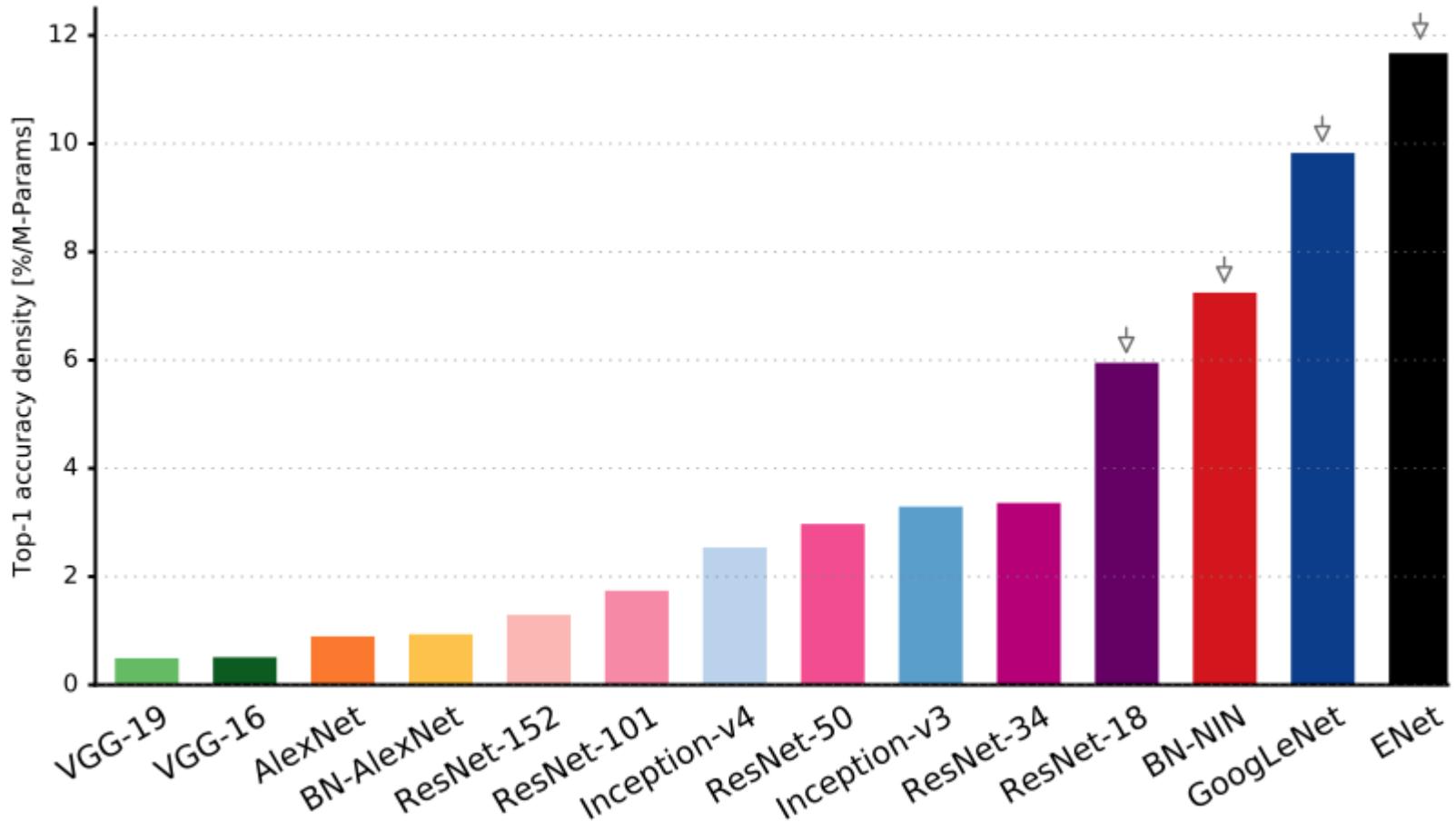
(Why would this work?)

# Models have been getting better, but bigger...



Top1 vs. operations, size  $\propto$  parameters.  
Top-1 one-crop accuracy versus  
amount of operations required for a  
single forward pass. See also [here](#)

... But they haven't grown better just because they're growing larger



# Adversarial attacks on AI



Hummingbird

Minor perturbation →



Hammer



Hare

Minor perturbation →



Desk

