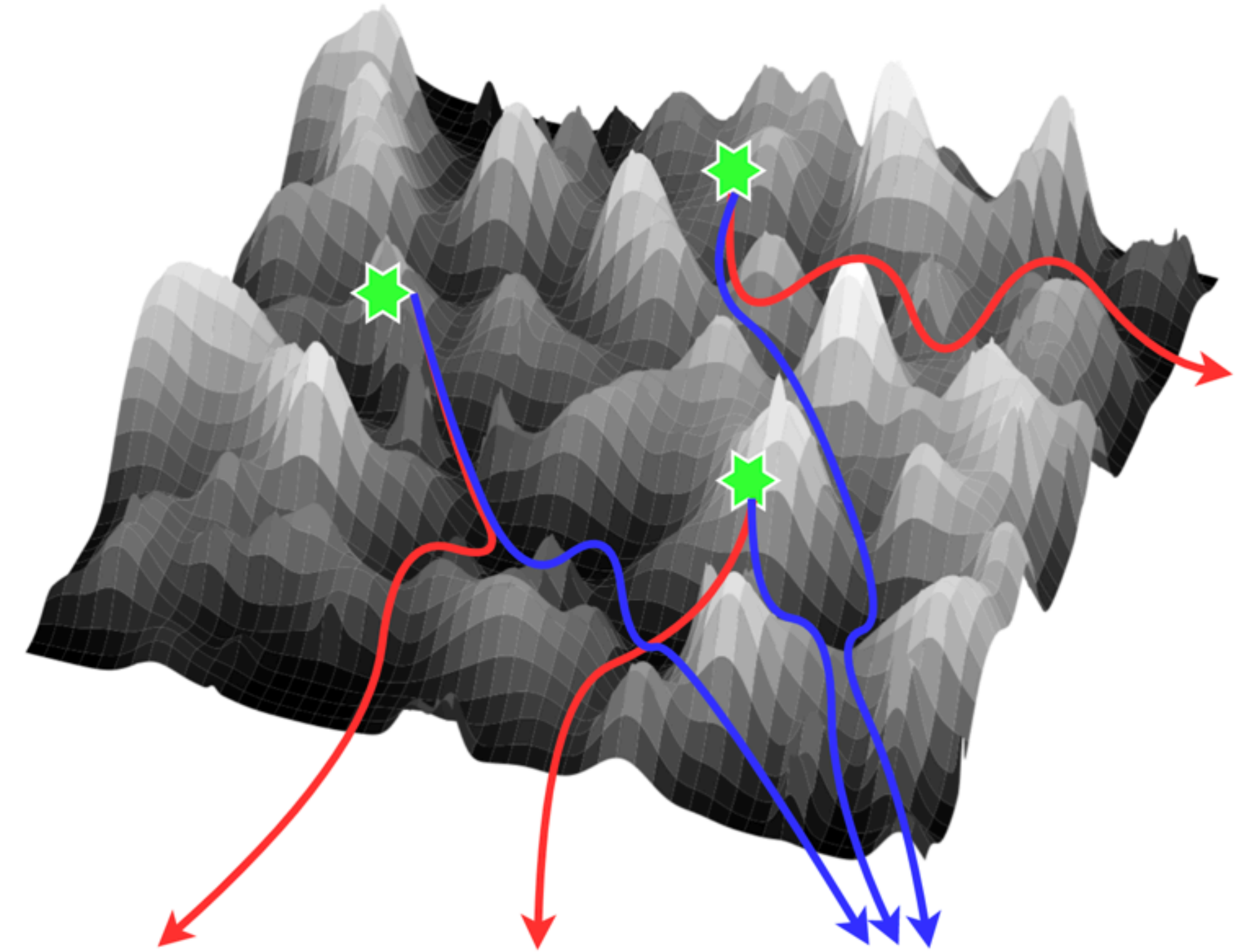


# An introduction to neural networks

Sebastian Goldt (SISSA, Trieste)



Junior Math Days @ SISSA — dec 2023



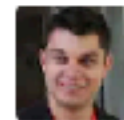
# New machine learning breakthroughs...



<https://twitter.com/StelfieTT>



# New machine learning breakthroughs...



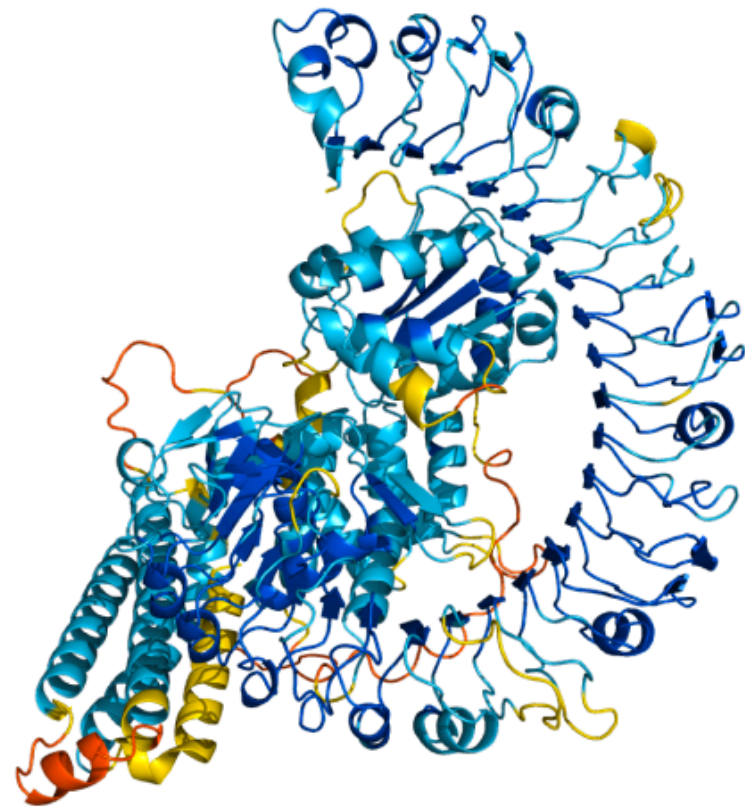
Hey ChatGP, I'm interested in the theory of neural networks. Do you know anything about that?



Yes, I'm familiar with the theory of neural networks. Neural networks are a type of machine learning algorithm that are modeled after the structure and function of the human brain. They consist of layers of interconnected "neurons," which process and transmit information. Neural networks are commonly used for tasks such as image and speech recognition, and they have been instrumental in many of the recent advances in artificial intelligence. 👍 🗨️





# You heard the news



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Yes, I'm familiar with the theory of neural networks. Neural networks are a type of machine learning algorithm that are modeled after the structure and function of the human brain. They consist of layers of interconnected "neurons," which process and transmit information. Neural networks are commonly used for tasks such as image and speech recognition, and they have been instrumental in many of the recent advances in artificial intelligence.  

In the words of the New York Times:

*“The Navy revealed the embryo of an electronic computer that it expects will be able to walk, talk, see, write, reproduce itself and be conscious of its own existence”*

# NEW NAVY DEVICE LEARNS BY DOING

Psychologist Shows Embryo  
of Computer Designed to  
Read and Grow Wiser

WASHINGTON, July 7 (UPI)  
—The Navy revealed the embryo of an electronic computer today that it expects will be able to walk, talk, see, write, reproduce itself and be conscious of its existence.

The embryo—the Weather Bureau's \$2,000,000 "704" computer—learned to differentiate between right and left after fifty attempts in the Navy's demonstration for newsmen.

The service said it would use this principle to build the first of its Perceptron thinking machines that will be able to read and write. It is expected to be finished in about a year at a cost of \$100,000.

*New York Times,  
July 8, 1958*

# The plan for today

## **What is a neural network?**

From neurons to networks.

## **The importance of feature learning**

The secret sauce of neural networks

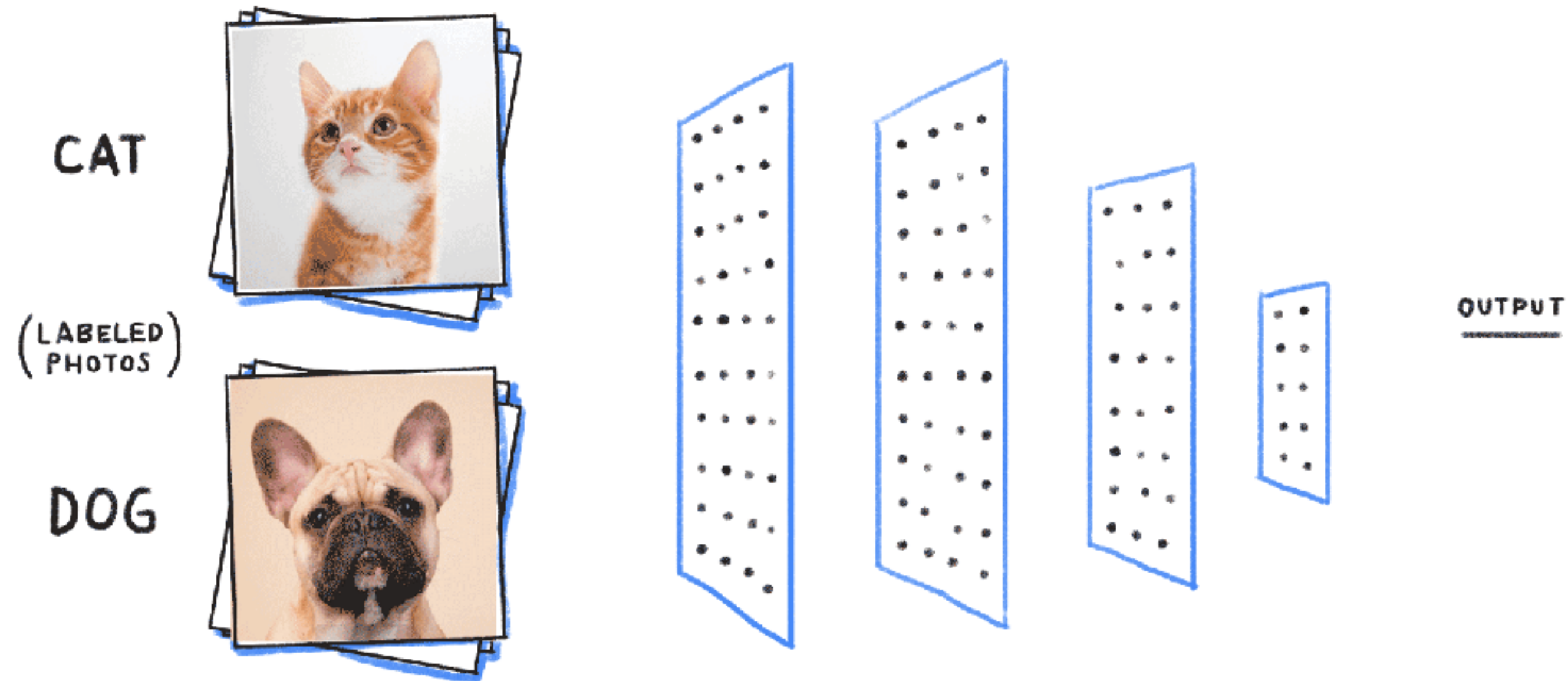
# Part I

**What is a neural network?**



# What is a neural network?

A neural network is a (complicated) function



*Animation courtesy of Aakash Srivastava*

Let's take a closer look at one of these small black dots...



# A single neuron

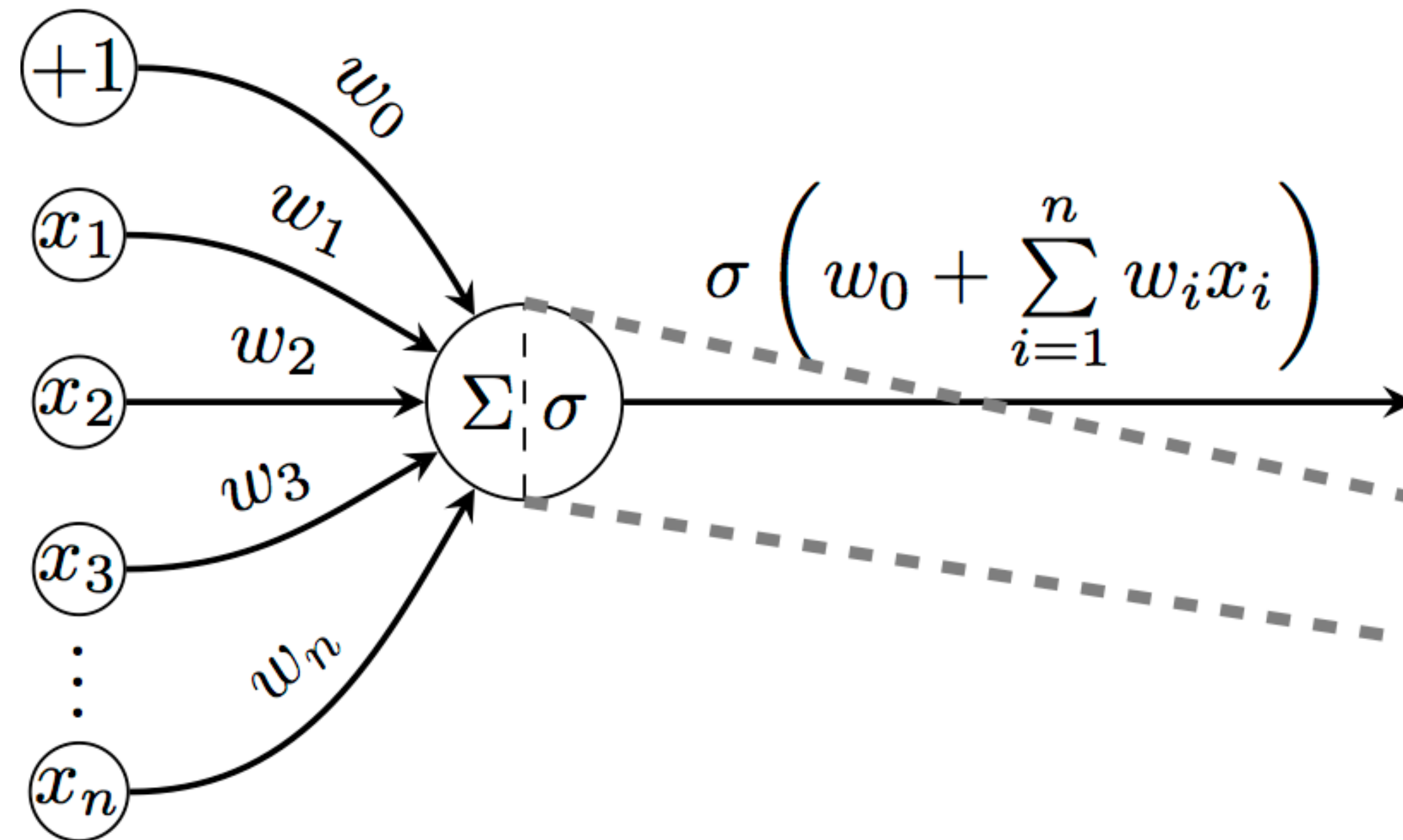
Many inputs, one output

A LOGICAL CALCULUS OF THE  
IDEAS IMMANENT IN NERVOUS ACTIVITY

WARREN S. MCCULLOCH AND WALTER PITTS

FROM THE UNIVERSITY OF ILLINOIS, COLLEGE OF MEDICINE,  
DEPARTMENT OF PSYCHIATRY AT THE ILLINOIS NEUROPSYCHIATRIC INSTITUTE,  
AND THE UNIVERSITY OF CHICAGO

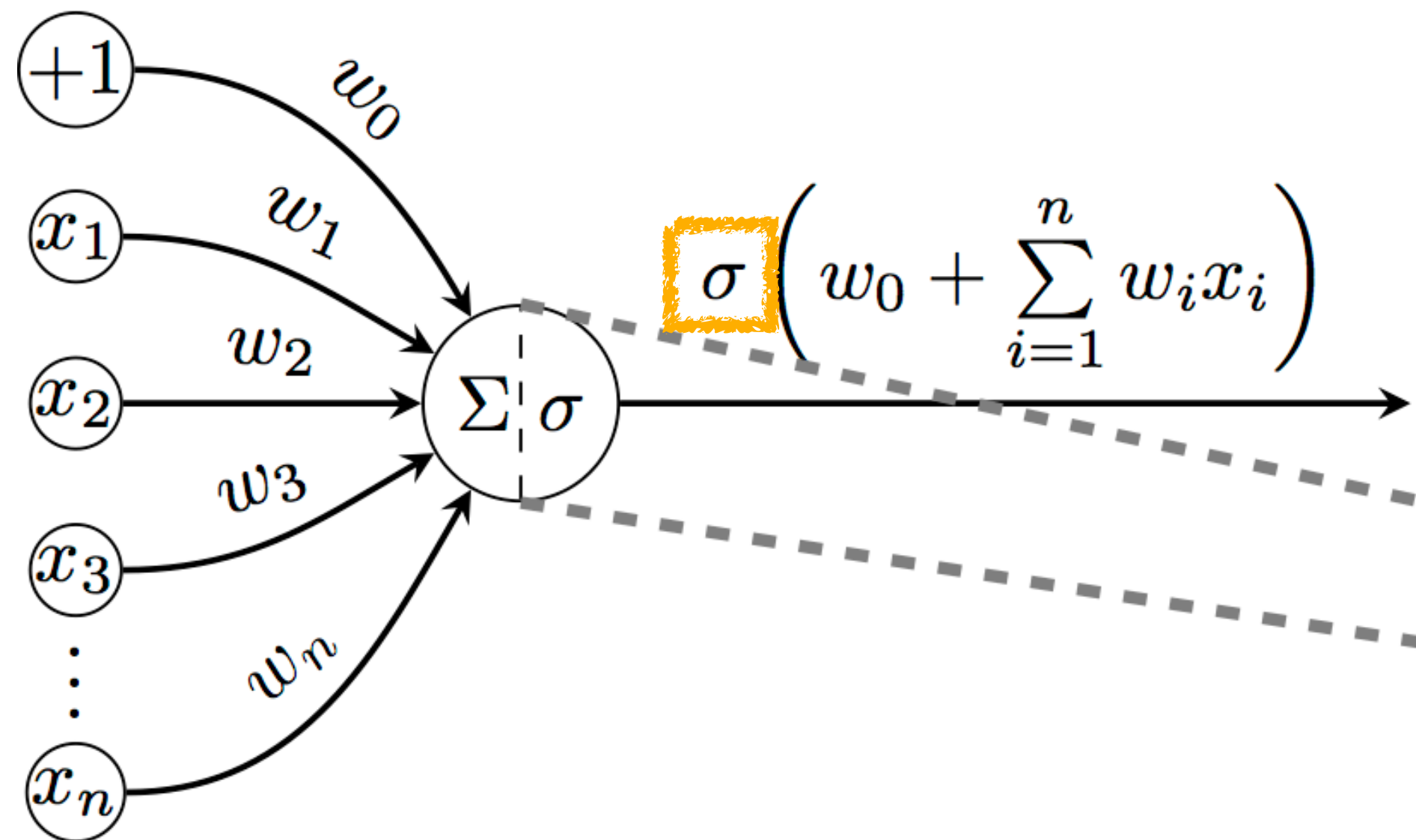
(1943)



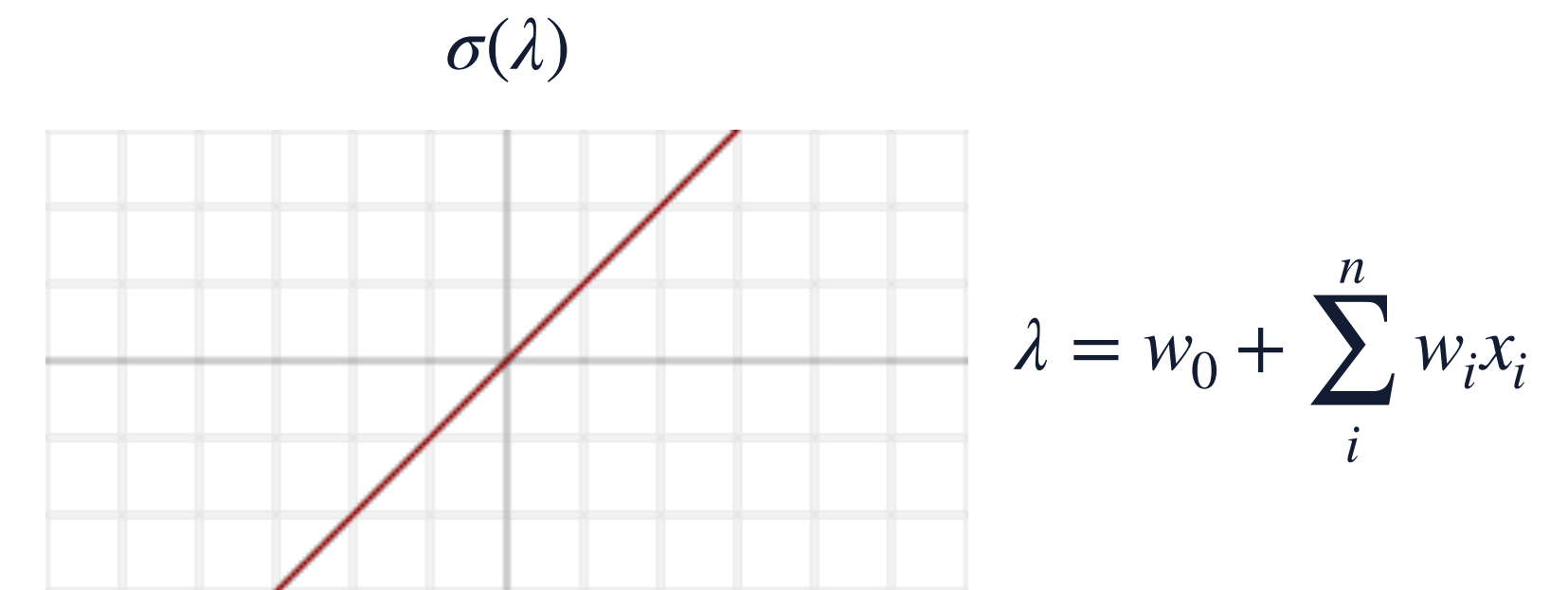
*Illustration by Petar Veličković*  
<https://github.com/PetarV-/TikZ>

# The activation function

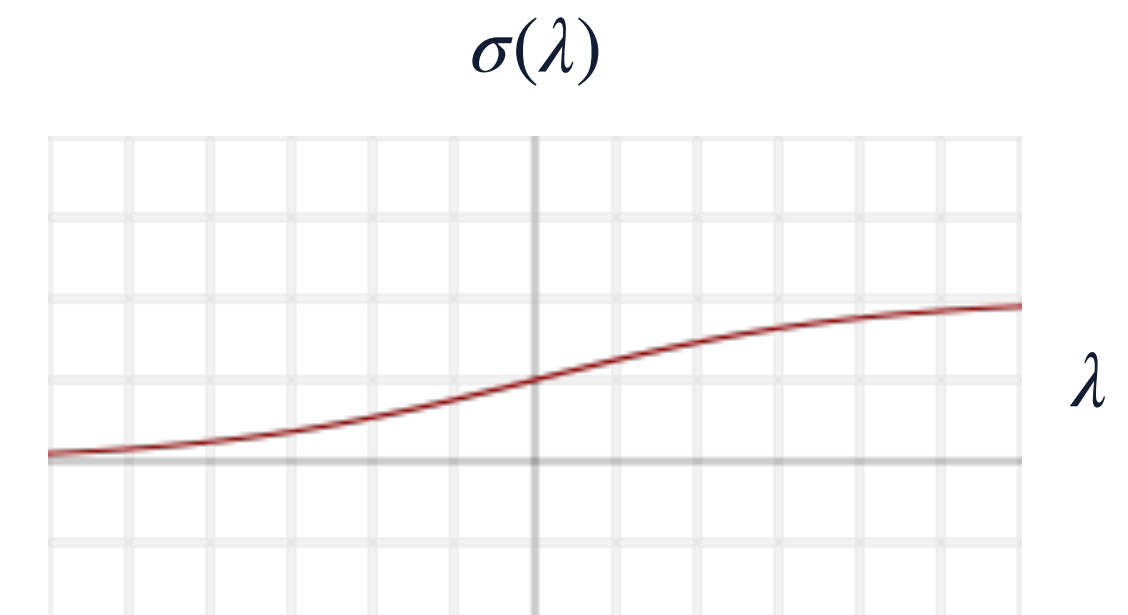
More than just summing up



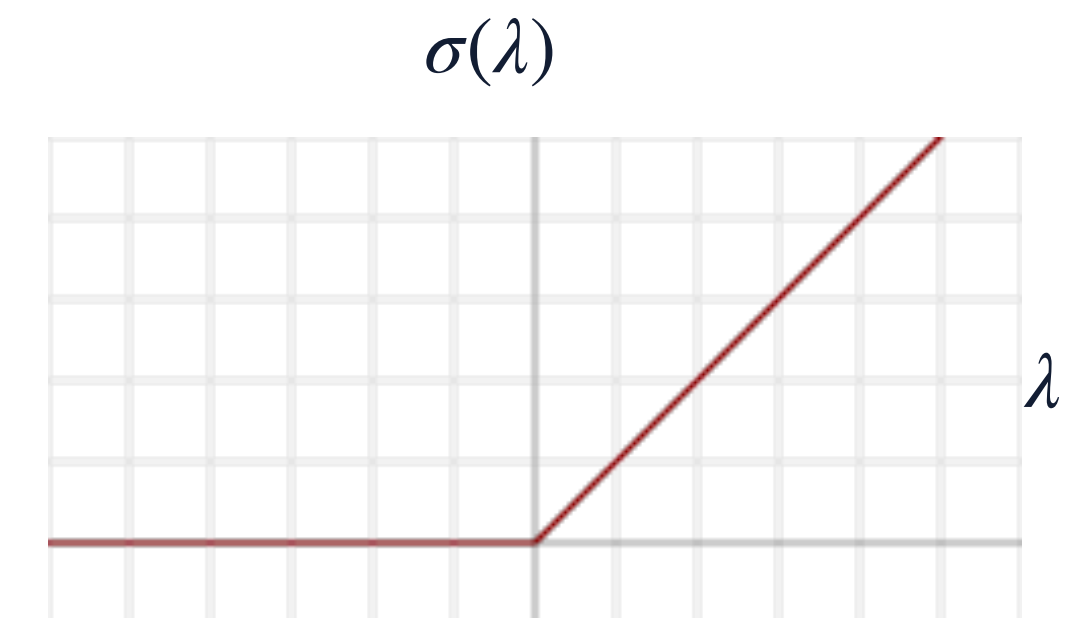
Linear



Sigmoid



ReLU





# A lot of neurons

Neurons can be assembled into layers

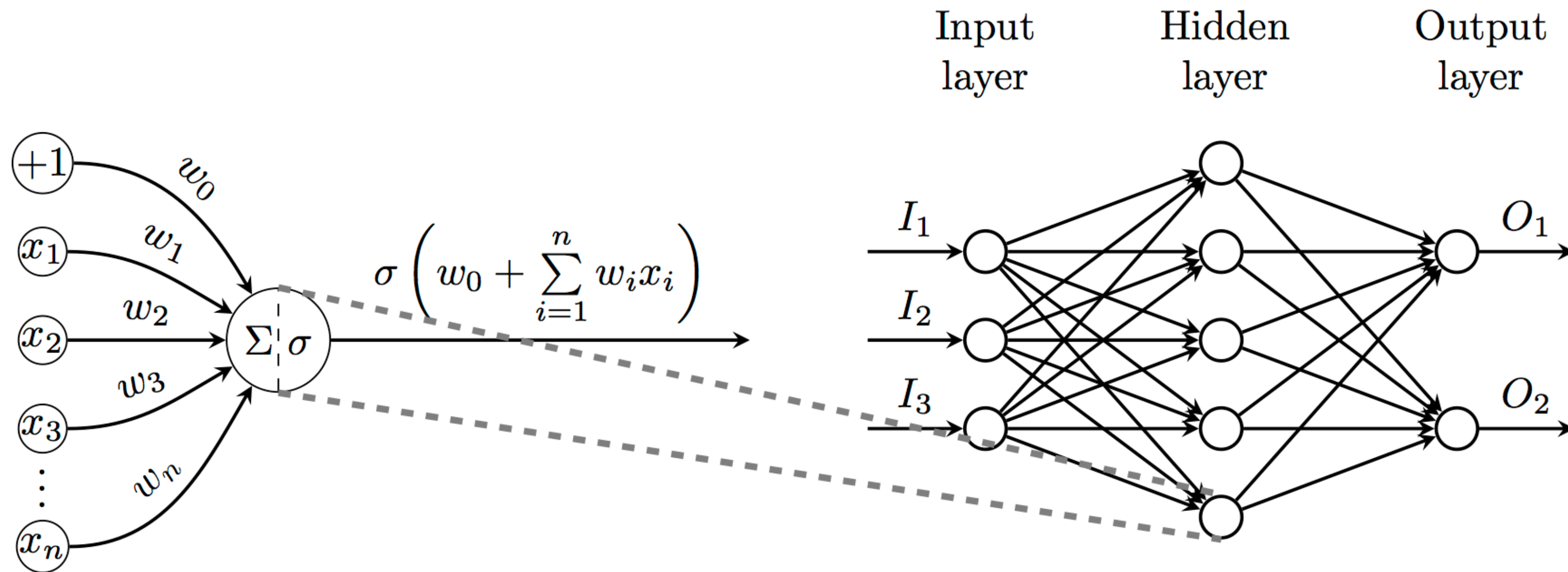
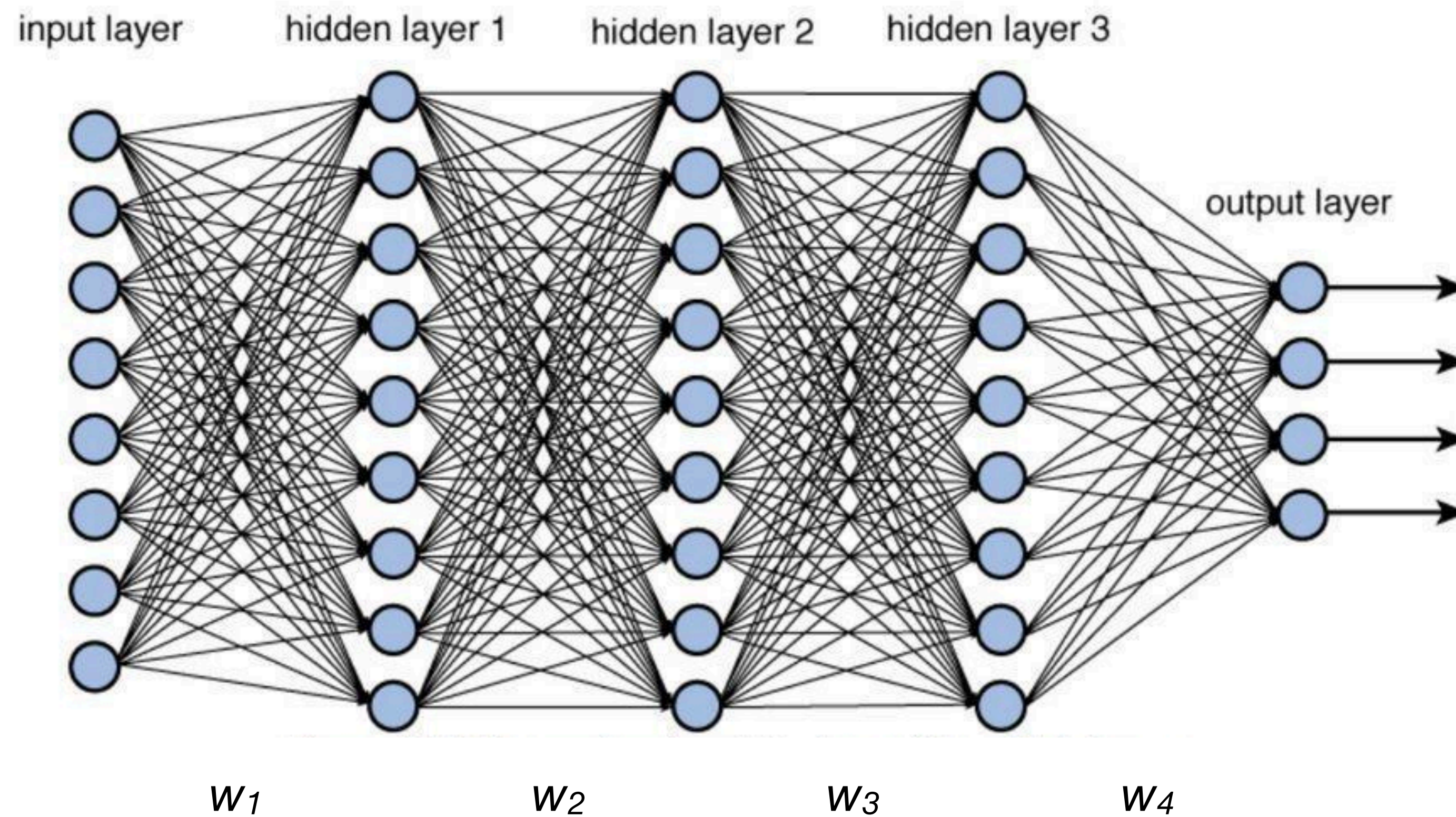


Illustration by Petar Veličković  
<https://github.com/PetarV-/TikZ>



# A lot of layers

Deep neural networks stack layers of neurons



$$y = w_4 \sigma (w_3 \sigma (w_2 \sigma (w_1 x)))$$



# A mostly complete chart of neural networks

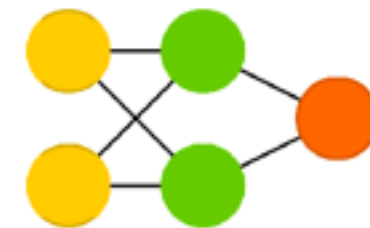
$y = w_4\sigma(w_3\sigma(w_2\sigma(w_1x)))$  is not the whole story

- Input Cell
- Backfed Input Cell
- △ Noisy Input Cell
- Hidden Cell
- Probabilistic Hidden Cell
- △ Spiking Hidden Cell
- Capsule Cell
- Output Cell
- Match Input Output Cell
- Recurrent Cell
- Memory Cell
- △ Gated Memory Cell
- Kernel
- Convolution or Pool

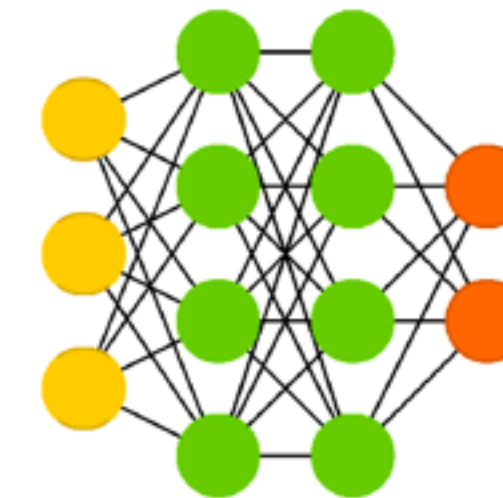
Perceptron (P)



Feed Forward (FF)

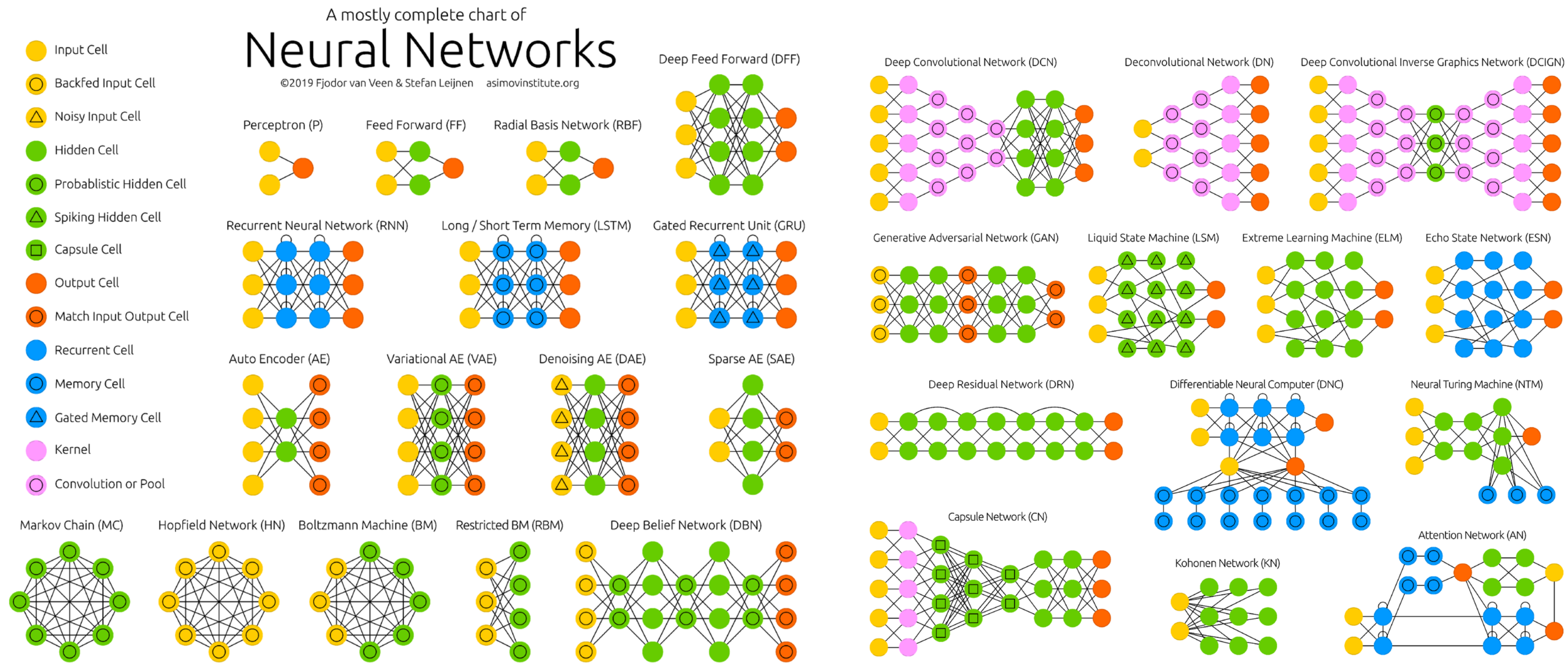


Deep Feed Forward (DFF)



# A mostly complete chart of neural networks

$y = w_4 \sigma(w_3 \sigma(w_2 \sigma(w_1 x)))$  is not the whole story



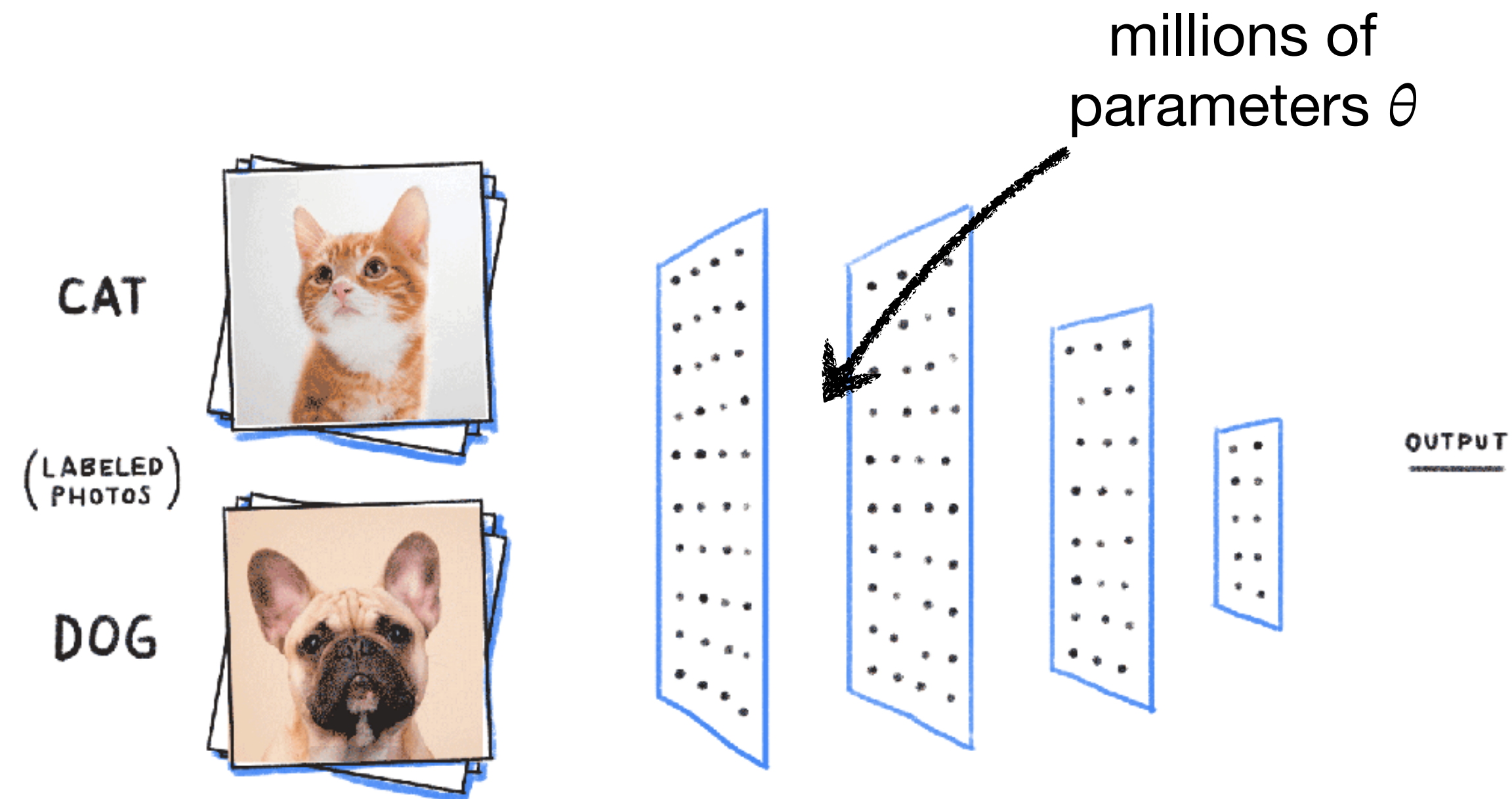




**Which architecture  
for which data?**

# Training a neural network

Having the architecture, how do we find the weights?



*Animation courtesy of Aakash Srivastava*



# Training a neural network

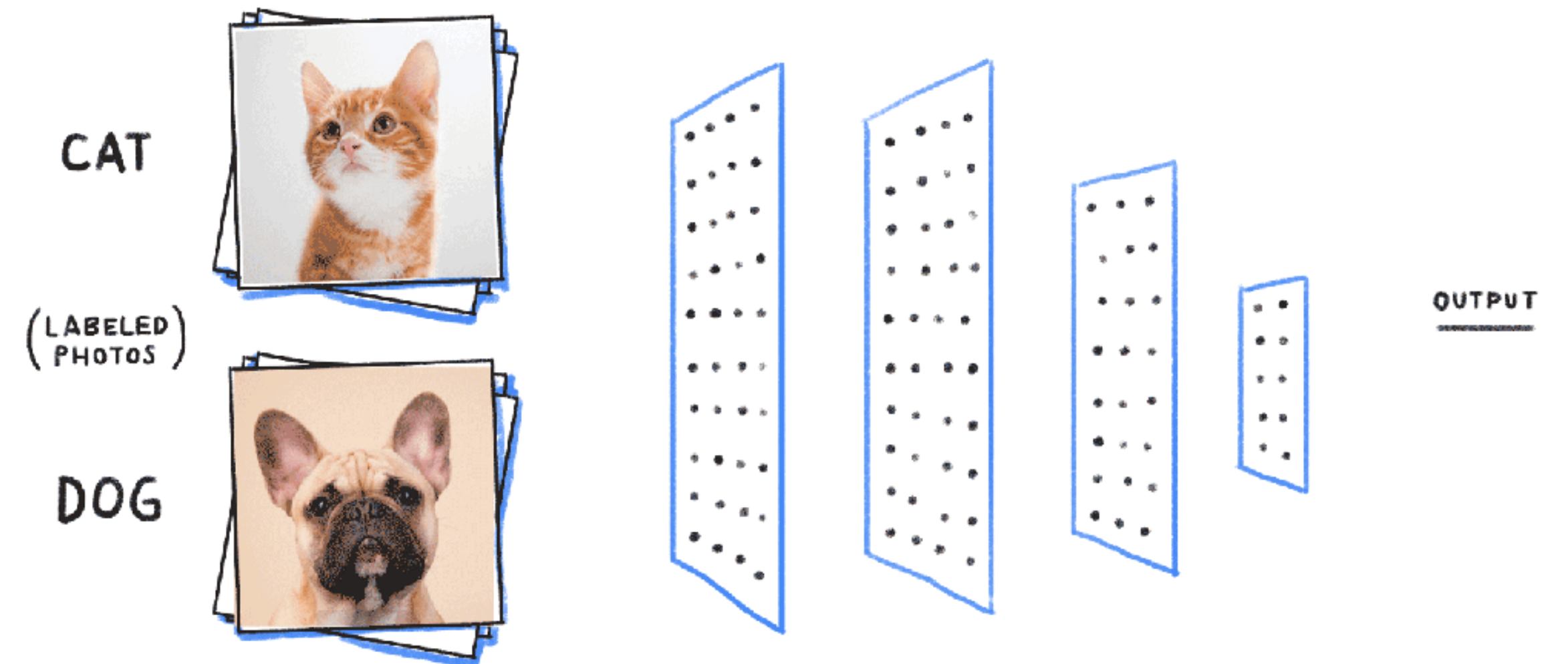
Having the architecture, how do we find the weights?

- Given a data set  $\mathcal{D} = \{(x_i, y_i)\}_{i=1}^N$

$\text{loss}(\theta) = \#$  of mis-classified  
training images

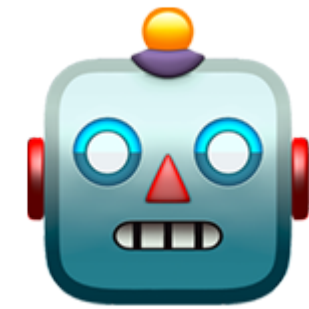
$$\theta^* = \underset{\theta}{\text{argmin}} \text{loss}(\theta)$$

$$\epsilon_t = \text{loss}(\theta^*)$$



*Animation courtesy of Aakash Srivastava*

# Training a neural network in theory



It's not so easy!

**TRAINING A 3-NODE NEURAL NETWORK  
IS NP-COMPLETE**

Avrim Blum\*  
MIT Lab. for Computer Science  
Cambridge, Mass. 02139 USA

Ronald L. Rivest†  
MIT Lab. for Computer Science  
Cambridge, Mass. 02139 USA

*Advances in Neural Information Processing (1989)*



# TRAINING A 3-NODE NEURAL NETWORK IS NP-COMPLETE

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*Advances in Neural Information Processing (1989)*

*Extends a previous result by Judd (1987)*

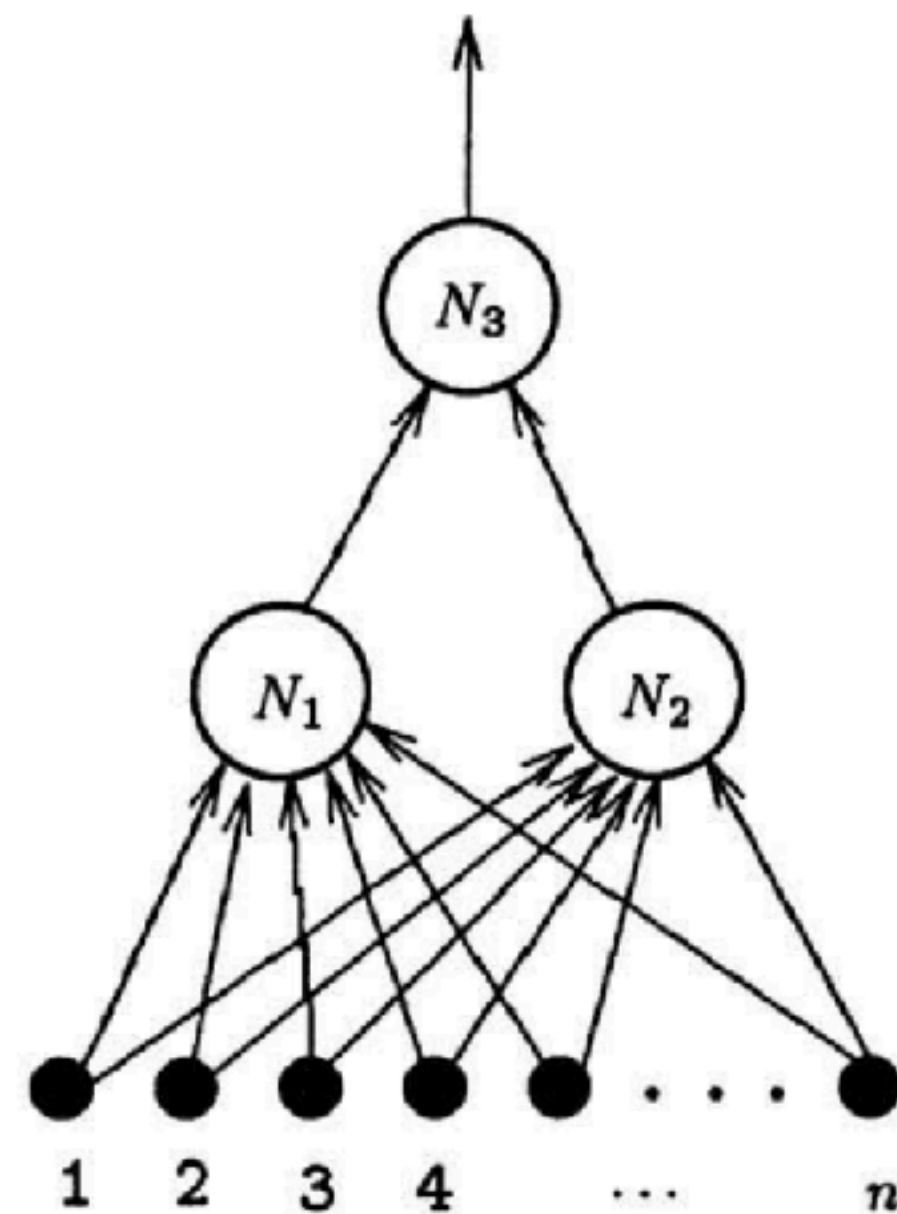


Figure 1: The three node neural network.

**Given:** A set of  $O(n)$  training examples on  $n$  inputs

**Question:** Do there exist linear threshold functions such that the three-node network fits the training set?

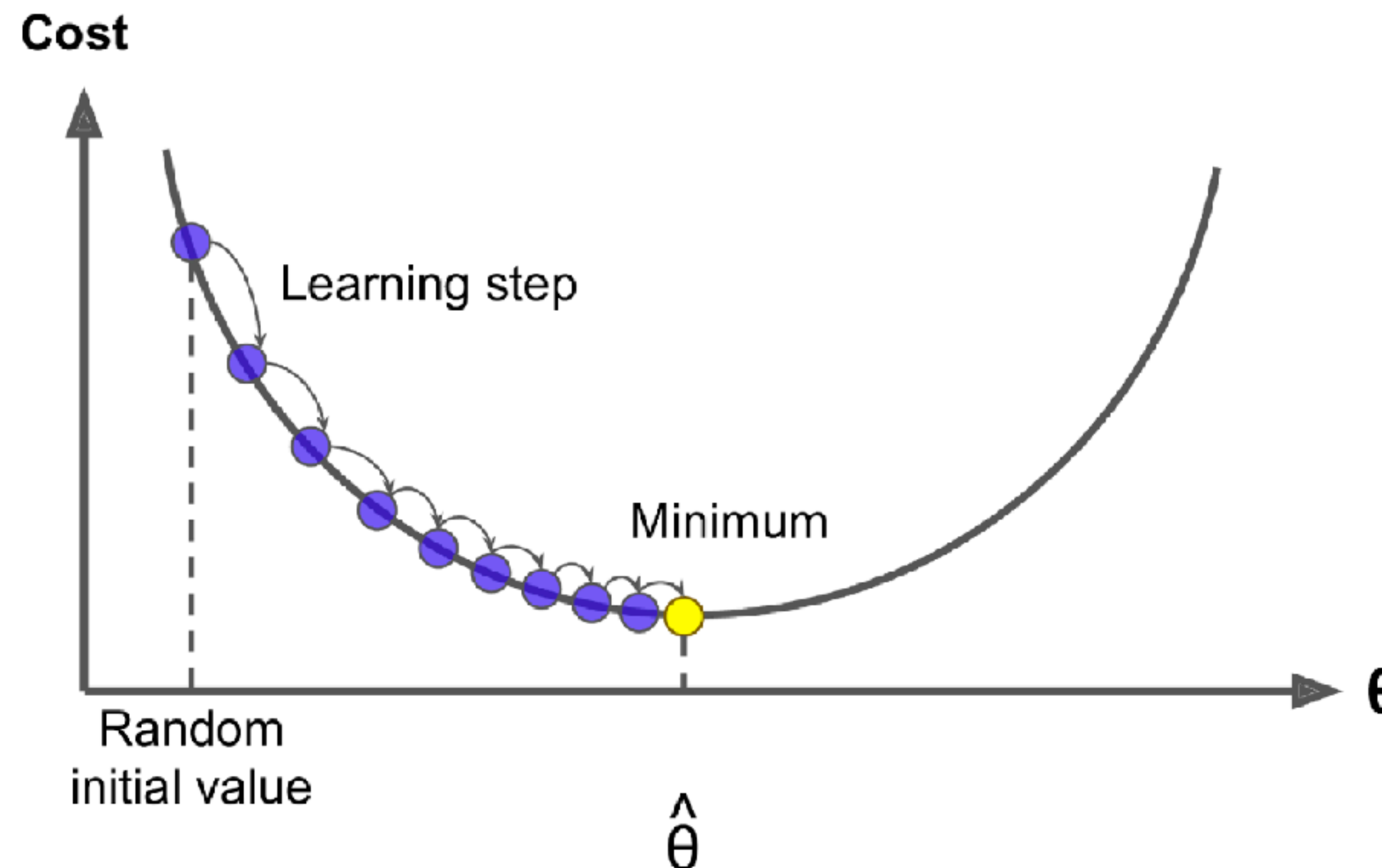
*NP hard*



# Training a neural network in practice

Forget you can't do it, and do it anyway!

We use **gradient descent** (& its friends: stochastic, momentum)

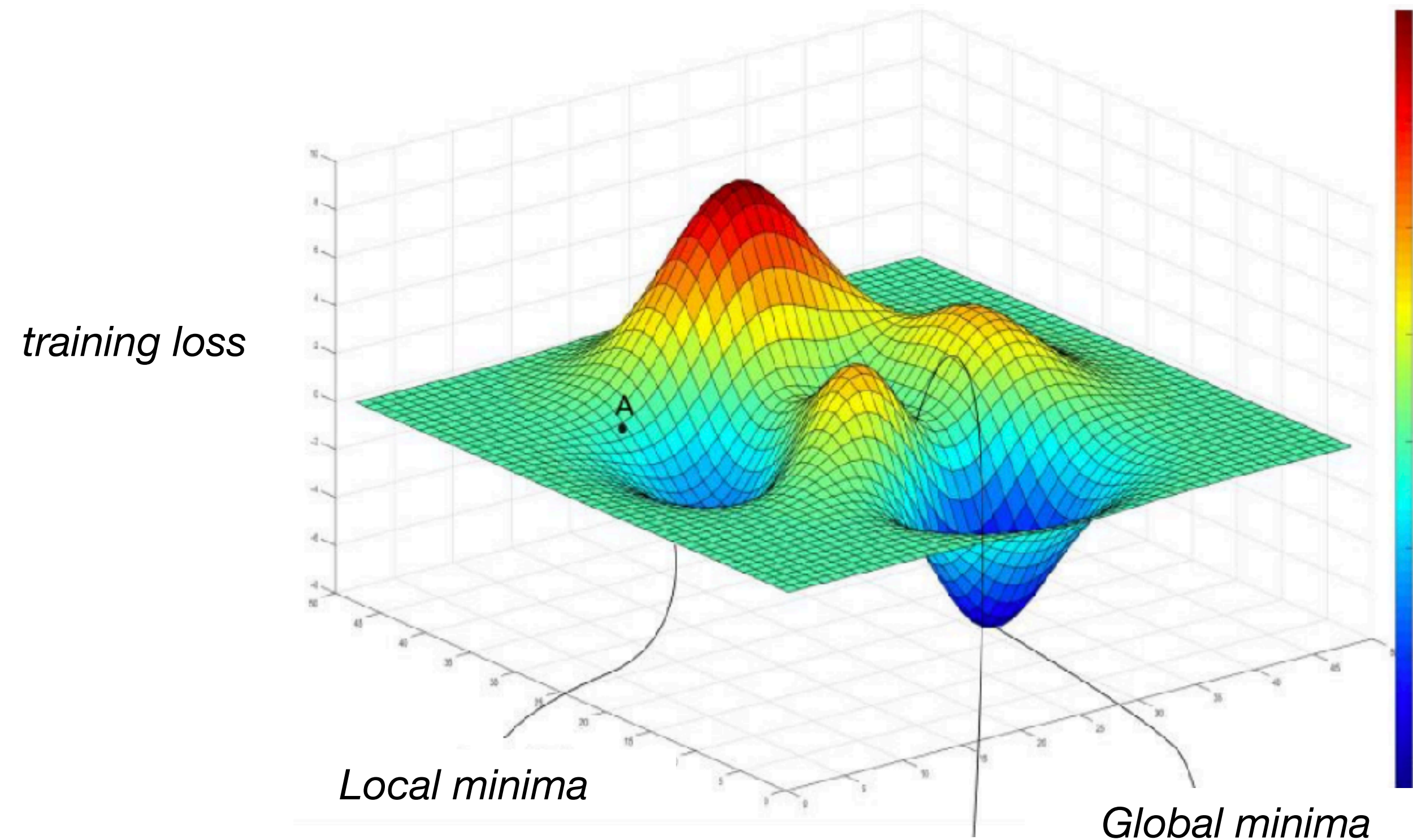




# Training a neural network in practice

Forget you can't do it, and do it anyway!

We use **gradient descent** (& its friends: stochastic, momentum)



*Saddle points  
(gradient = 0,  
but not local minima)*

*Cartoon by Andrej Risteski*

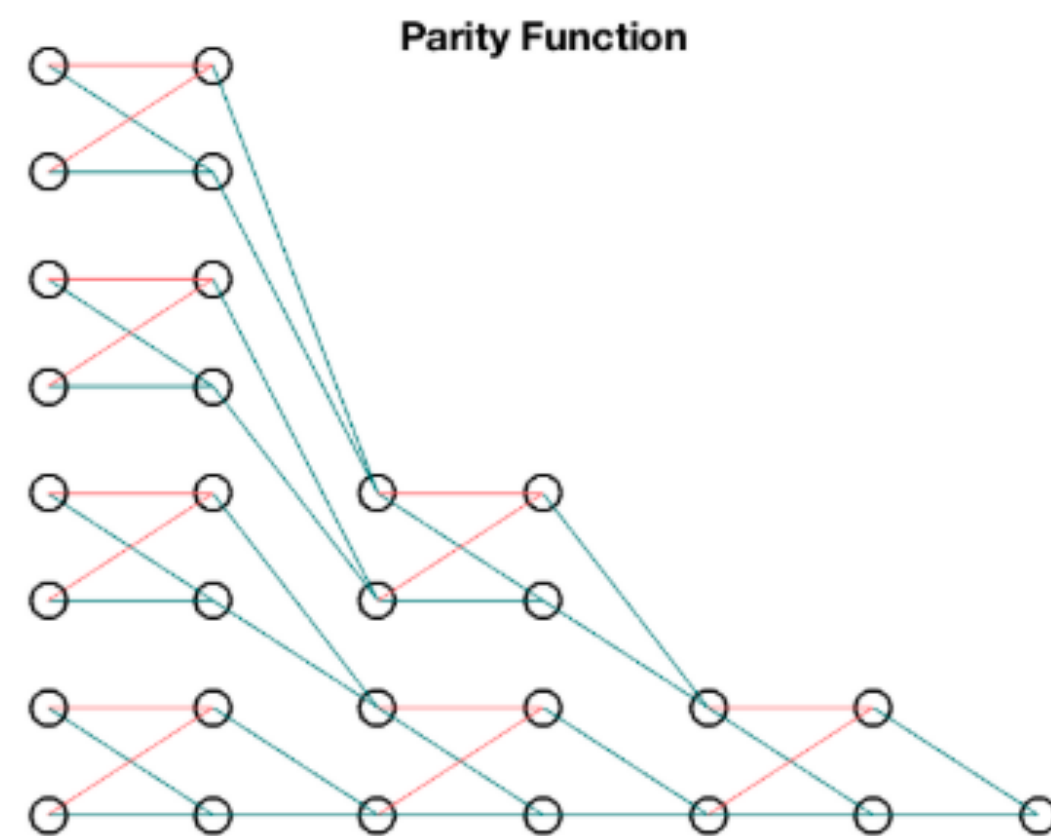
# Can we learn **anything** with gradient descent?

ARE EFFICIENT DEEP REPRESENTATIONS  
LEARNABLE?

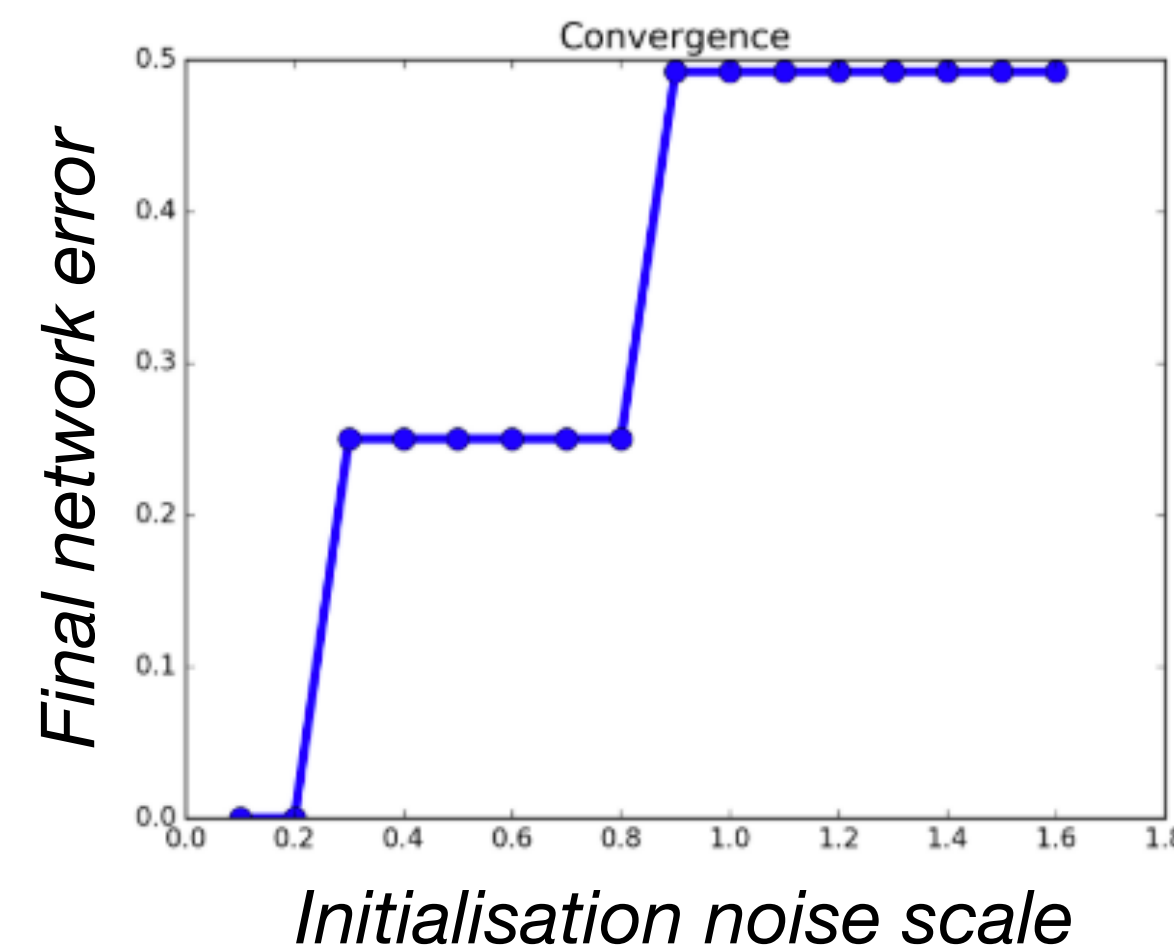
Maxwell Nye  
Massachusetts Institute of Technology  
mnye@mit.edu

Andrew Saxe  
Harvard University  
asaxe@fas.harvard.edu

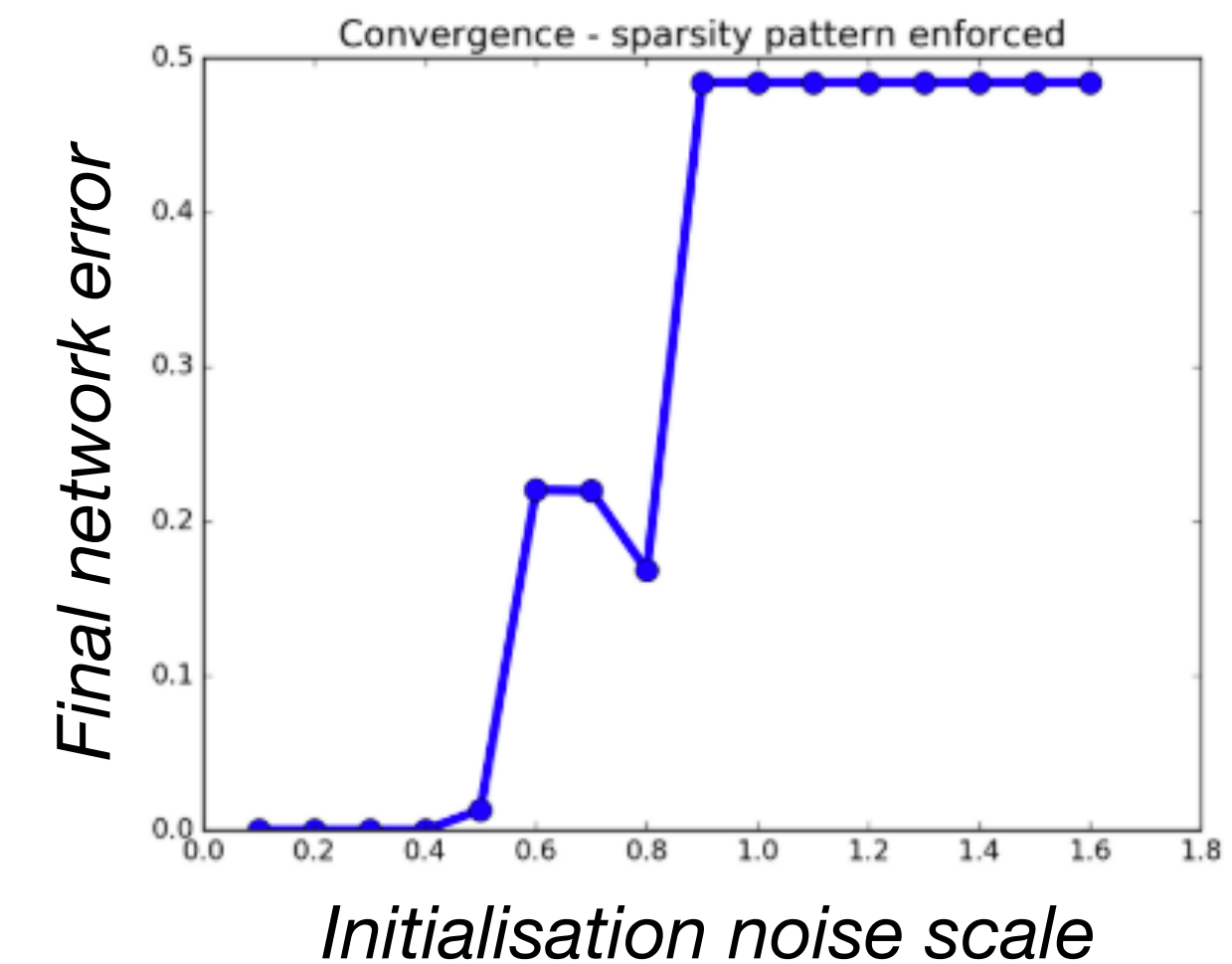
- **Parity function:** 1 if a binary string has *odd* number of ones.
- $(0, 1, 1, 1) \rightarrow 1$
- $(1, 1) \rightarrow 0$
- etc.



The parity function can be  
**expressed exactly**  
by this neural network



Gradient descent  
struggles to learn  
this function...



... even if we explicitly  
impose the sparsity  
of the target network.



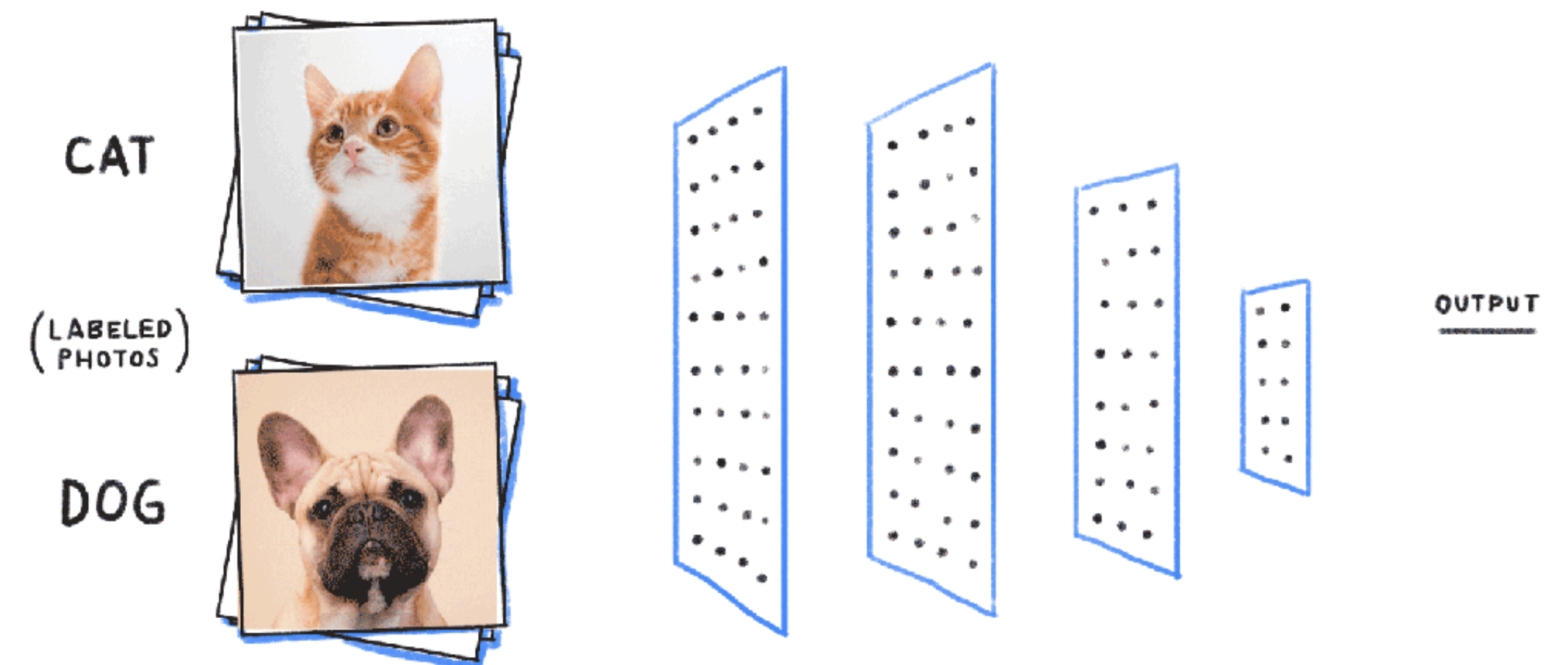
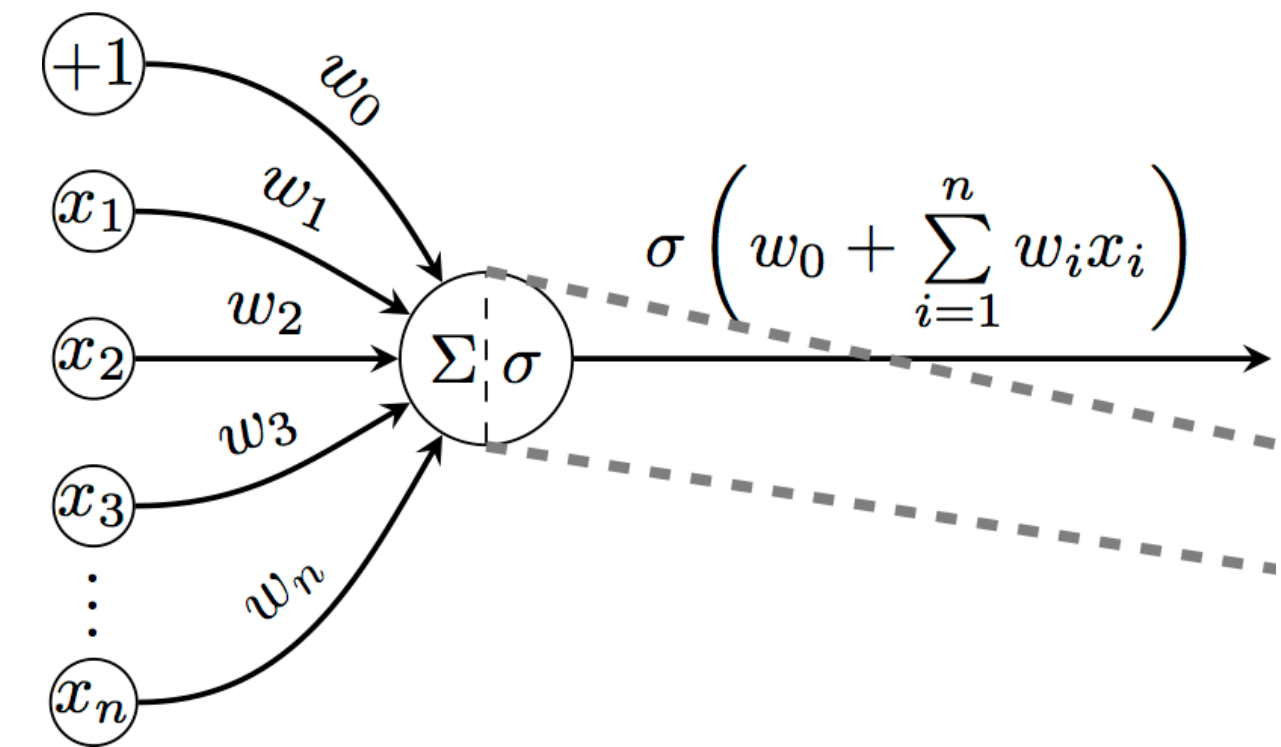
**What can GD learn, and why  
does it work on real data?**



# A first summary

What are neural networks?

- A **Neuron** can be represented as a single number, nonlinearity (weighted sum of inputs).
- Many neurons yield a neural **network**.
- The ordering & wiring of neurons determines the **architecture** of the neural network.
- The weights of a neural network are found by **training** the network (using stoch grad desc)



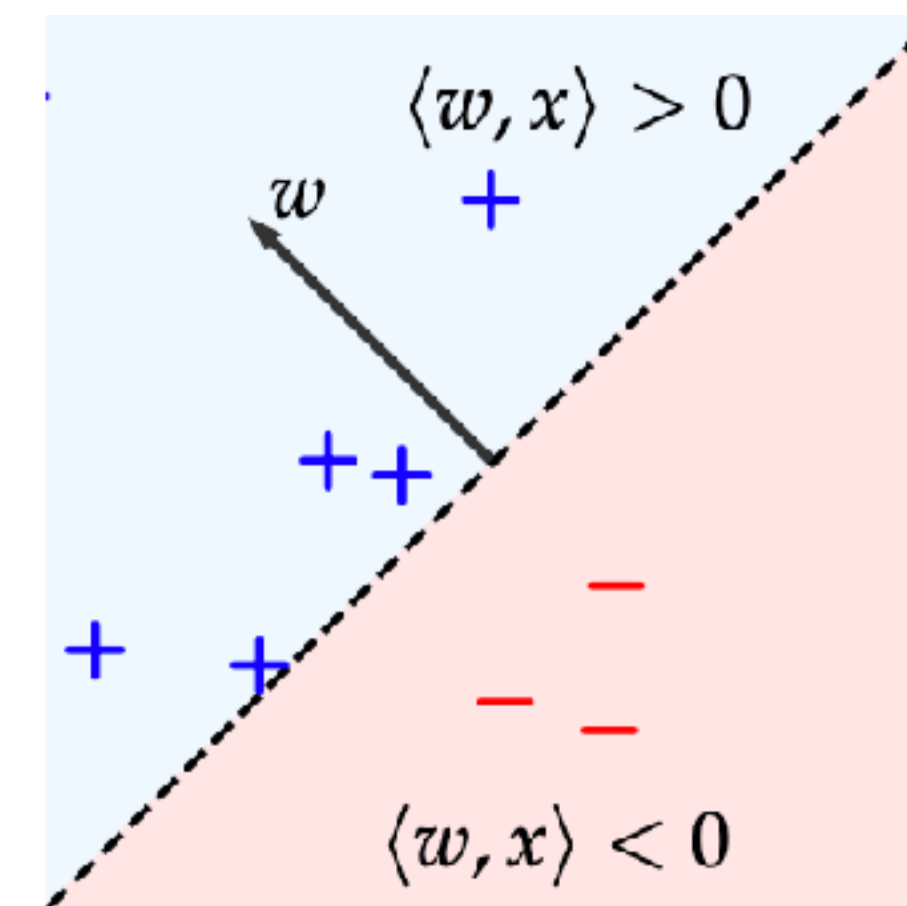
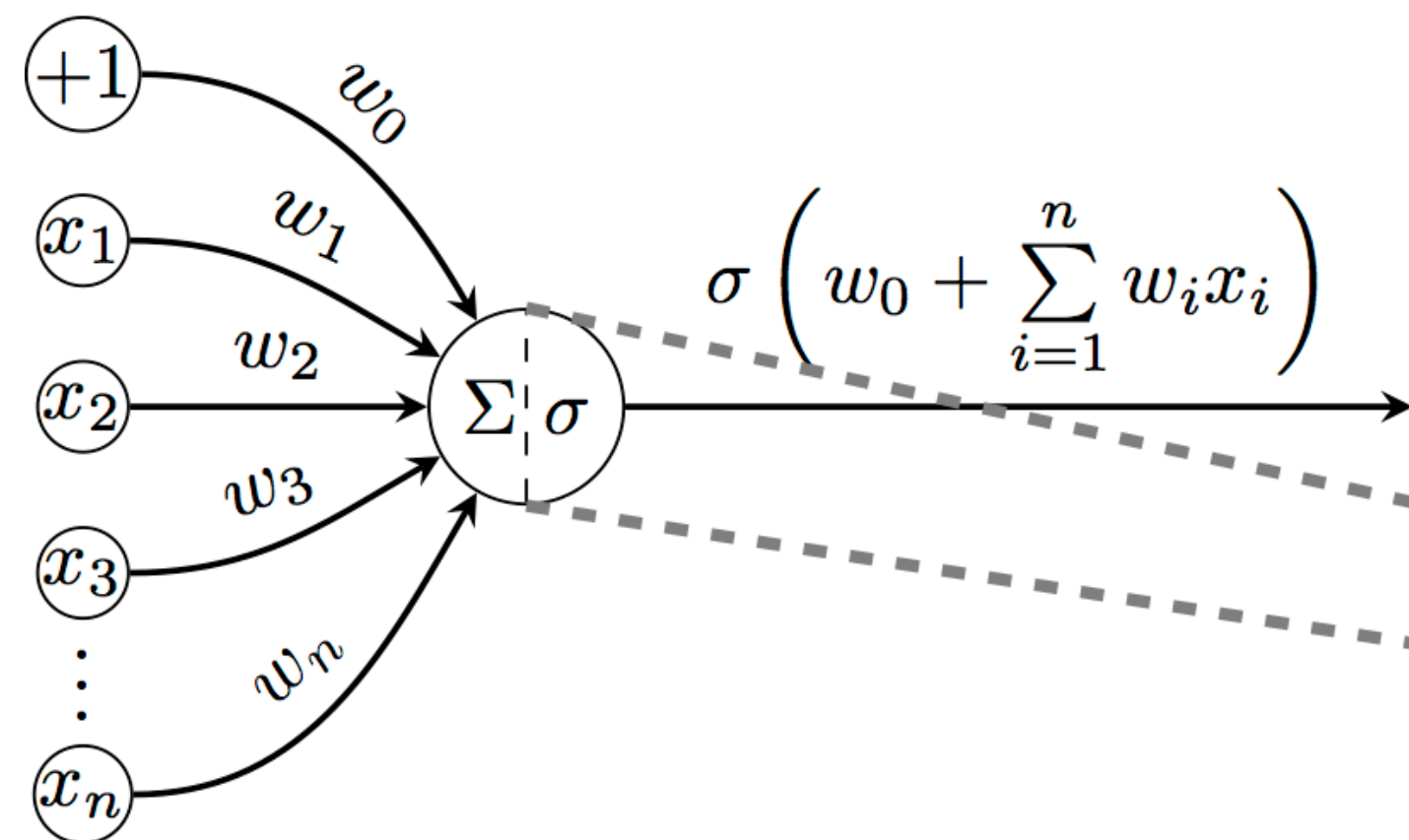
# Part II

## Feature learning

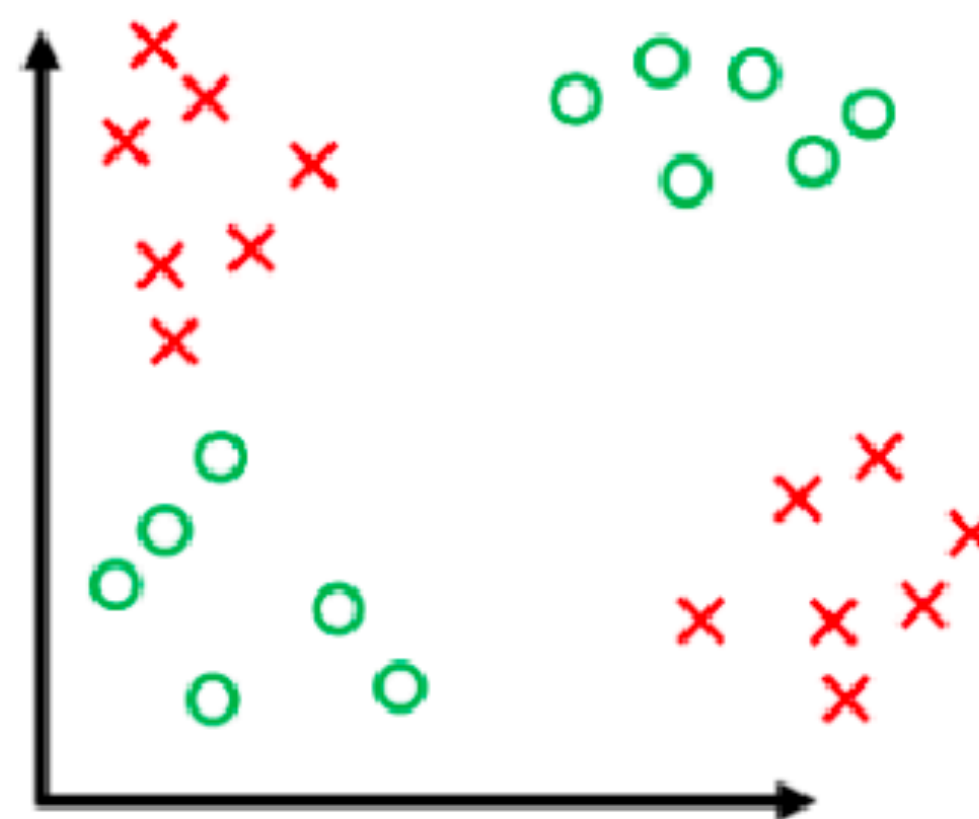
a.k.a. the *secret sauce* in neural networks

# What can a single neuron learn?

Turns out, it's rather limited



Minsky and Papert (1969)



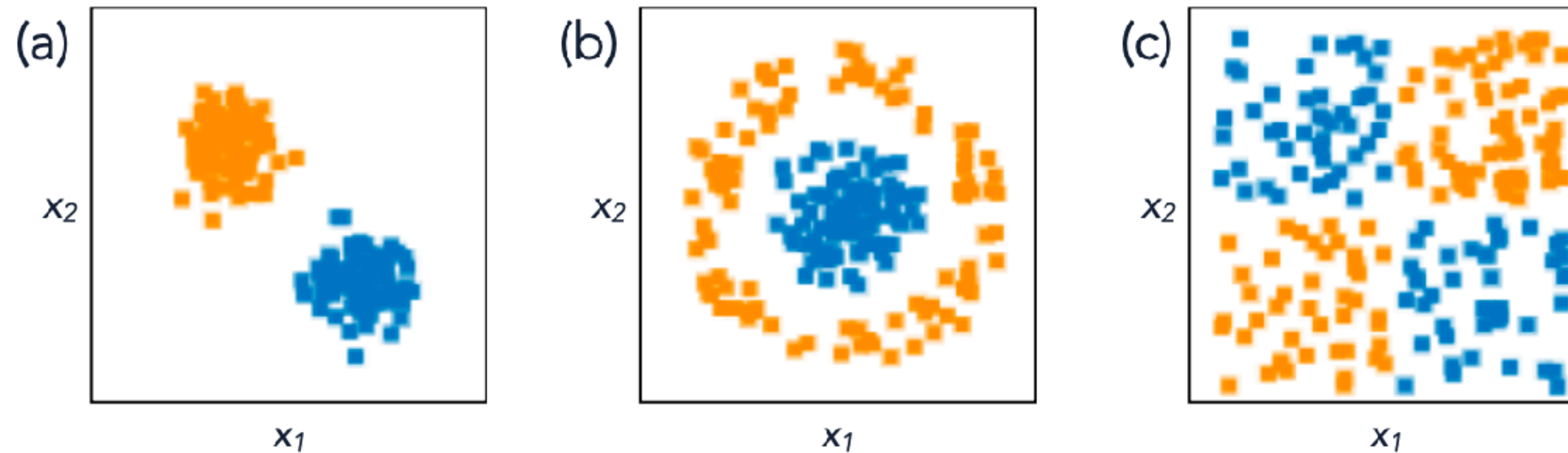
Illustrations by Petar Veličković  
<https://github.com/PetarV-/TikZ>  
and Hardt & Brecht (2021)



# What can a perceptron learn?

Your turn

Three data sets for binary classification (two classes):

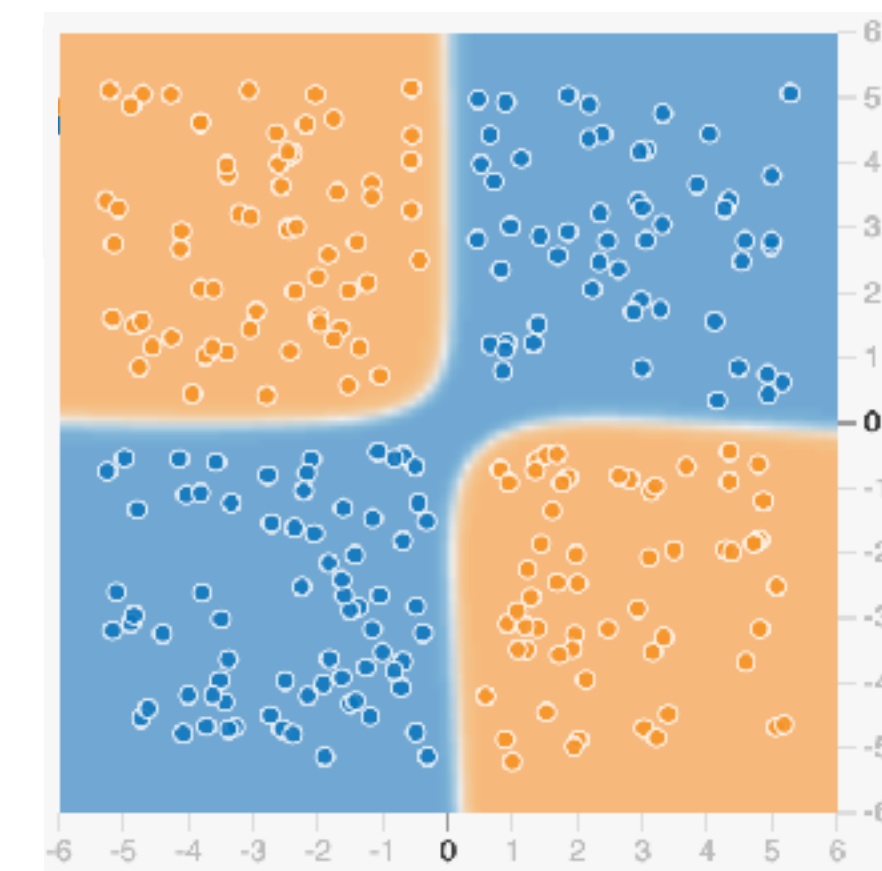
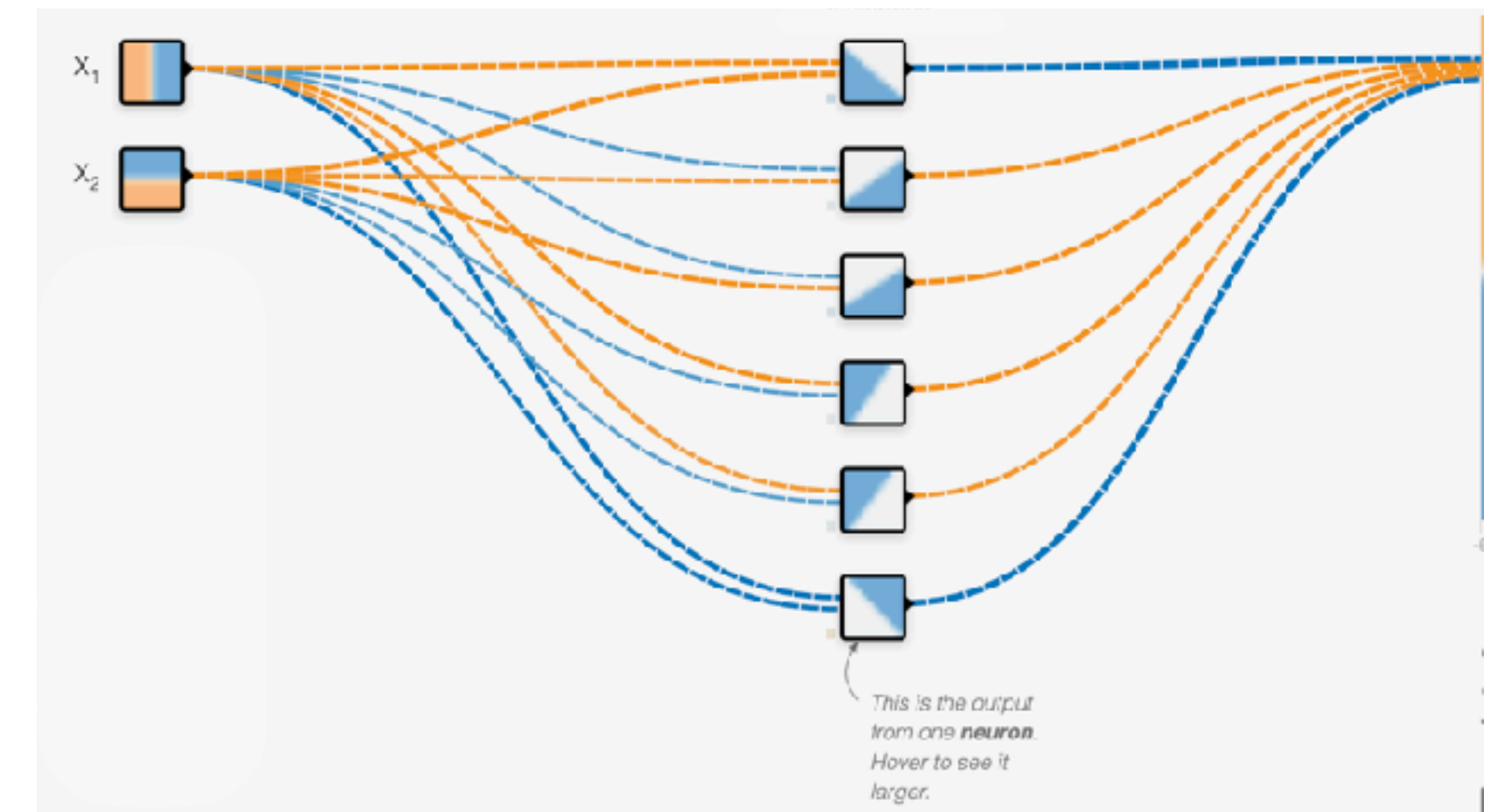


[playground.tensorflow.org](https://playground.tensorflow.org)

# Feature learning: research questions

## A summary

- Neural networks discover the right **features** directly from their data.
- What are those features? How are they related to the **data statistics**?
- How does the **dynamical** process of learning unwrap the statistics of inputs?
- Relevant for machine learning and **neuroscience**!





# Acknowledgements



Ezter  
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Data science



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Math



Riccardo  
Data science



Sara  
Neuroscience



SISSA data science group (Oct 2023)



[datascience.sissa.it](https://datascience.sissa.it)



@sebastiangoldt