

# Phase transitions in Artificial Intelligence

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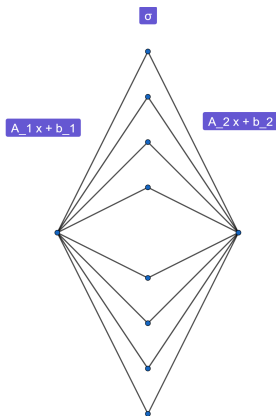
## Phase transitions (examples)

“Sudden changes of behavior”: When a relevant function has a **discontinuity** in one of its derivatives.

- ▶ Gas-liquid.
- ▶ Ferromagnetism.
- ▶ Message/disease propagating in a network (or graph).

Examples from several areas of science/engineering, including AI.

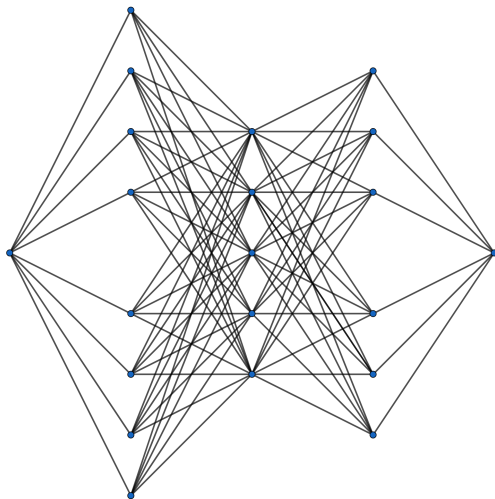
- Neural network  $\Rightarrow$  function  $f : \mathbb{R}^{N_{in}} \rightarrow \mathbb{R}^{N_{out}}$ , which to an input associates an output. Schematically, for one layer



- For  $b_1 \in \mathbb{R}^{N_{in}}$ ,  $b_2 \in \mathbb{R}^{N_{out}}$  and  $A_1 \in M_{N \times N_{in}}$ ,  $A_2 \in M_{N_{out} \times N}$

$$f(x) = A_2 \sigma(A_1 x + b_1) + b_2.$$

► Neural network  $\Rightarrow$  function  $f : \mathbb{R}^{N_{in}} \rightarrow \mathbb{R}^{N_{out}}$ . An example with three layers:



$$f(x) = A_4\sigma(A_3\sigma(A_2\sigma(A_1x + b_1) + b_2) + b_3) + b_4.$$

- ▶ Neural network  $\Rightarrow$  function  $f : \mathbb{R}^{N_{in}} \rightarrow \mathbb{R}^{N_{out}}$ . For one layer

$$f(x) = A_2 \sigma(A_1 x + b_1) + b_2,$$

with

$$\sigma : \mathbb{R}^N \rightarrow \mathbb{R}^N$$

a fixed function (activation function).

- ▶ The network can be taught pairs  $\{(x_i, y_i)\}_{i \in I}$ : Minimize

$$\mathcal{L}(\mathbf{A}, \mathbf{b}) := \sum_{i \in I} (f(x_i) - y_i)^2,$$

with  $\mathbf{A} = (A_1, A_2)$  and  $\mathbf{b} = (b_1, b_2)$ .

- ▶ Evolve  $\mathbf{A}(t), \mathbf{b}(t)$  with  $-\nabla \mathcal{L}$ . Then, the values  $f_t(x_i)$  evolve according to:<sup>1</sup>

$$\frac{d}{dt}(f_t(x_i) - y_i) = - \sum_{j \in I} \Theta_{ij}(f_t(x_j) - y_j),$$

for a matrix  $\Theta$  known as Neural Tangent Kernel (NTK).

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<sup>1</sup>When  $N \rightarrow +\infty$

## Spectral gap and spectral ratio

- ▶ From the equation

$$\frac{d}{dt}(f_t(X) - Y) = -\Theta(f_t(X) - Y),$$

we find

$$f_t(X) - Y = e^{-\Theta t}(f_0(X) - Y).$$

- ▶ If

$$\lambda_{min} := \text{smaller eigenvalue } \Theta,$$

we obtain

$$|f_t(X) - Y| \lesssim e^{-\lambda_{min} t}.$$

$$\implies \lambda_{min} \sim \text{learning rate!}$$

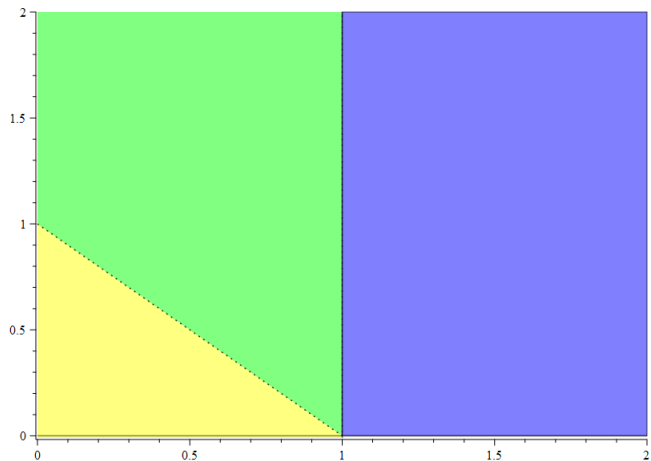
- ▶ If

$$\lambda_{max} := \text{largest eigenvalue } \Theta.$$

$$\implies \frac{\lambda_{max}}{\lambda_{min}} \sim \text{stability of learning.}$$

- ▶ Varying the network parameters ... phase transitions in  $\lambda_{min}$  e  $\frac{\lambda_{max}}{\lambda_{min}}$ .

## Phase diagram for $\text{tr } \Theta$ (L. Carvalho, J. Costa, J. Mourão, O.)



Phase diagram for the behavior of  $\text{tr } \Theta$  in wide and deep networks.

## To do:

- ▶ Find relevant quantities to study.
- ▶ Learn techniques which can be used to establish the existence of phase transitions (Crandall–Rabinowitz theorem, Lyapunov–Schmidt reduction).
- ▶ Prove the existence of phase transitions in the problem at hand.
- ▶ Understand what is the best phase (for our goals).



**Thank you!**