

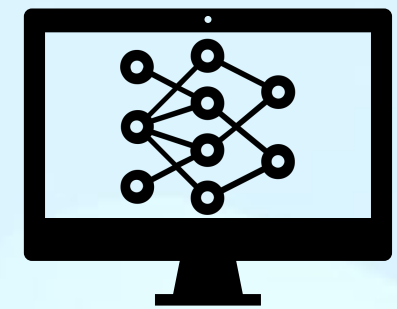
Brain-inspired AI

Neuromorphic computing and beyond

Steve Abreu @ Cover • April 18, 2023

Introduction

Motivation for brain-inspired AI



Sample efficiency

200M words²

2,000,000M words¹

10,000x

Online learning

Lifelong learning

Expensive re-training

Representations

Multimodal, Causal,
Hierarchical

Limited?

Energy efficiency

20 years of food

Train: 500 years of food³
20 questions: 500ml + 🍌

1) GPT-3, 2) Fermi estimate: 500 tokens/minute, 10 hours per week, 20 years of life, 3) from BLOOM, based on 2,000kcal/day

Introduction

Motivation against brain-inspired AI



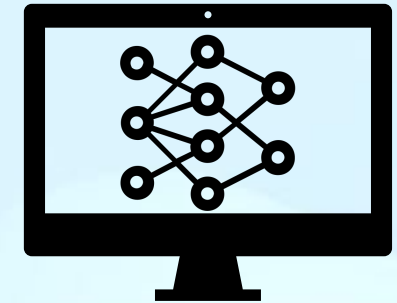
Constant sugar supply

Temperature sensitive & fragile

Slow ($\sim 10\text{ms/op}$)

Forgets things

Emotional



Electricity

Robust to environment

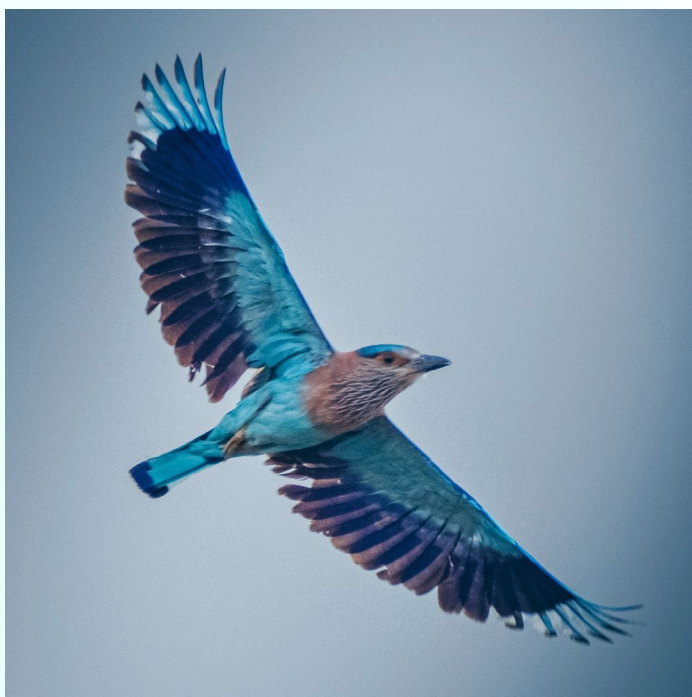
Fast ($\sim 1\text{ns}$)

Non-volatile memory

No emotions (*yet?*)

Lesson #1

Deciding *how* to “take inspiration” is difficult



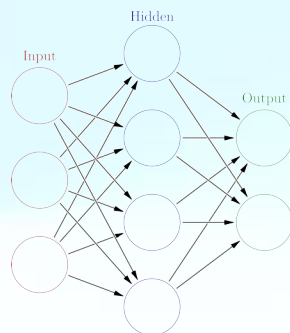
Deep learning

Example of brain-inspired AI

A LOGICAL CALCULUS OF THE
IDEAS IMMANENT IN NERVOUS ACTIVITY

WARREN S. MCCULLOCH AND WALTER PITTS

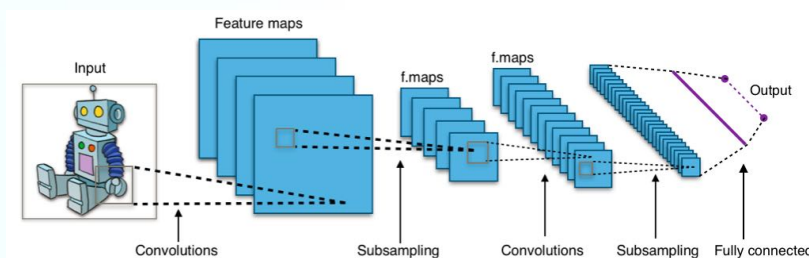
1943



RECEPTIVE FIELDS OF SINGLE NEURONES IN
THE CAT'S STRIATE CORTEX

1959

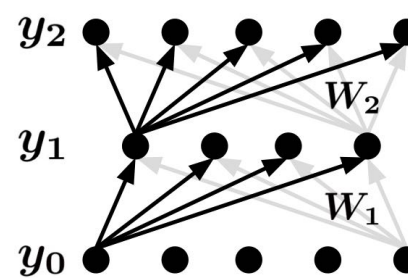
BY D. H. HUBEL* AND T. N. WIESEL*



Learning not brain-inspired

A

Forward Pass



Output: $y_l = f(h_l)$
Transfer Function: $f(h)$
Activation: $h_l = W_l y_{l-1}$

B

Error Computation

Loss Function: L

$$\frac{\partial L}{\partial W_i} = \delta_i \times y_{i-1}$$

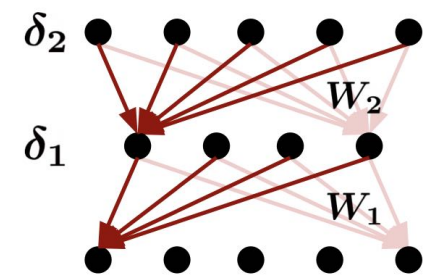
Examining loss at W_1 :

$$\frac{\partial L}{\partial W_1} = \left(\delta_2 \frac{\partial h_2}{\partial y_1} \frac{\partial y_1}{\partial h_1} \right) \times y_0$$

W_2

C

Backward (Error) Pass



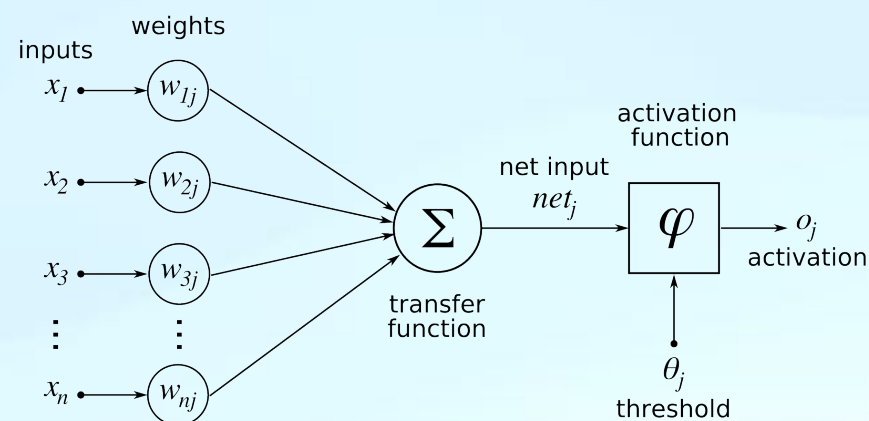
The Weight Transport Problem
Weights of the forward network need to be copied for the backward (error) pass. How can this “copying” be achieved biologically?

Transformers and LLMs?



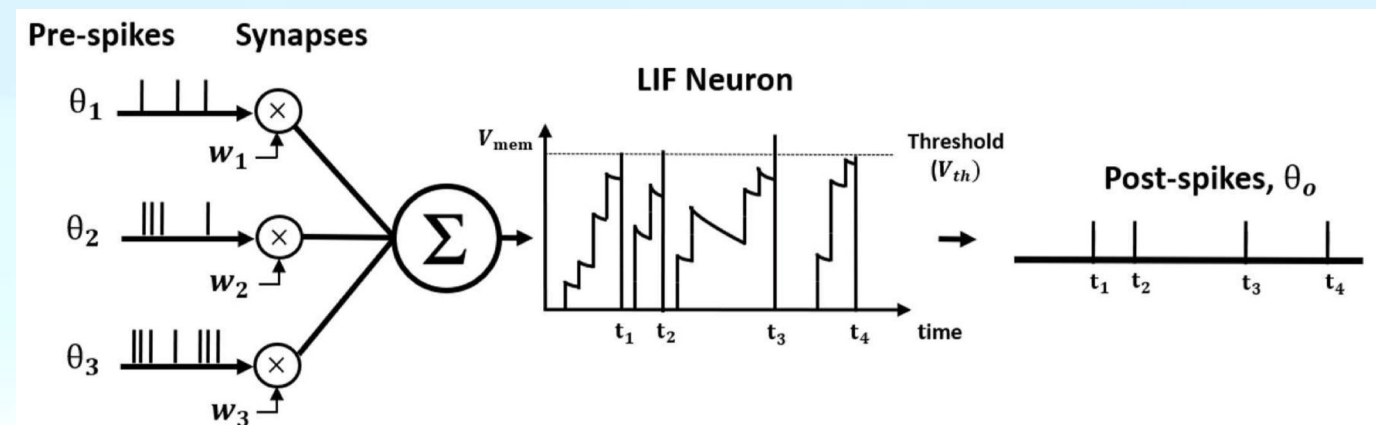
Neuromorphic computing

Taking brain-inspired AI to the next level

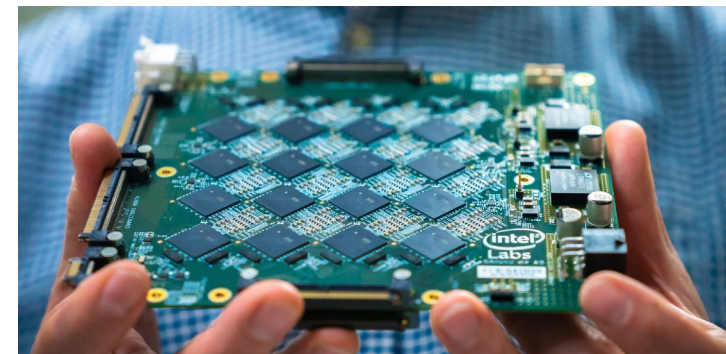
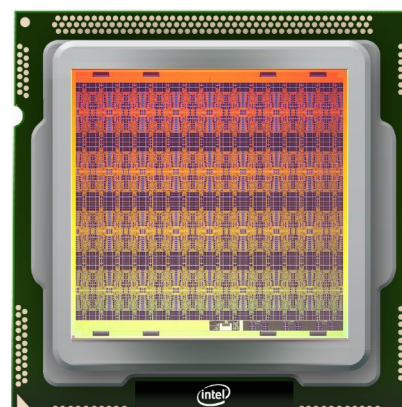
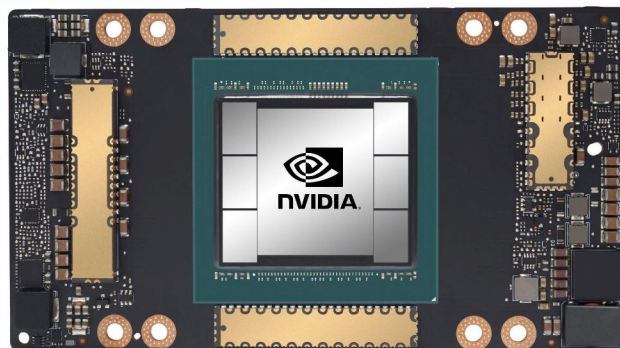


$$\vec{x} \quad W\vec{x} \quad f(W\vec{x})$$

Artificial neuron



Spiking neuron



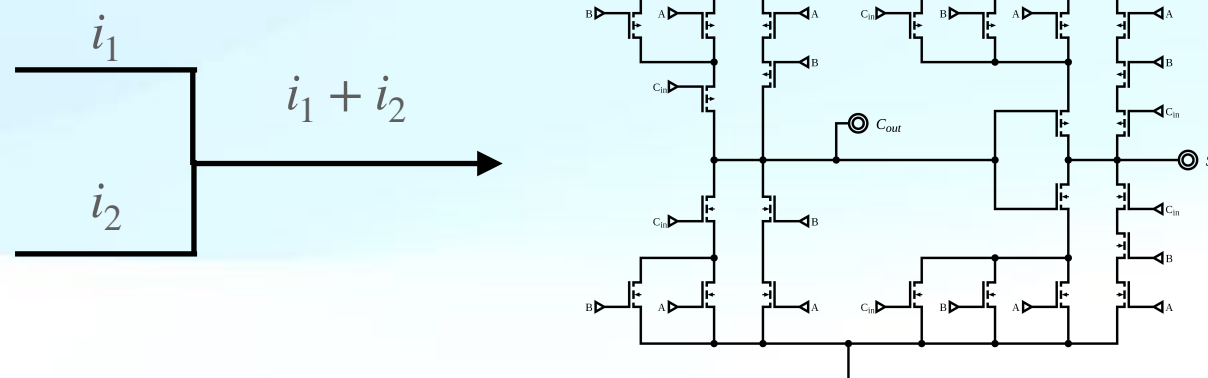
Neuromorphic computing

Principle #1: mixed-signal (analog/digital) computation

Computation

Analog

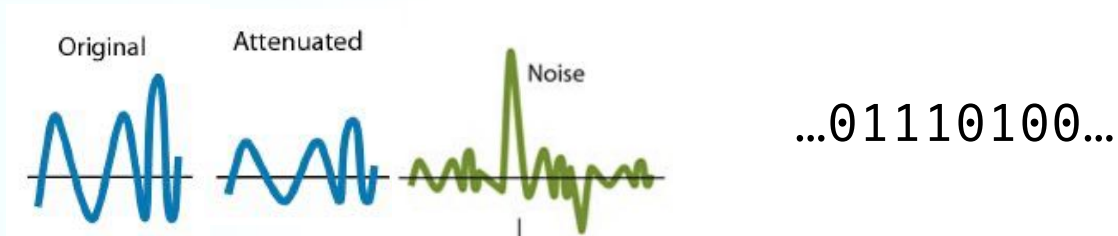
Digital



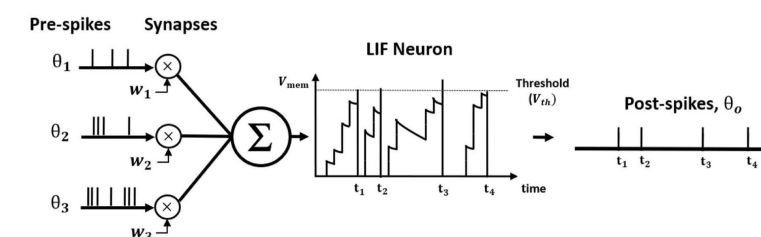
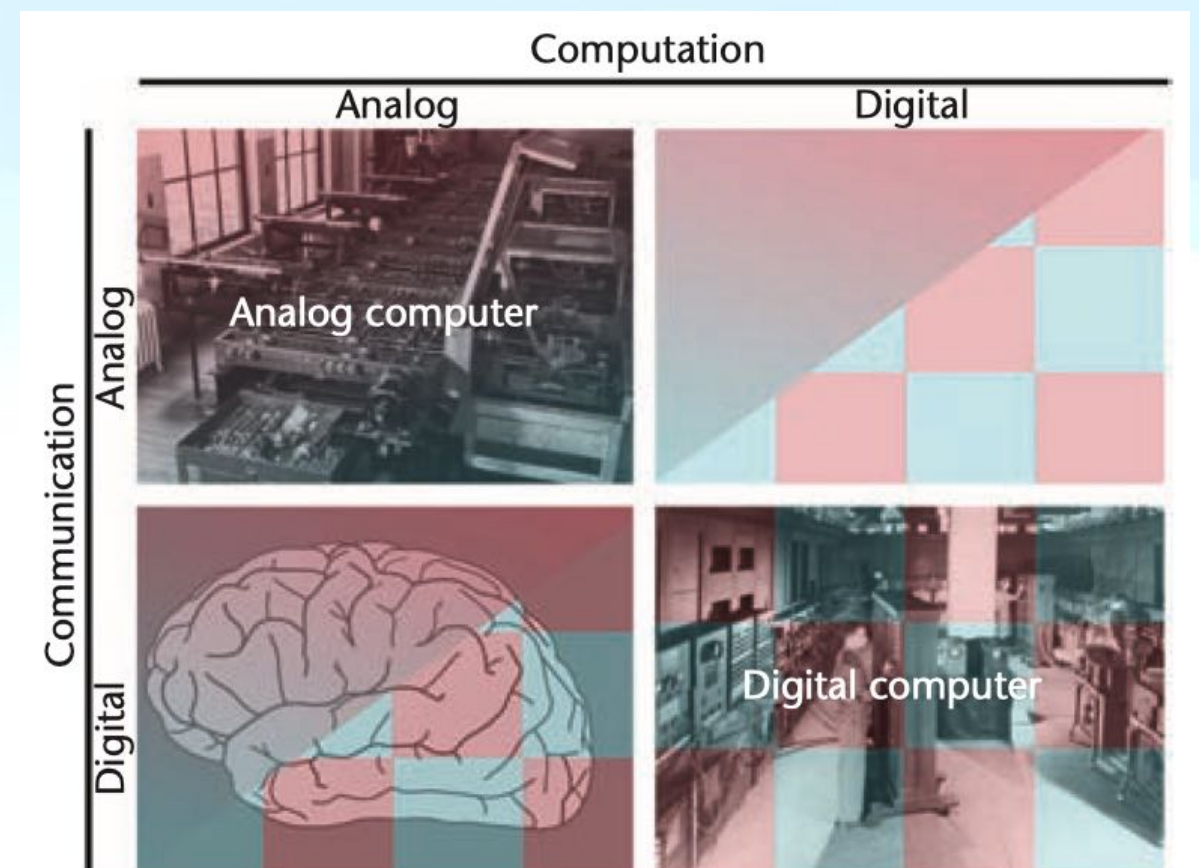
Communication

Analog

Digital



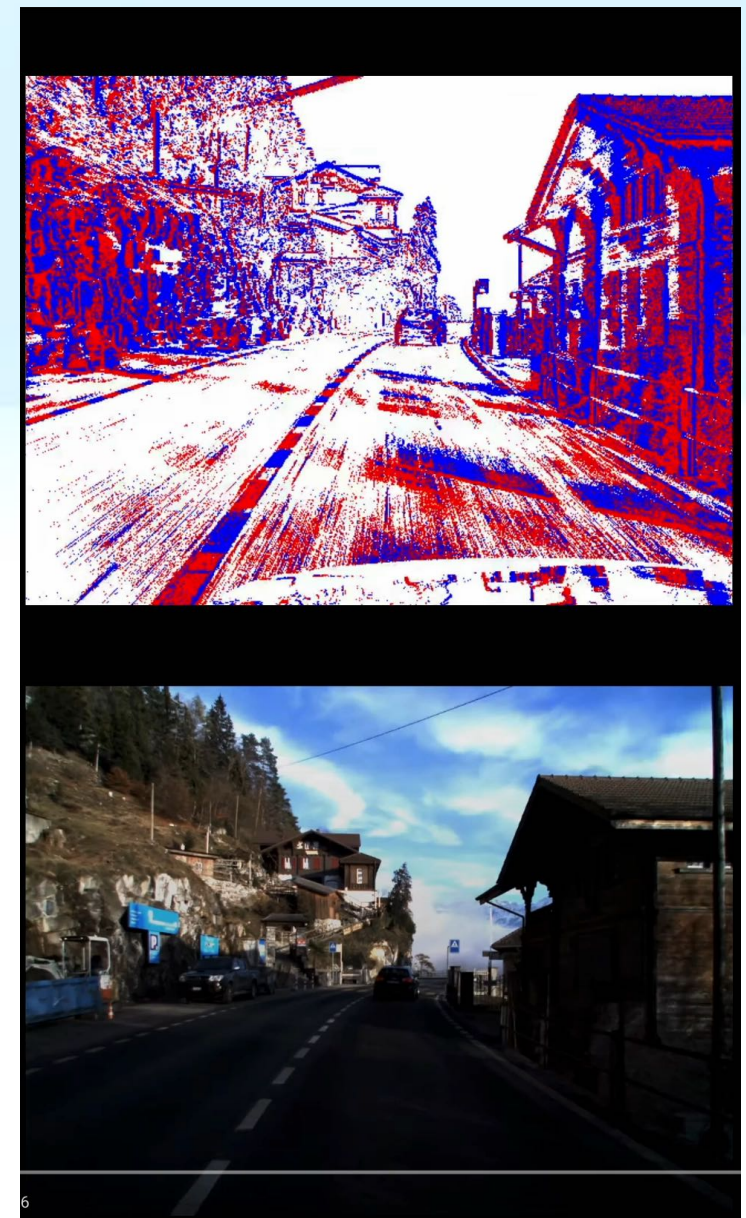
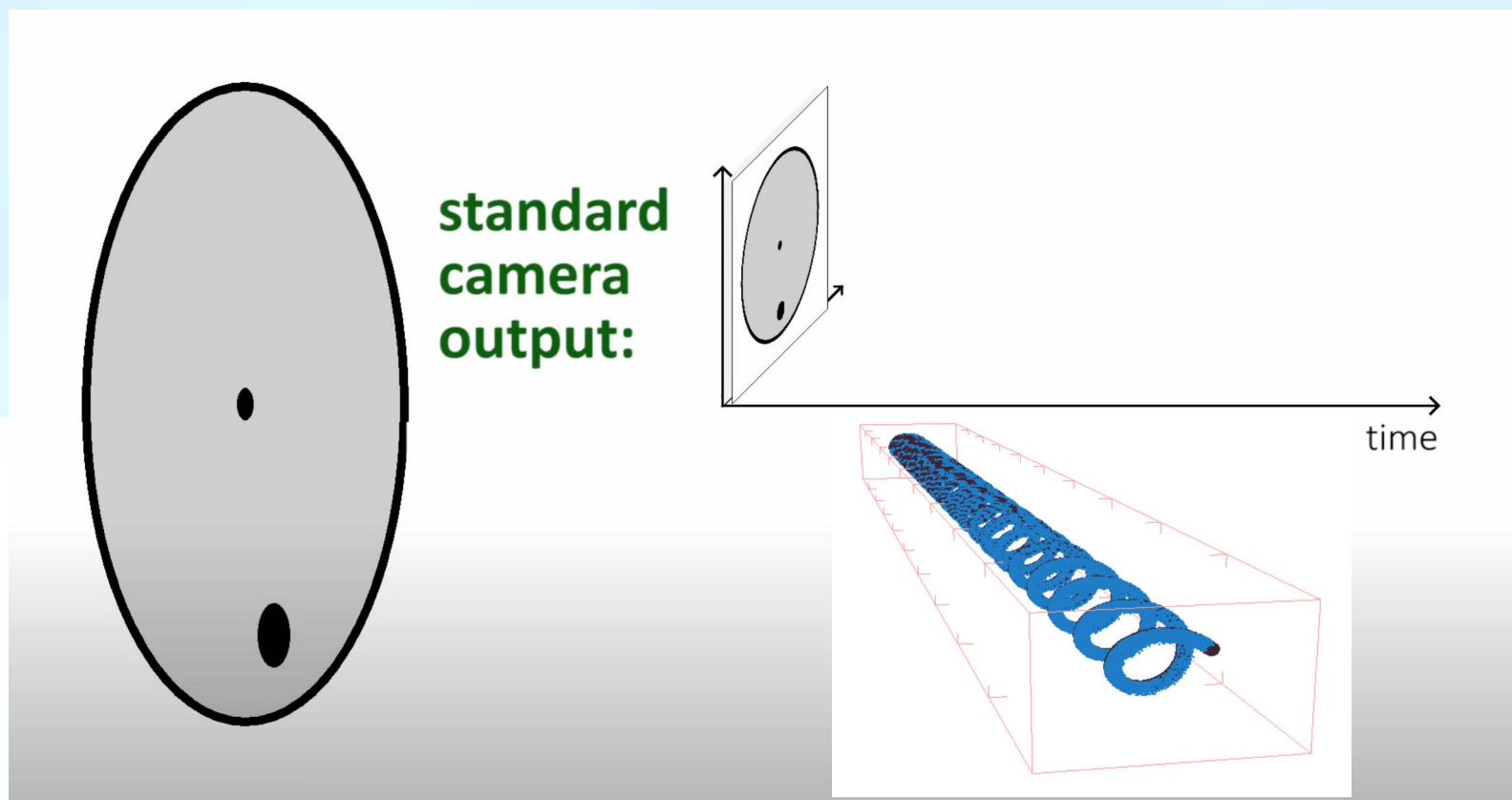
Algorithms and hardware!



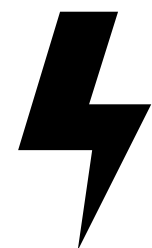
Neuromorphic computing

Principle #2: event-based sensing and processing

<https://youtu.be/W4yW78y4F7A?t=4>

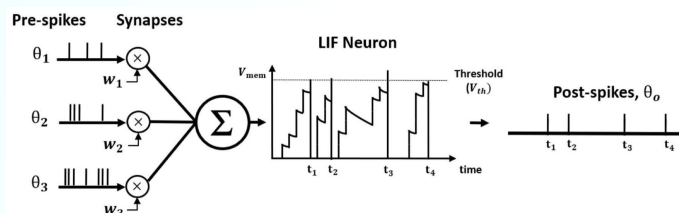


Sparse event stream
~ 1 million fps



Difficult to process
on CPUs/GPUs!

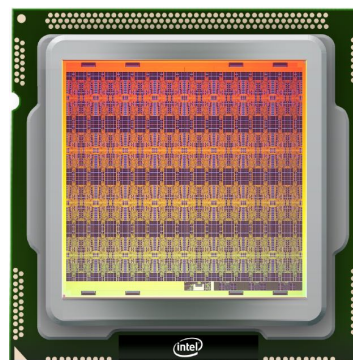
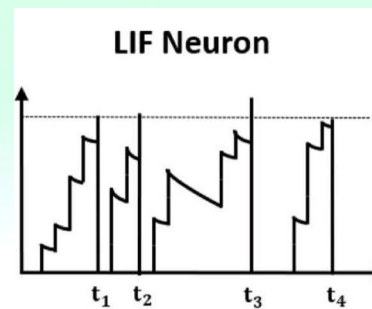
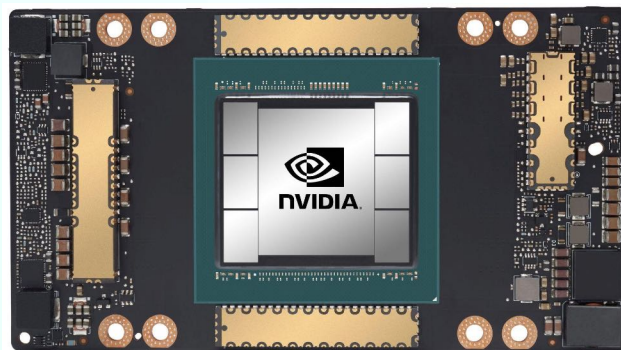
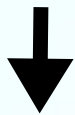
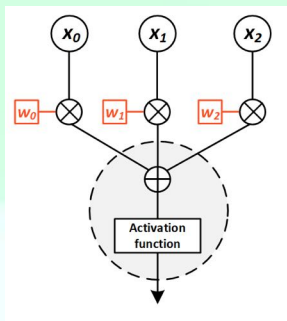
→ neuromorphic chip!



Lesson #2 (a meta-principle)

Computation is physical → hardware matters!

- Hardware and algorithms must be *aligned*



?

Beyond neuromorphic (and GPUs)

A universe of possible computers to be explored

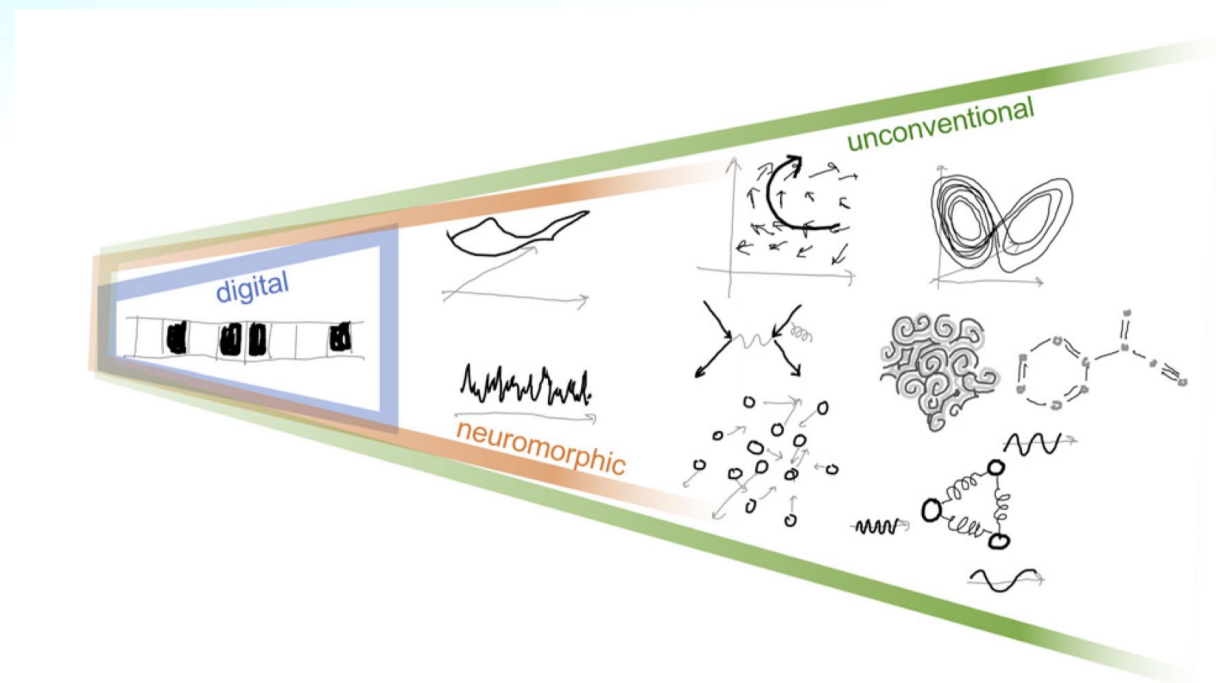
- There is a whole *universe* of alternative computer hardware

*“Current AI is what works
on GPUs”*

The Hardware Lottery

Sara Hooker (Google)

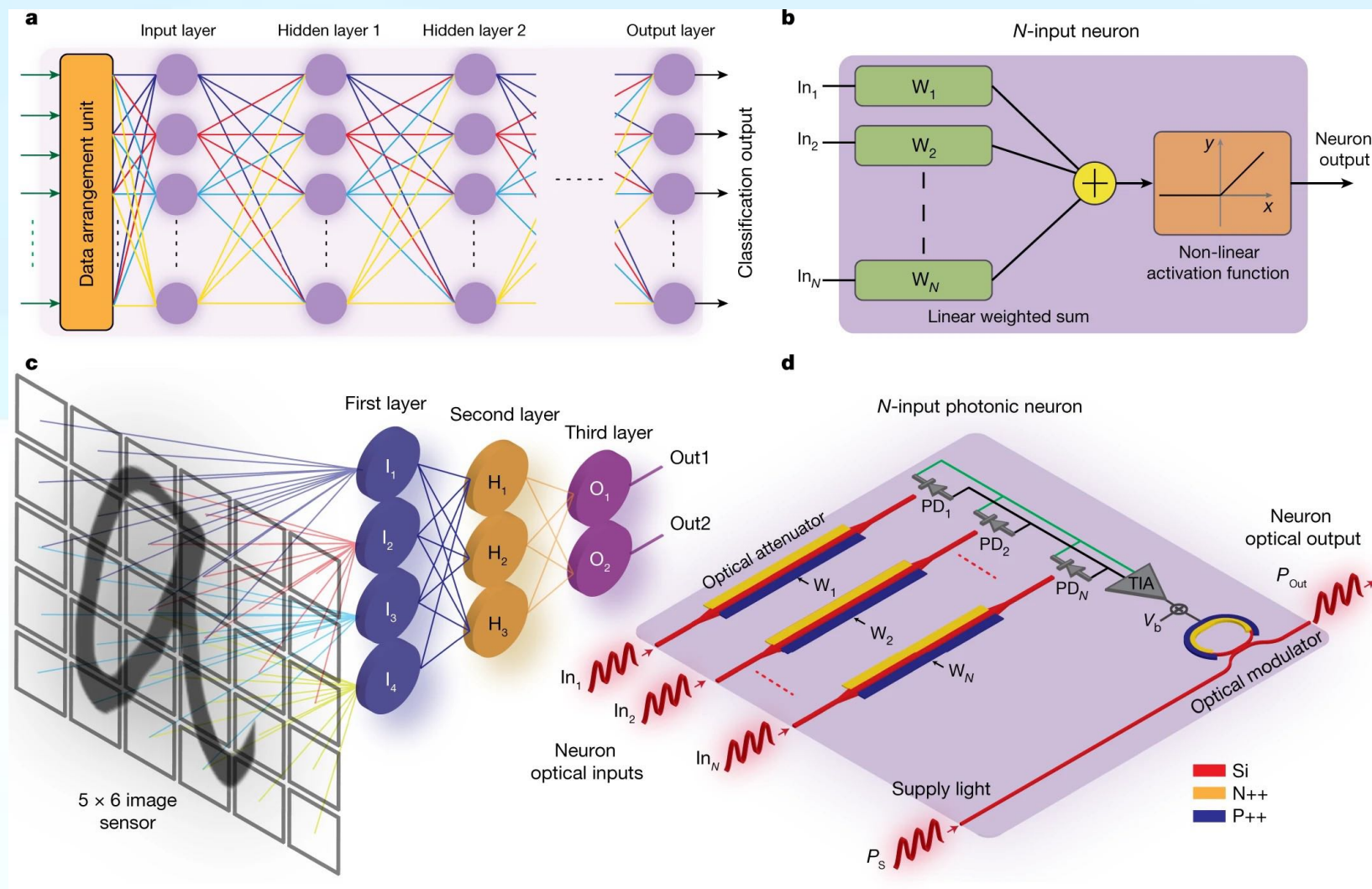
2019



Herbert Jaeger, 2021. *Towards a generalized theory comprising digital, neuromorphic and unconventional computing*

Beyond neuromorphic (and GPUs)

Photonic computing



Ashtiani et al, 2022. *An on-chip photonic deep neural network for image classification*. Nature.

Moving into mainstream AI:

TPU v4: An **Optically** Reconfigurable...

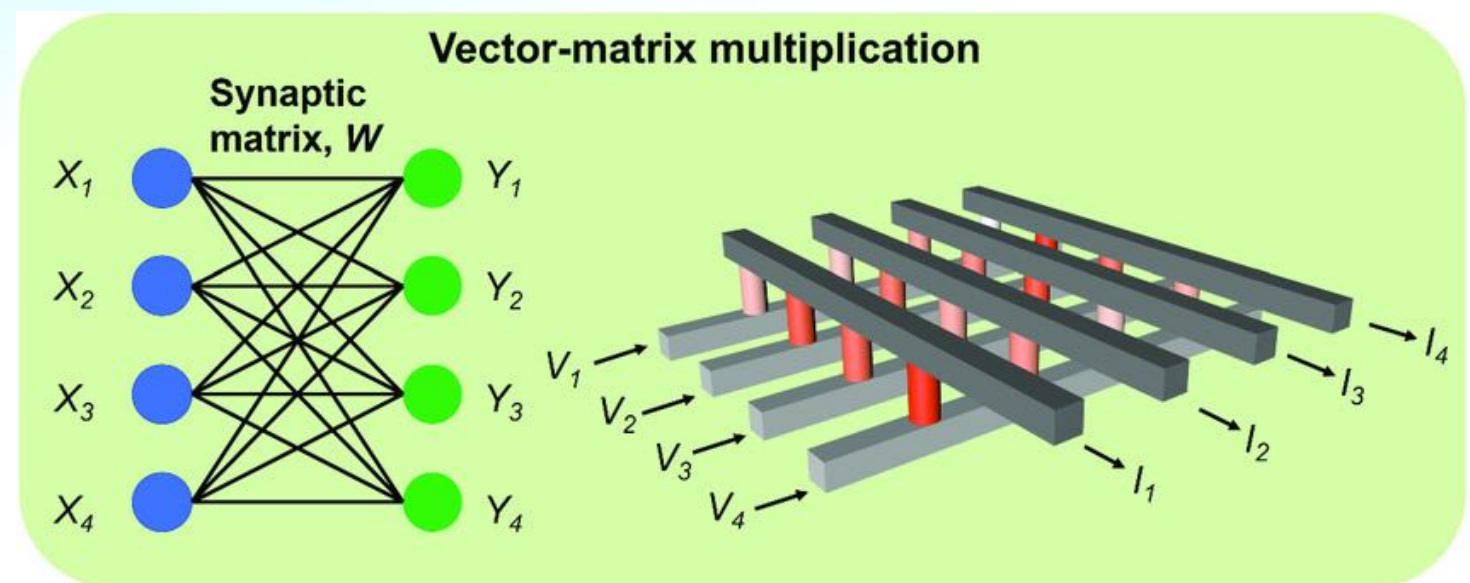
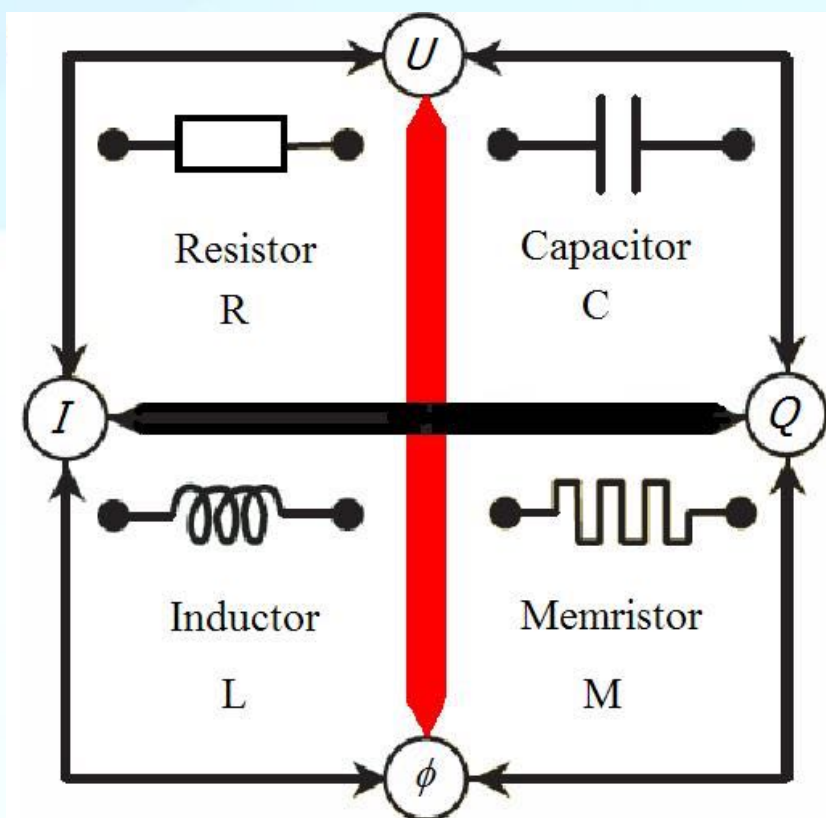
arXiv:2304.01433, April 2023

+ much faster
+ less energy

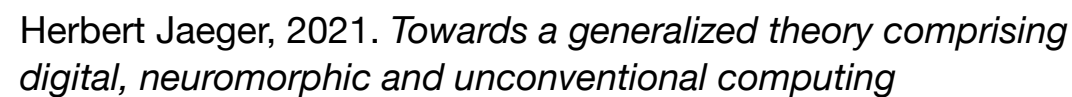
Beyond neuromorphic (and GPUs)

Cognitive materials: memristors & co

memristors act like synapses



An explosion of hardware diversity in AI



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thank you



MINDS

ai.rug.nl/minds/

slides on stevenabreu.com



rijksuniversiteit
 groningen



Image sources

- Slide 4 left: https://en.wikipedia.org/wiki/Artificial_neural_network
- Slide 4 right: https://en.wikipedia.org/wiki/Convolutional_neural_network
- Slide 6 top: https://fr.wikipedia.org/wiki/Ader_Avion_III#/media/Fichier:Ader_Avion_III_001.tif
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- Slide 7 right: <http://dx.doi.org/10.3389/fnins.2020.00119>
- Slide 7 bottom: <https://techxplore.com/news/2022-06-neuromorphic-architecture-deep-neural-networks.html>
- Slide 10 left: <https://www.goodtherapy.org/blog/psychpedia/neuroplasticity>
- Slide 10 right: <https://arxiv.org/pdf/2003.03988.pdf>
- Slide 10 bottom: <https://techxplore.com/news/2022-06-neuromorphic-architecture-deep-neural-networks.html>
- Slide 12: TODO
- Brain <https://unsplash.com/photos/58Z17lnVS4U>
- Loihi chip <https://qbi.uq.edu.au/brains-chip-neuromorphic-computing>
- GPU <https://developer.nvidia.com/blog/nvidia-ampere-architecture-in-depth/>
- memristive crossbar array <https://doi.org/10.1002/aisy.202000149>
- memristor https://www.science20.com/alpha_meme/detractors_not_grasping_the_memristor-93011