

Solid State Neuroscience

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why do Neuroscience in a Solid State Physics Lab??

An analogy

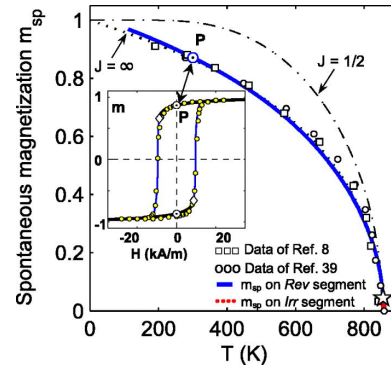
Material



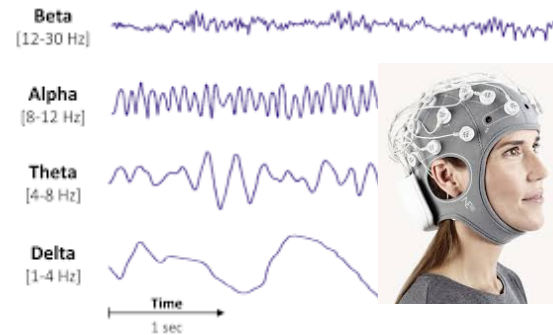
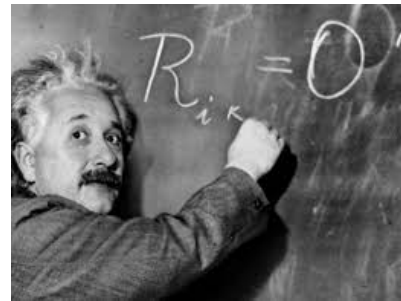
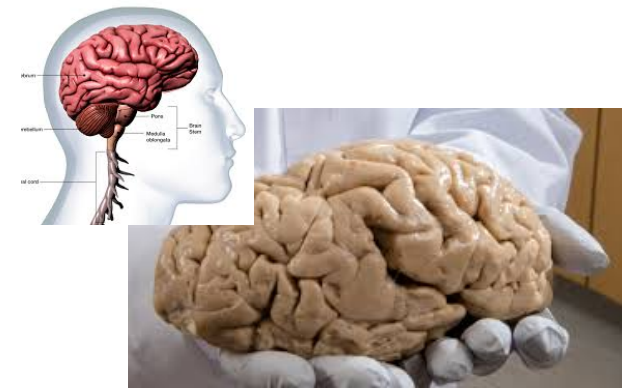
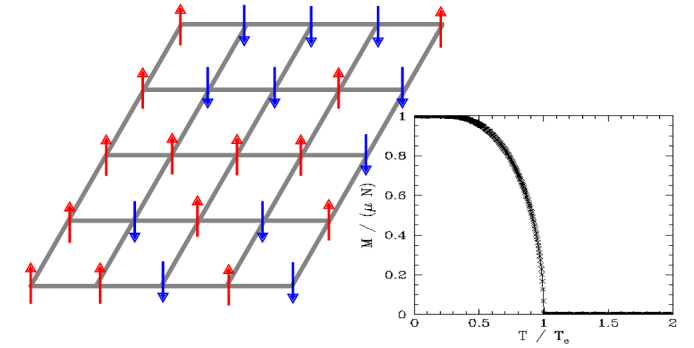
Functionality



State of matter

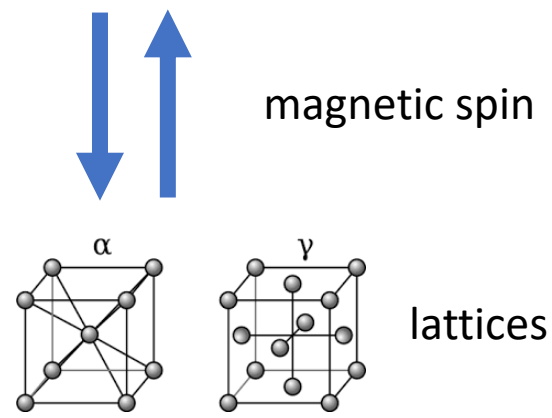
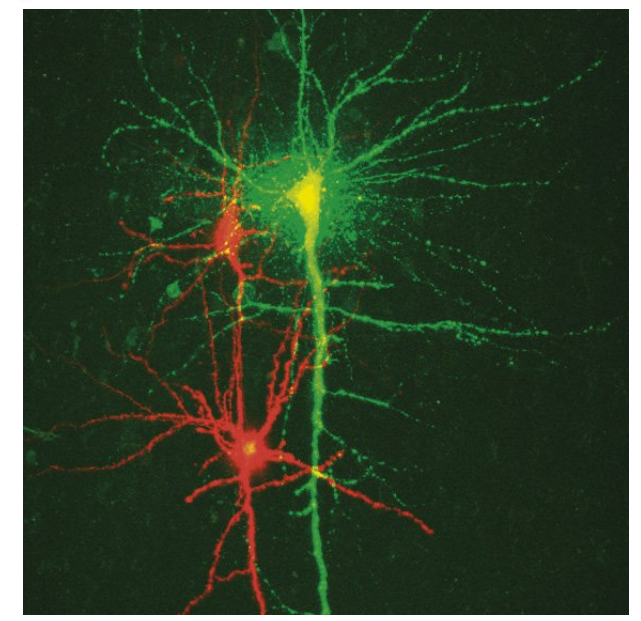
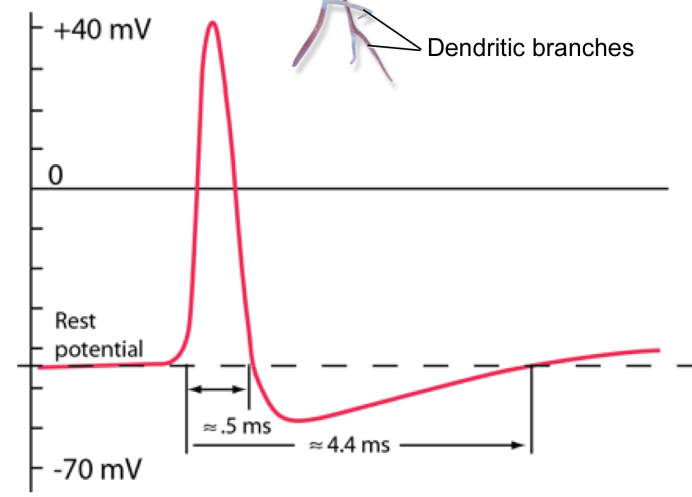
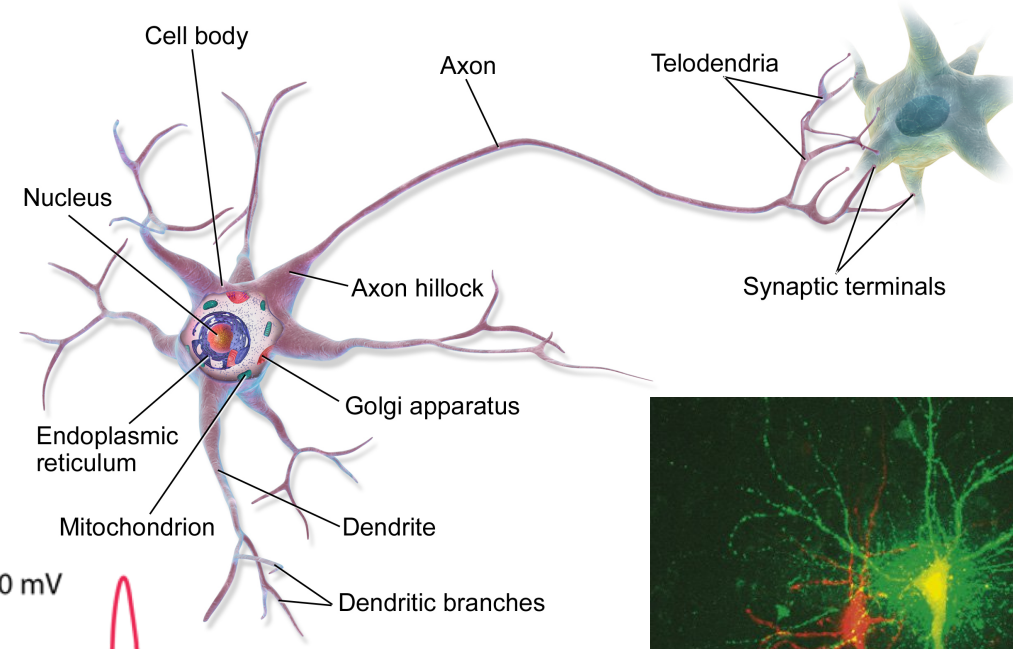


Modeling



?

Degrees of freedom?
 Interactions?
 Networks?



Neural networks

Mathematical models

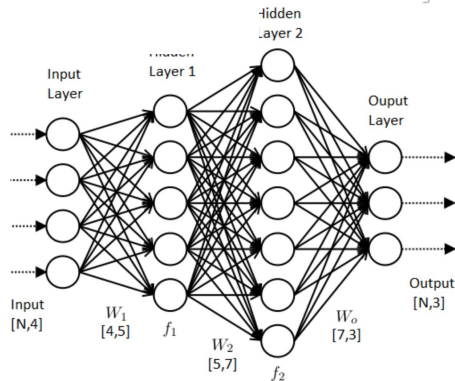
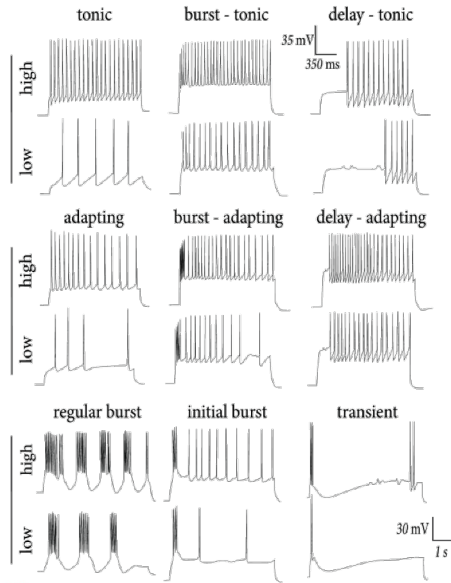
Differential equations that are discretized

$$C_m \frac{dV}{dt} = -g_L(V - E_L) + g_L \Delta_T e^{\frac{V - V_T}{\Delta_T}} - u + I$$

$$\tau_w \frac{du}{dt} = a(V - E_L) - u$$

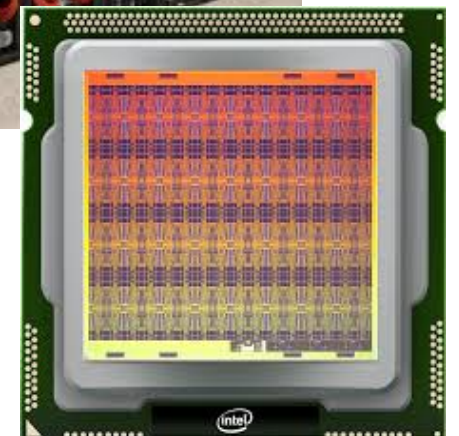
$$\text{if } V > V_{peak} = \begin{cases} V \rightarrow V_r \\ u \rightarrow u + b \end{cases}$$

- C_m = membrane capacitance (pF)
- E_L = leak reversal potential (mV)
- g_L = leak conductance (nS)
- V_T = spike threshold (mV)
- V_r = resting potential (mV)
- τ_w = adaptation time constant (ms)
- Δ_T = spike slope factor (mV)
- a = subthreshold adaptation (nS)
- b = spike triggered adaptation (pA)
- I = total current (pA)



Neuromorphic chips

IBM *TrueNorth*, Intel *Loihi*, FPGAs
 Conventional computers optimized to run the mathematical models



AI driven

Long-time simulations
 Mathematical artifacts?

Is there an alternative route?
 Make neurons directly in hardware!
New neuromorphic functionalities

Neuromorphic materials

resistive switching & *memristors*

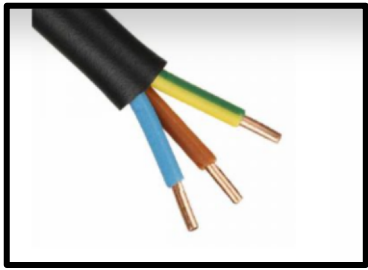
memory + resistors

sudden change in resistance under applied electric field (voltage)

hysteresis in the I - V

Electric breakdown in **Mott insulators**:
volatile effect

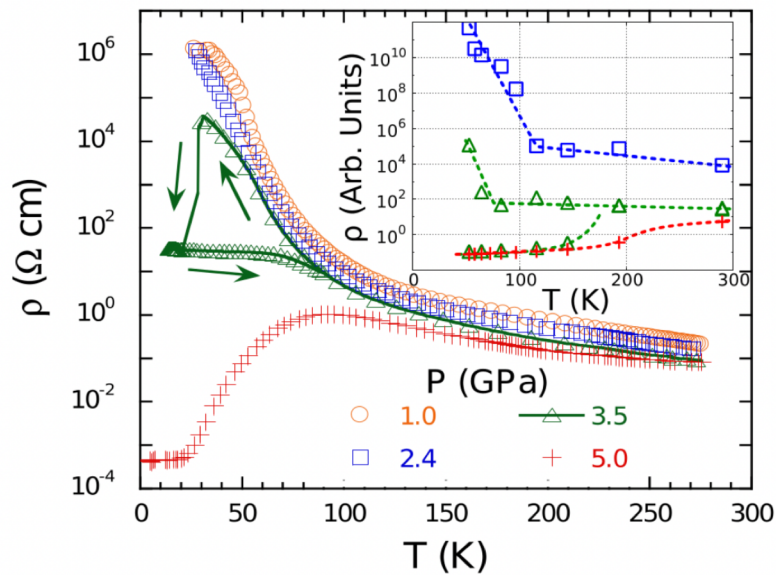
a new (and unexpected) neuromorphic functionality



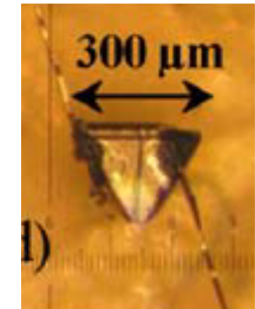
Ordinary matter:
cable made of metal (copper) and insulator (plastic)

Mott Materials (quantum matter) Should be metals.... But are insulators

insulator to metal (1st order) transition
10 orders of magnitude!!



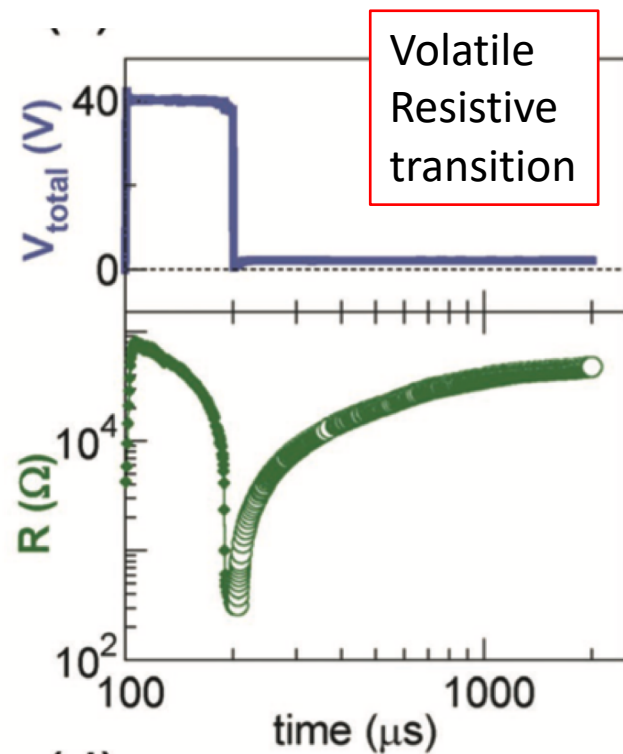
- Strongly Correlated Material
- Famous High T_c Superconductors



Acha, Camjayi @ UBA
Phys Rev Lett (2014)

Metastability of the
1st order MIT (DMFT)

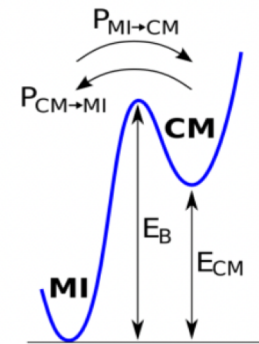
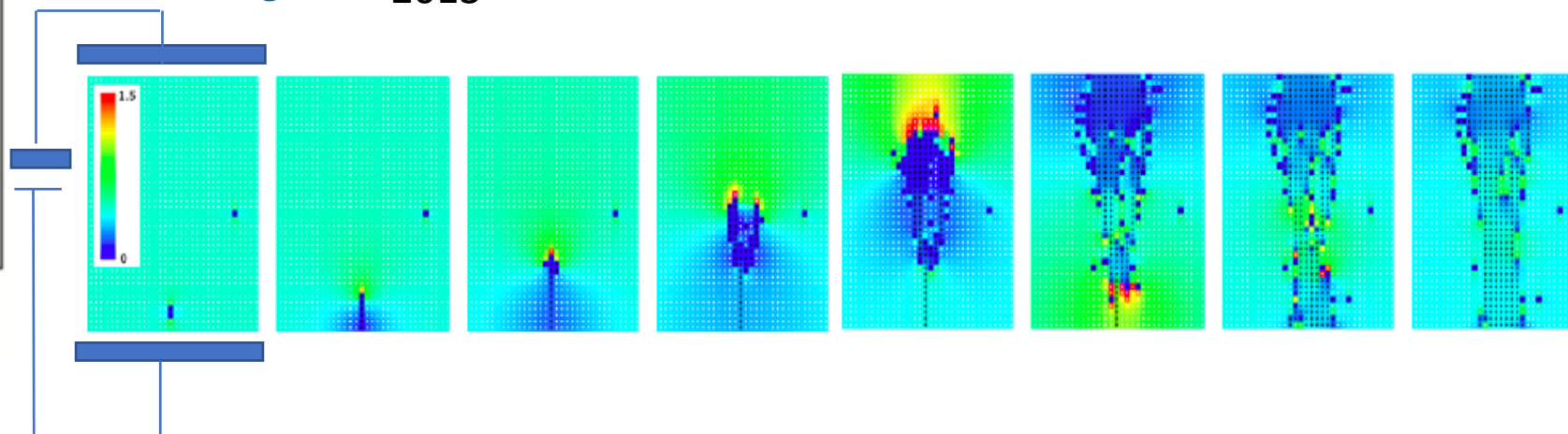
Surprise: Electrically driven Mott transition



ADVANCED MATERIALS

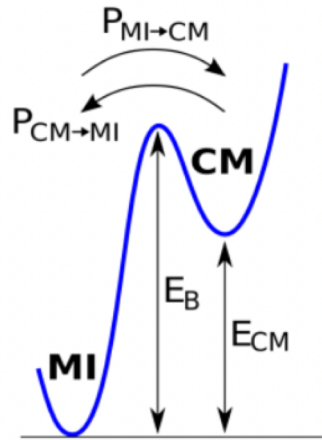
Universal Electric-Field-Driven Resistive Transition in Narrow-Gap Mott Insulators

Pablo Stoliar✉, Laurent Cario, Etienne Janod, Benoit Corraze, Catherine Guillot-Deudon, Sabrina Salmon-Bourmand, Vincent Guiot, Julien Tranchant, Marcelo Rozenberg✉ 2013

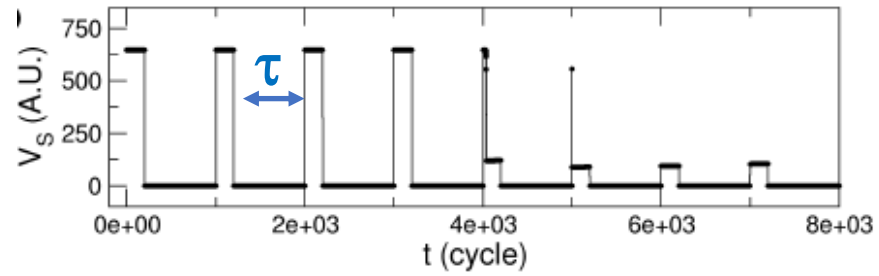


How about a non trivial validation?

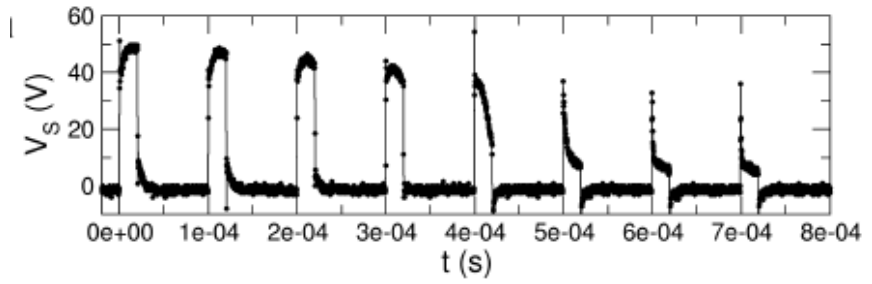
Metastability of the 1st order MIT (DMFT)



Model prediction



Experiment

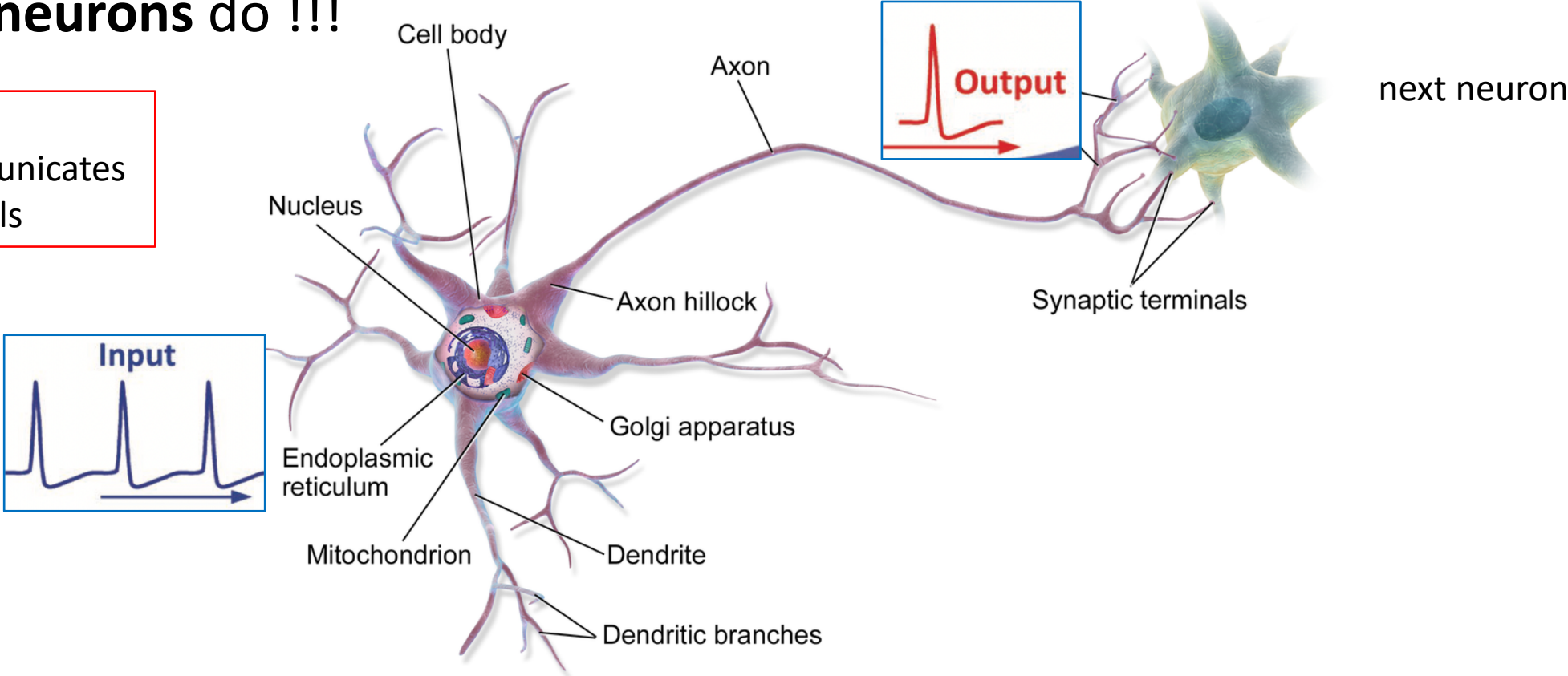


Electric pulses provoke a collapse of resistance, then a surge of current!

1st Aha! Moment !!!!

it is like what **neurons** do !!!

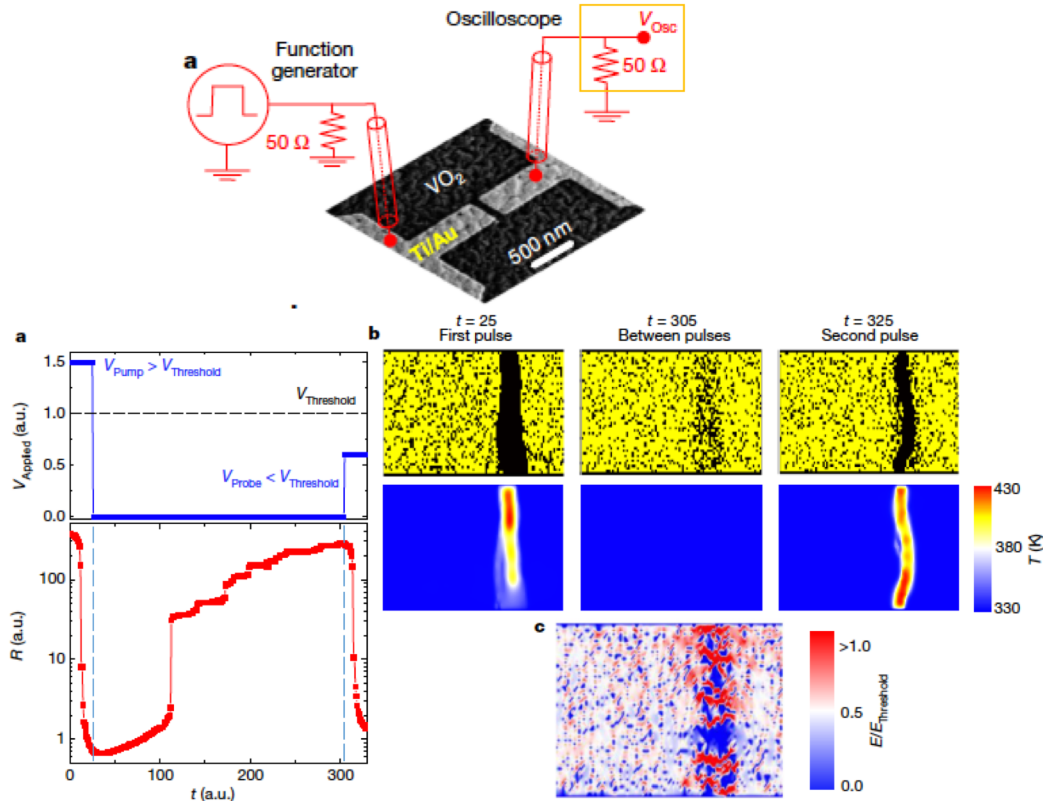
A neuron is a cell
Generates and communicates
through electric signals



Towards physical artificial neurons

Mott materials (VO_2) are promising

But also difficult to make and control, and we are beginning to understand (filamentary conduction)



Del Valle et al, Nature 2019

Del Valle, Ramirez, MR, Schuller JAP 2018 (review)

Kalcheim et al Nat Comm 2020

Del Valle et al, Science 2021

R.Rocco et al PRApp (2022)

Editor's Suggestion

Silicon neurons are a reality

Everything can be done with conventional electronics
But *silicon neurons* are not that simple

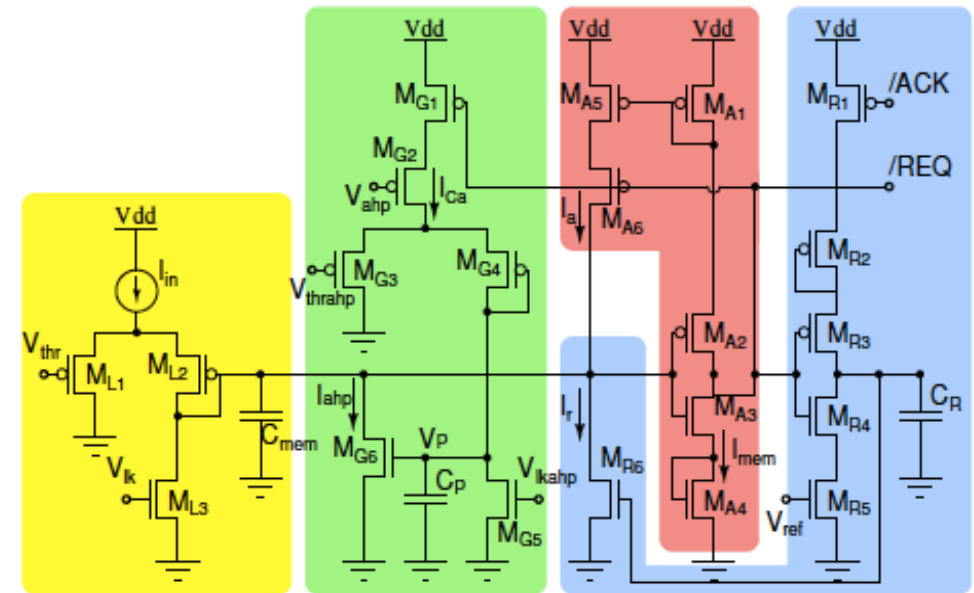


Fig. 2: Adaptive exponential I&F neuron circuit schematic.

Indiveri, Stefanini, Chicca IEEE (2010)

Indiveri et al Front Neurosci 2011 (review)

Can't we have the "best of both worlds"?

Simplicity of Mott memristors

Reliability of conventional silicon electronics

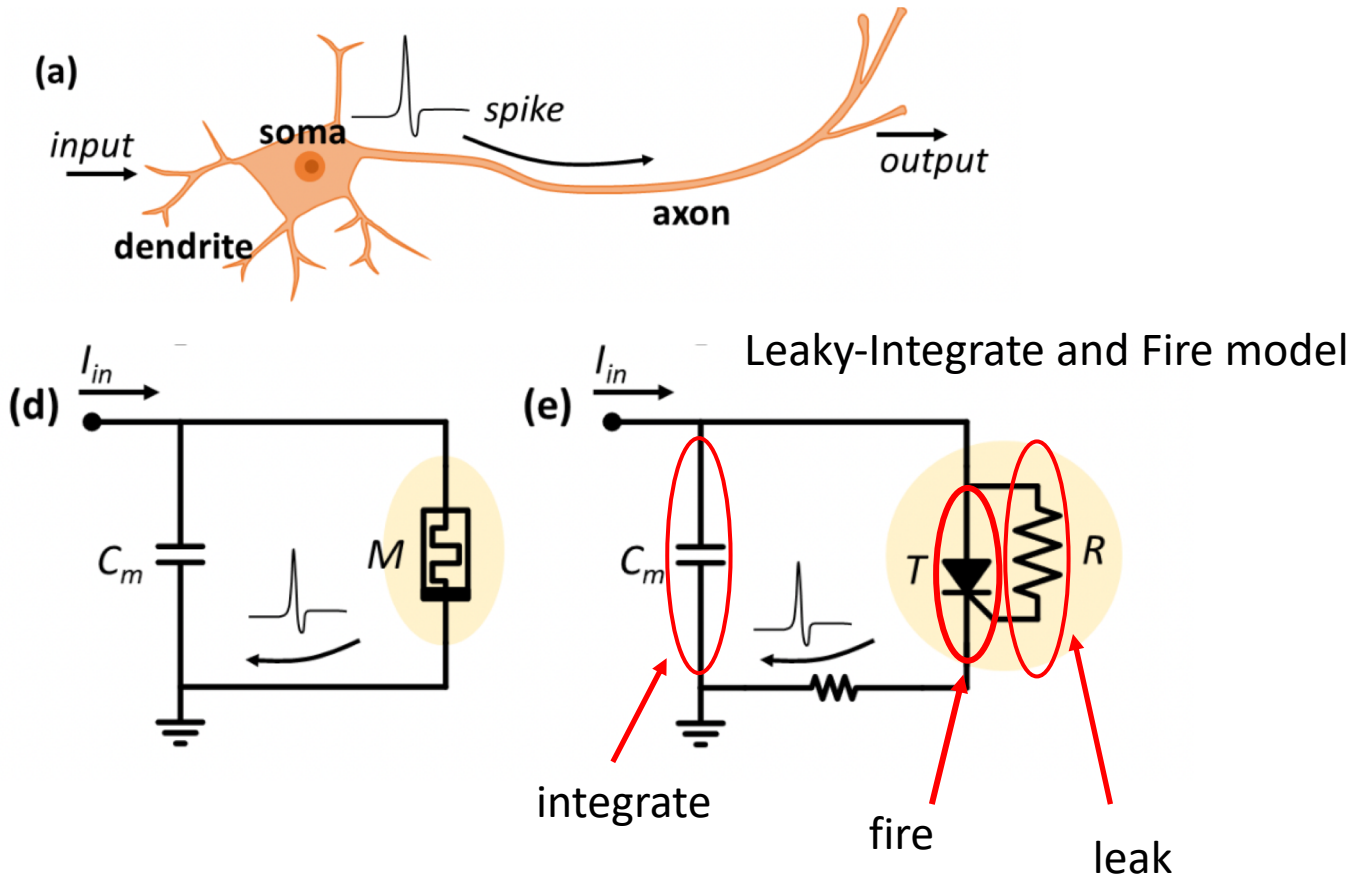
Best of both worlds: Memristive Silicon Spiking Neuron

aka Ultra-Compact Neuron

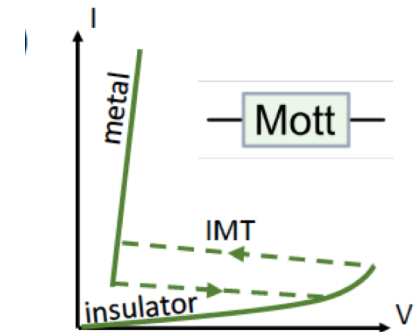
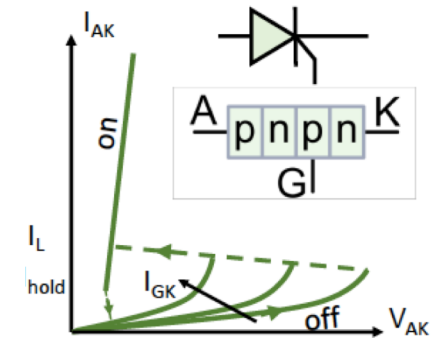
MR, O. Schneegans and P. Stolar, Sci Rep (2019)

2nd Aha! Moment !!!

memory-resistive effect \leftrightarrow hysteresis

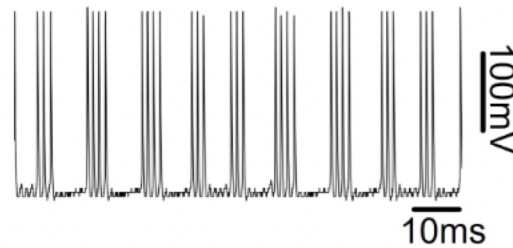
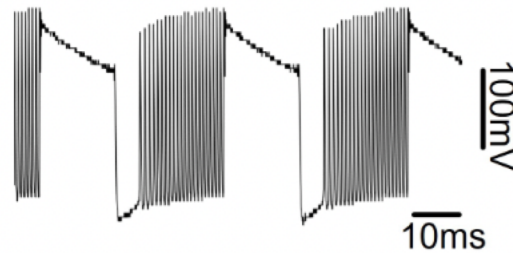
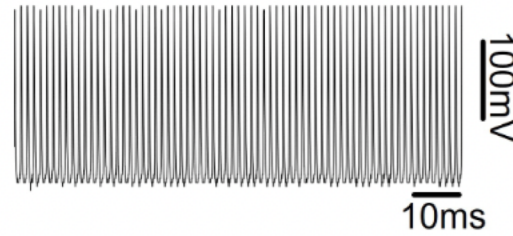
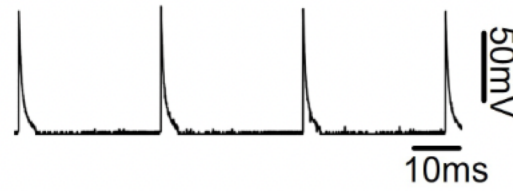
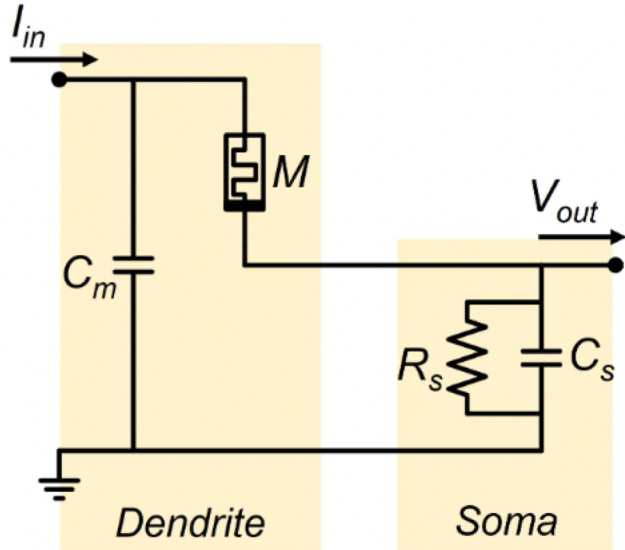
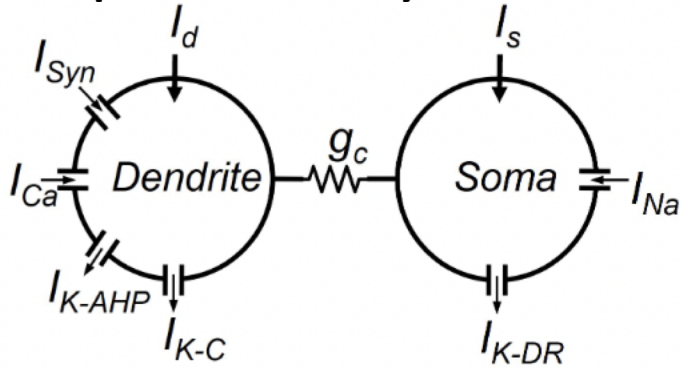


Thyristor
Threshold control diode



Memristive Bursting neuron

2-compartment Pinsky-Rinzel model



$$\begin{aligned}
 C_m V_s' &= -I_{\text{Leak}}(V_s) - I_{\text{Na}}(V_s, h) - I_{\text{K-DR}}(V_s, n) \\
 &\quad + (g_c/p)(V_d - V_s) + I_s/p \\
 C_m V_d' &= -I_{\text{Leak}}(V_d) - I_{\text{Ca}}(V_d, s) - I_{\text{K-AHP}}(V_d, q) \\
 &\quad - I_{\text{K-C}}(V_d, Ca, c) - I_{\text{Syn}}/(1-p) \\
 &\quad + (g_c/(1-p))(V_s - V_d) \\
 &\quad + I_d/(1-p)
 \end{aligned} \tag{1}$$

The kinetic equation for each of the gating variables h, n, s, c and q takes the form

$$y' = (y_\infty(U) - y)/\tau_y(U). \tag{2}$$

$$Ca' = -0.13I_{\text{Ca}} - 0.075Ca$$

$$I_{\text{Leak}}(V_s) = \bar{g}_L(V_s - V_L)$$

$$I_{\text{Leak}}(V_d) = \bar{g}_L(V_d - V_L)$$

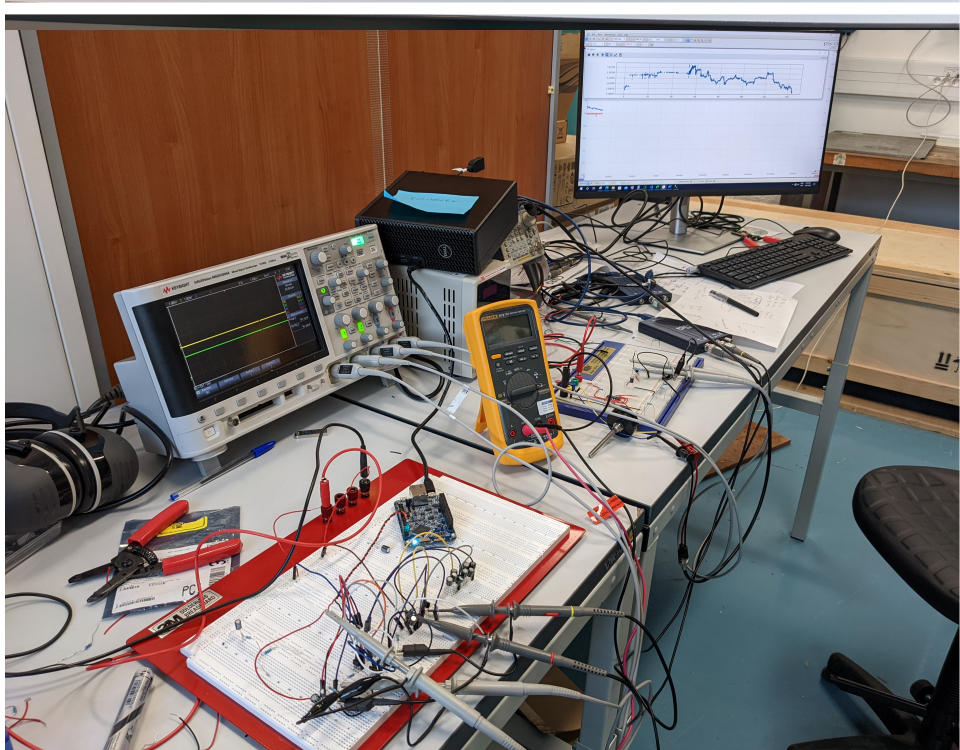
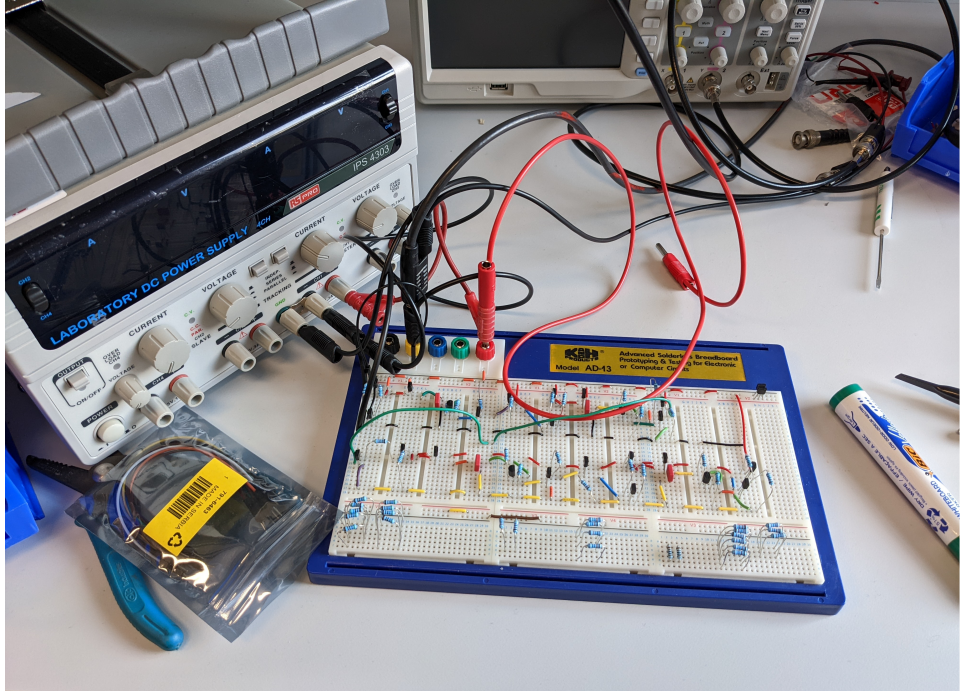
$$I_{\text{Na}} = \bar{g}_{\text{Na}} m_\infty^2(V_s) h (V_s - V_{\text{Na}})$$

$$I_{\text{K-DR}} = \bar{g}_{\text{K-DR}} n (V_s - V_{\text{K}})$$

$$I_{\text{Ca}} = \bar{g}_{\text{Ca}} s^2 (V_d - V_{\text{Ca}})$$

$$I_{\text{K-C}} = \bar{g}_{\text{K-C}} c \chi(Ca) (V_d - V_{\text{K}})$$

$$I_{\text{K-AHP}} = \bar{g}_{\text{K-AHP}} q (V_d - V_{\text{K}})$$



New Experimental Activity at LPS
Solid State Neuroscience Lab
Thanks to LPS for initial support!

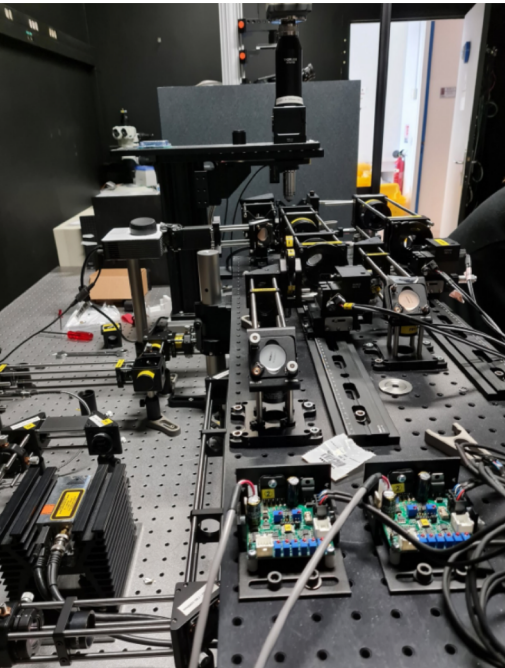
New Project:

**Brain-Machine-Interface:
memristive neural network
zebra fish neural network
w/G. Sumbre @ ENS**

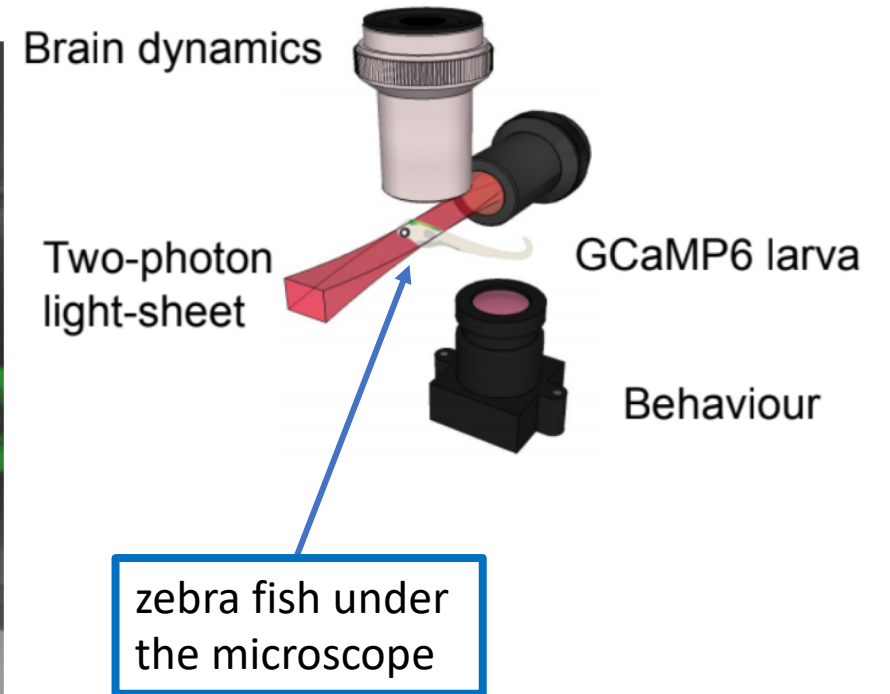
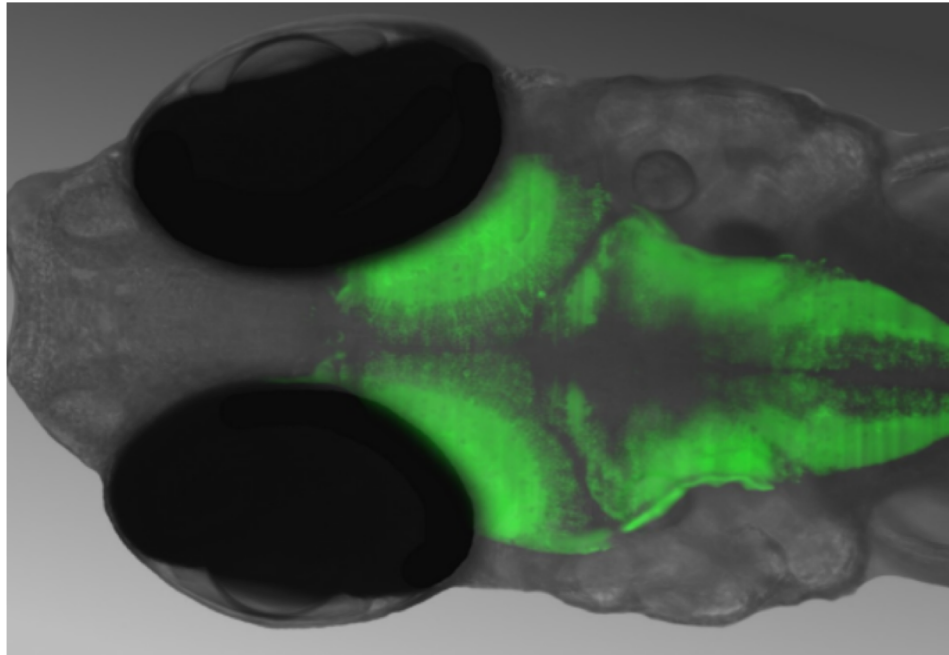
- Zebra fish larva optical recording and stimulation

Study fundamental questions of Neuroscience:

- Neural code? Spike timing versus spike rates
- State of matter? Criticality, chaos, phase diagram
- Dynamics? Waves, bumps, spike avalanches, epilepsy?
- Information How different networks communicate?
- Dynamical patterns, working memory



Brain of the zebra fish



"What I cannot create, I do not understand" (R. Feynman)



What I cannot create,
I do not understand.

Know how to solve every
problem that has been solved

Why const \times sort. P
TO LEARN:
Bethe Ansatz Probs.
Kondo
2-D Hall
local Temp
Non linear Classical Hydro

① $f = U(r, a)$
 $g = -4(r \cdot z) u(r, z)$
② $f = 2|k \cdot a| u(a)$

- Richard Feynman, February 1988 -

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Thank you