

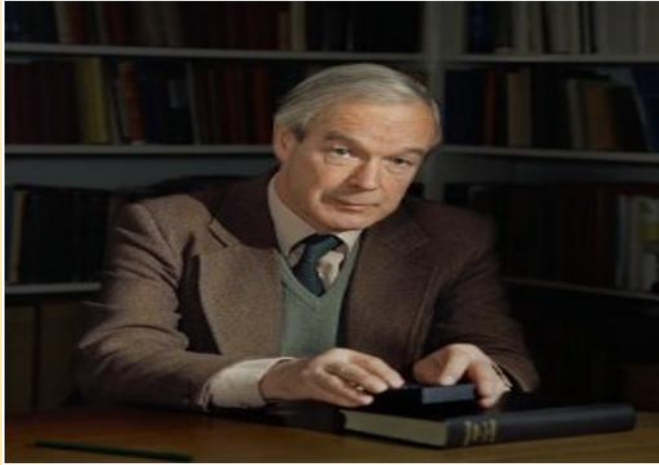
Hodgkin–Huxley ,

Memristor ,

and

Edge of Chaos

Journal of Physiology, Vol 117, pp. 500-544, (1952)



Sir A. L. Hodgkin



Sir A. F. Huxley

1961 Nobel Prize
in Physiology

A QUANTITATIVE DESCRIPTION OF MEMBRANE CURRENT AND ITS APPLICATION TO CONDUCTION AND EXCITATION IN NERVE*

■ A. L. HODGKIN and A. F. HUXLEY
Physiological Laboratory,
University of Cambridge,
Cambridge, U.K.

This article concludes a series of papers concerned with the flow of electric current through the surface membrane of a giant nerve fibre (Hodgkin *et al.*, 1952, *J. Physiol.* **116**, 424-448; Hodgkin and Huxley, 1952, *J. Physiol.* **116**, 449-566). Its general object is to discuss the results of the preceding papers (Section 1), to put them into mathematical form (Section 2) and to show that they will account for conduction and excitation in quantitative terms (Sections 3-6).

1. Discussion of Experimental Results. The results described in the preceding papers suggest that the electrical behaviour of the membrane may be represented by the network shown in Fig. 1. Current can be carried through the

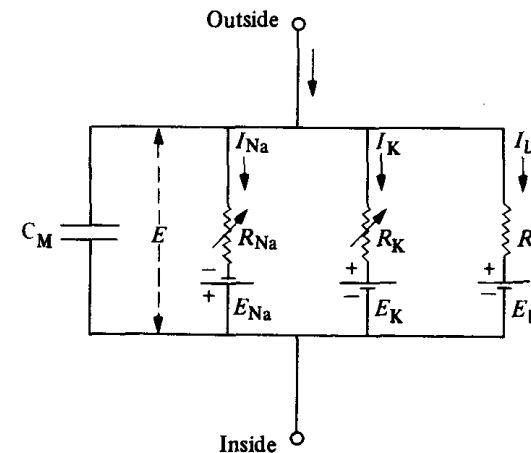


Figure 1. Electrical circuit representing membrane. $R_{Na}=1/g_{Na}$; $R_K=1/g_K$; $R_l=1/\bar{g}_l$. R_{Na} and R_K vary with time and membrane potential; the other components are constant.

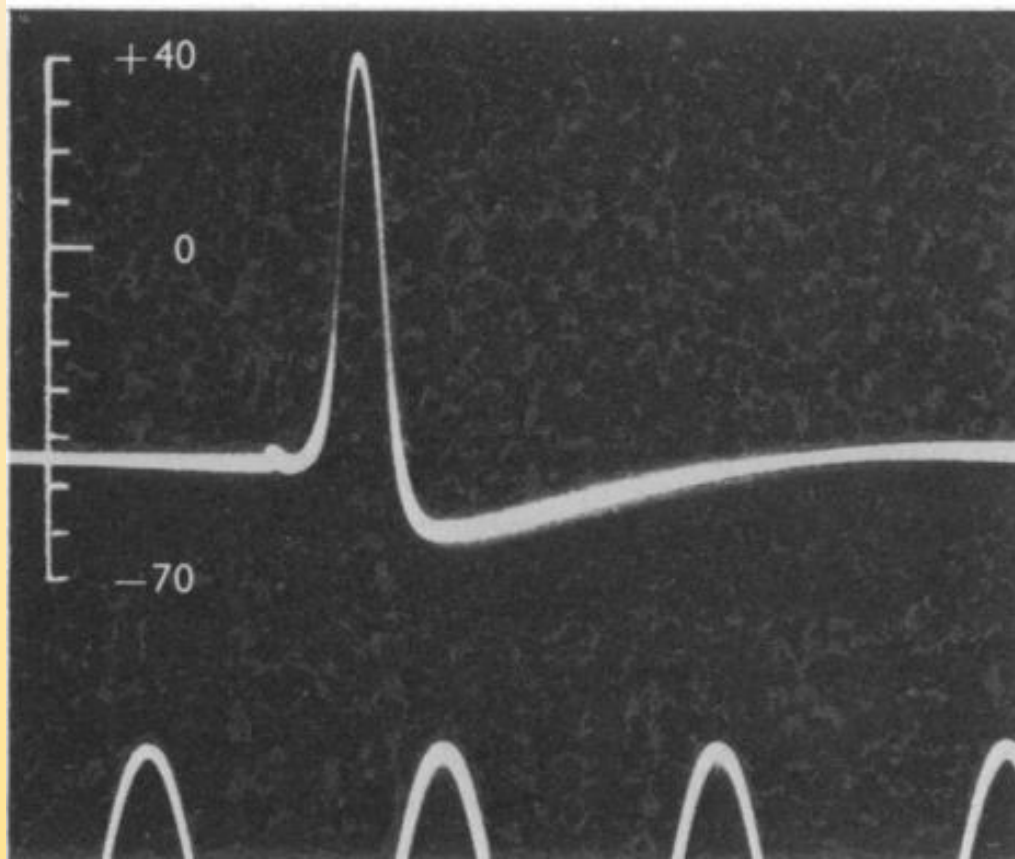
Hodgkin-Huxley Equations

$$\frac{dV}{dt} = -\frac{1}{C_M} \left[\bar{g}_K n^4 (V - E_K) + \bar{g}_{Na} m^3 h (V - E_{Na}) + \bar{g}_l (V - E_l) - I \right]$$

$$\frac{dn}{dt} = \left[\frac{0.01 (V + 10)}{\left(\exp \frac{V + 10}{10} - 1 \right)} (1 - n) \right] - \left[0.125 \exp \left(\frac{V}{80} \right) n \right]$$

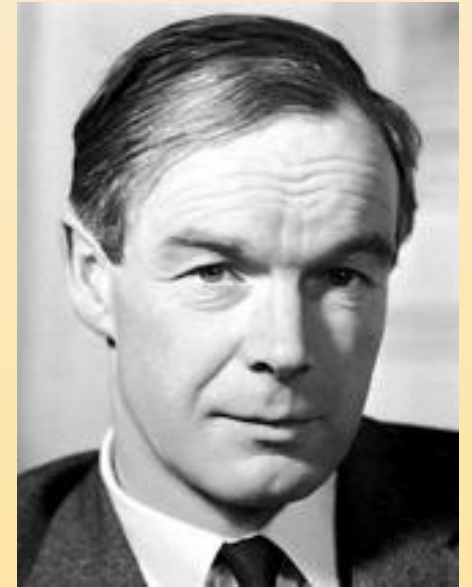
$$\frac{dm}{dt} = \left[\frac{0.1 (V + 25)}{\left(\exp \frac{V + 25}{10} - 1 \right)} (1 - m) \right] - \left[4 \exp \left(\frac{V}{18} \right) m \right]$$

$$\frac{dh}{dt} = \left[0.07 \exp \left(\frac{V}{20} \right) (1 - h) \right] - \left[\frac{1}{\left(\exp \frac{V + 30}{10} + 1 \right)} h \right]$$

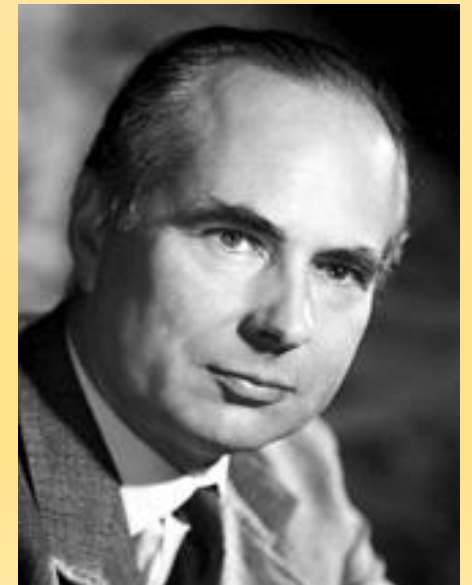


A. L. Hodgkin and A. F. Huxley

Action Potentials Recorded from inside the Nerve Fibre,
Nature, Lond. 144, 710-711, 1939

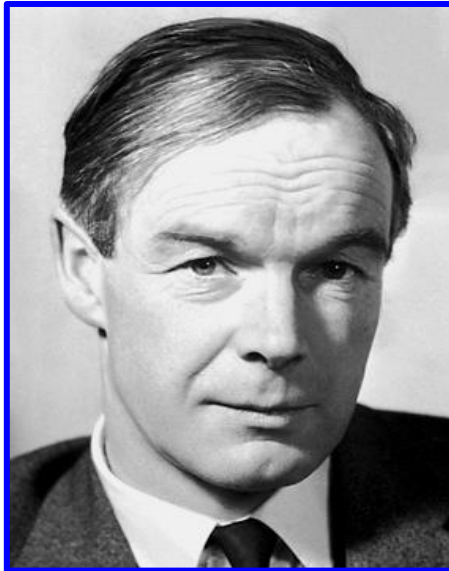


Alan Hodgkin



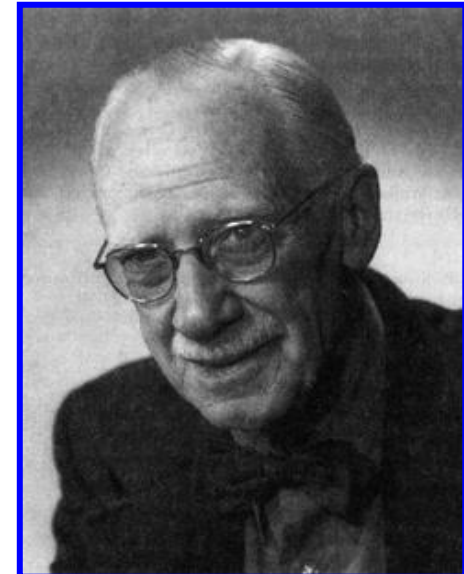
Andrew Huxley

Anomalous Inductance and Diodes in the Brain



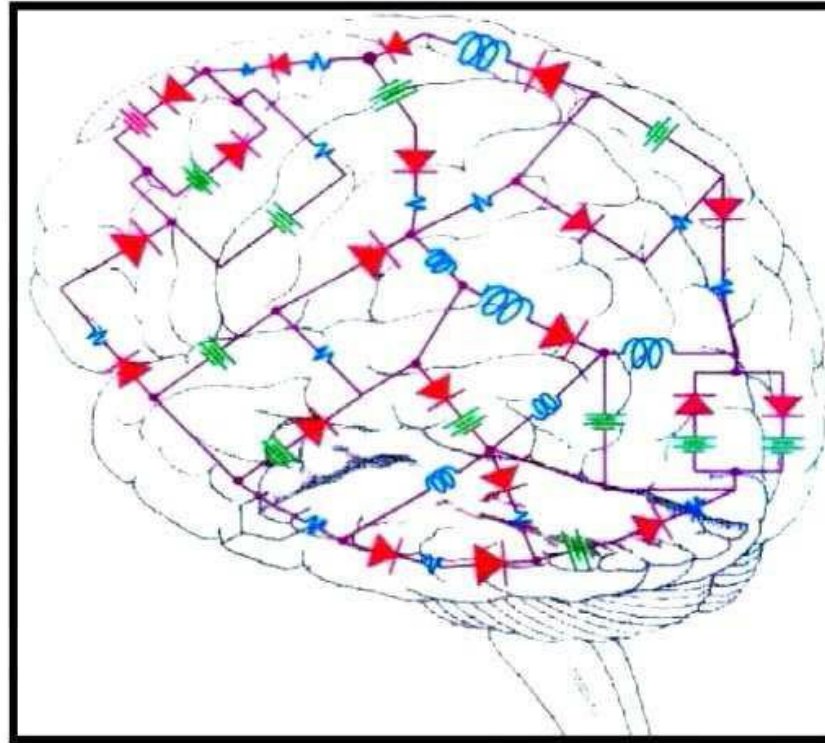
Alan L. Hodgkin

Hodgkin and Cole
have independently
measured from the
squid axon membrane
gigantic inductances
and *DC V-I Curves*
that resemble
rectifying diodes.



Kenneth S. Cole

*Where are the gigantic
magnetic fields and
rectifiers ?*



*the inductors and
diodes are
anomalous !*

Alan Hodgkin

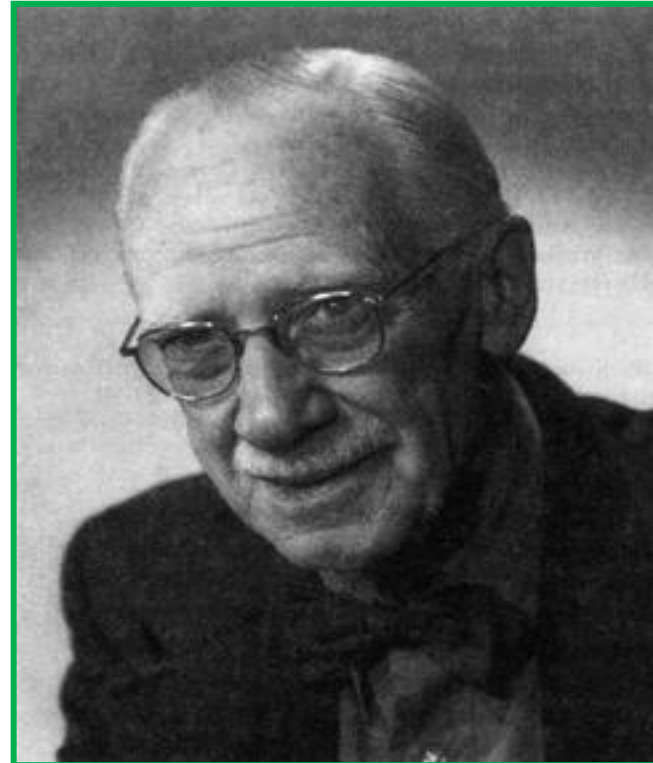
Kenneth Cole



Po_19-

Po_19-191

*The suggestion of
an inductive
reactance
anywhere in the
system was
shocking to the
point of being
unbelievable.*



Kenneth Cole

Hodgkin-Huxley Mystery

Although the Hodgkin-Huxley Equations have been around for more than 70 years, we still do not understand the complex nonlinear dynamics responsible for the generation of the *action potentials (spikes)*.

Hodgkin's Blunder

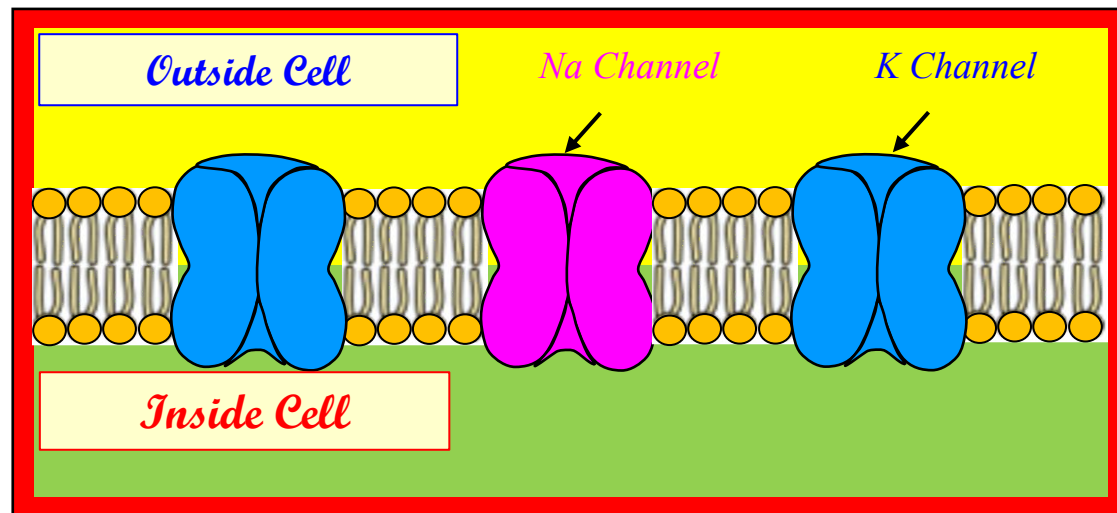
Hodgkin had struggled in vain searching for a *physical* interpretation of the *squid axon inductance*. He failed because he had mistaken the *axon* for a *time-varying conductance*, when in fact it has a simple explanation if the *Potassium* and *Sodium ion channels* are identified as *memristors*.

A. L. Hodgkin,
“The ionic basis of electrical activity in nerve and muscle,”
Biological Review, Vol. 26, pp. 339-409, 1951

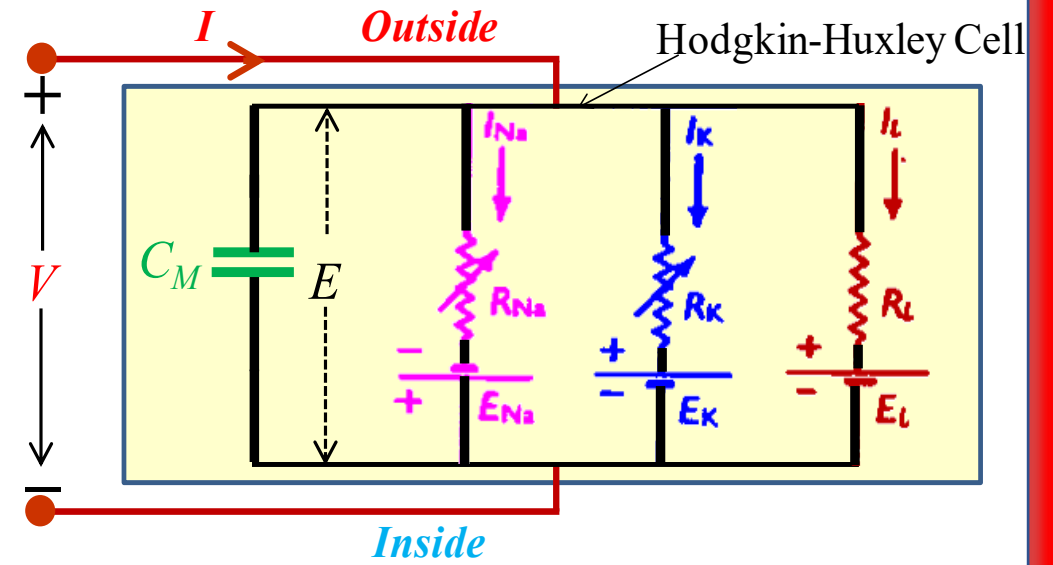
(a)



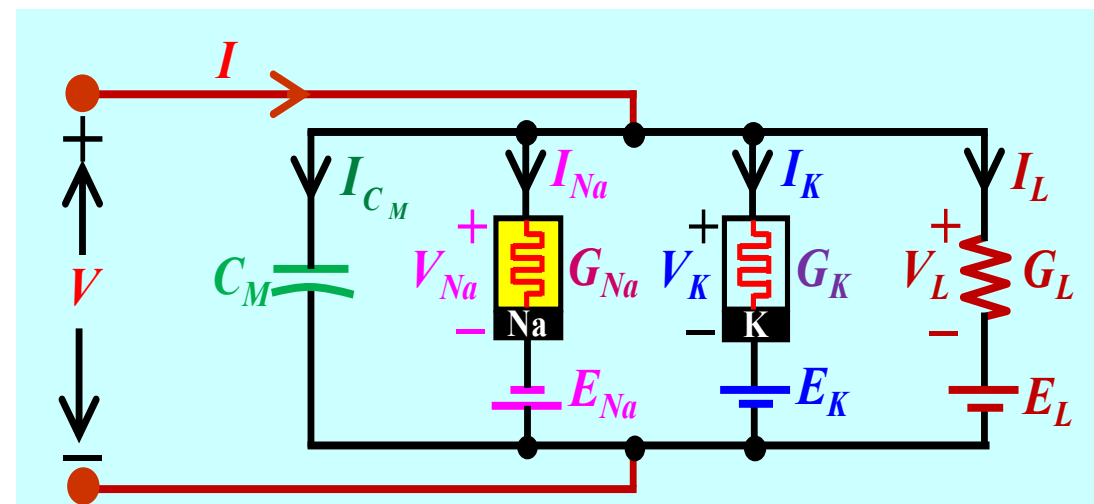
(b)



(c)



(d)



Wuh-26

International Journal of Bifurcation and Chaos, Vol. 22, No. 4
(2012) 1250098 (49 pages)

NEURONS ARE POISED NEAR THE EDGE OF CHAOS

Leon Chua, Valery Sbitnev, Hyonsuk Kim

Crisis in Circuit Theory

*Pre-1970 Definitions of the
3 Basic Circuit Elements*

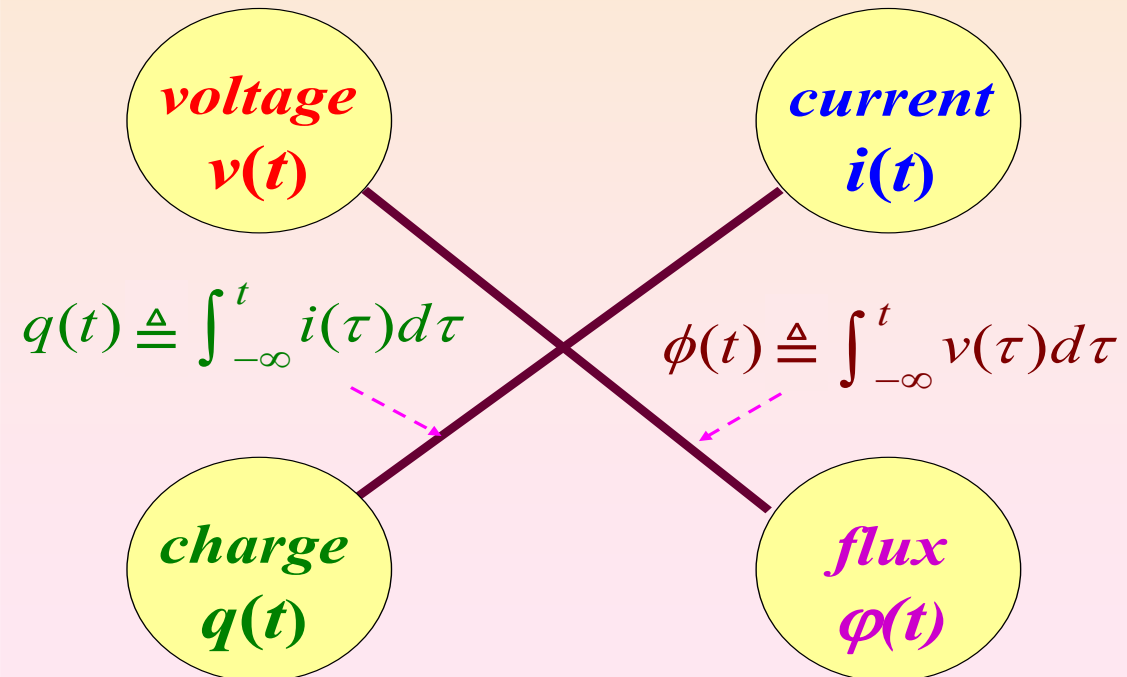
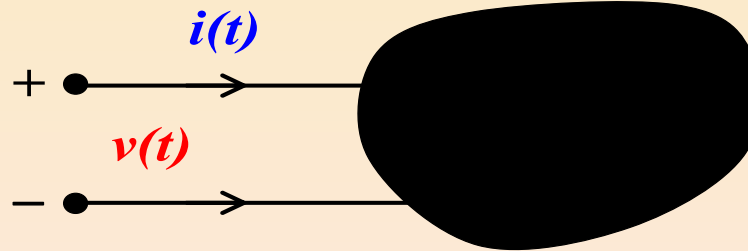
Capacitors, Resistors, and Inductors

give unrealistic or incorrect circuit solutions

*when the elements are
time-varying or nonlinear*

To Recover from
the perfect storm
Capacitors, Resistors, Inductors
must be
redefined
via an
AXIOMATIC APPROACH

Four Basic Circuit Variables



voltage, Volt V

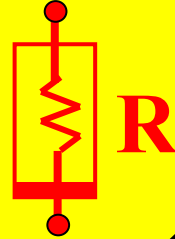
v

RESISTOR

current, Ampere A

i

$$R(v,i)=0$$



R

$$v = \frac{d\phi}{dt}$$

$$i = \frac{dq}{dt}$$

$$q \triangleq \int_{-\infty}^t i(\tau) d\tau$$

$$\phi \triangleq \int_{-\infty}^t v(\tau) d\tau$$

q

charge, Coulomb C

ϕ

flux, Weber Wb

voltage, Volt V

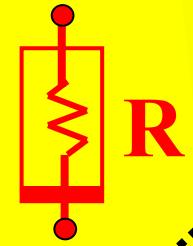
current, Ampere A

RESISTOR

v

i

$R(v,i)=0$



R

$v = \frac{d\phi}{dt}$

$i = \frac{dq}{dt}$

$q \triangleq \int_{-\infty}^t i(\tau) d\tau$

$\phi \triangleq \int_{-\infty}^t v(\tau) d\tau$

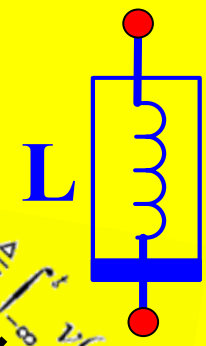
q

ϕ

charge, Coulomb C

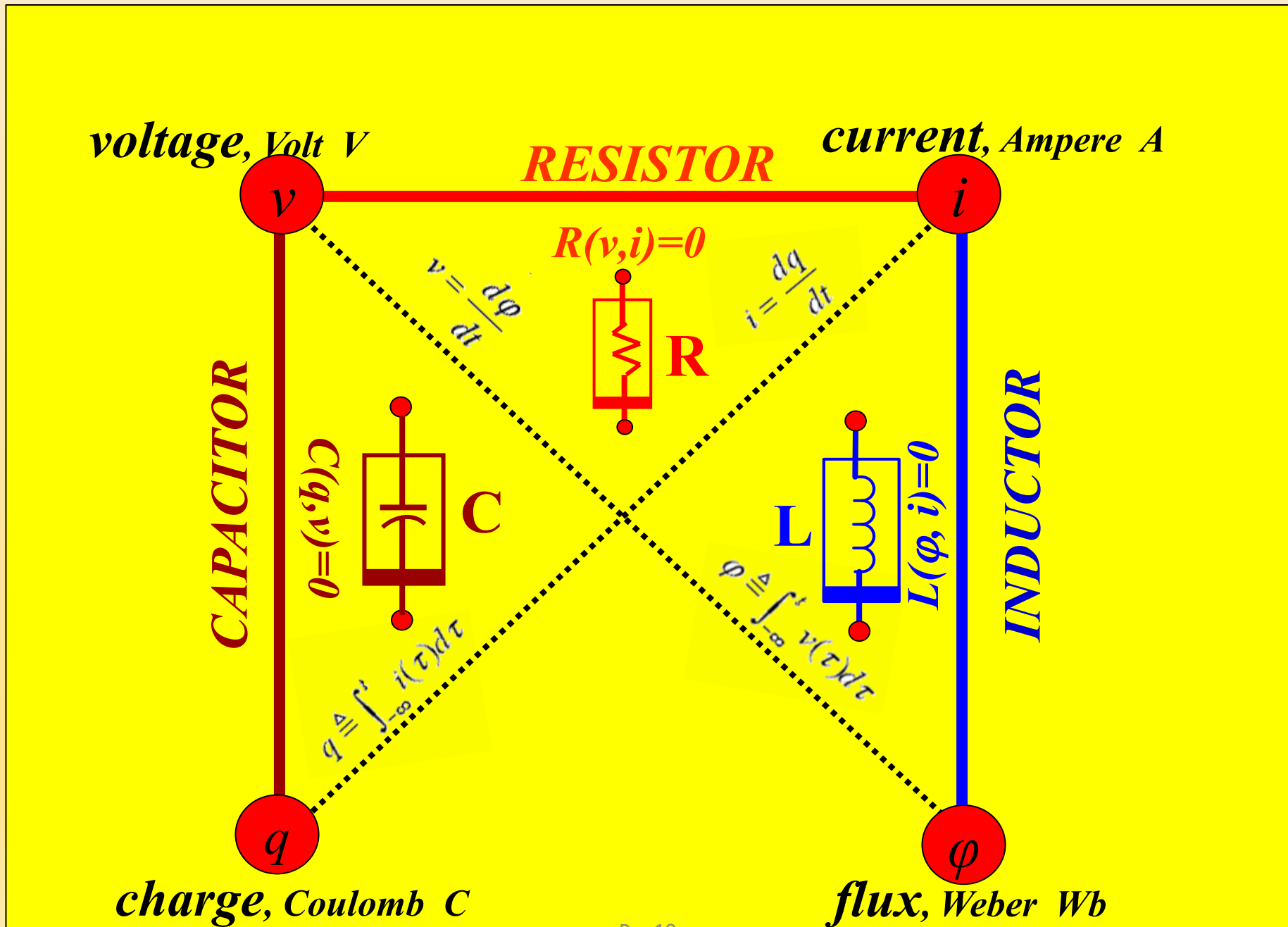
flux, Weber Wb

INDUCTOR

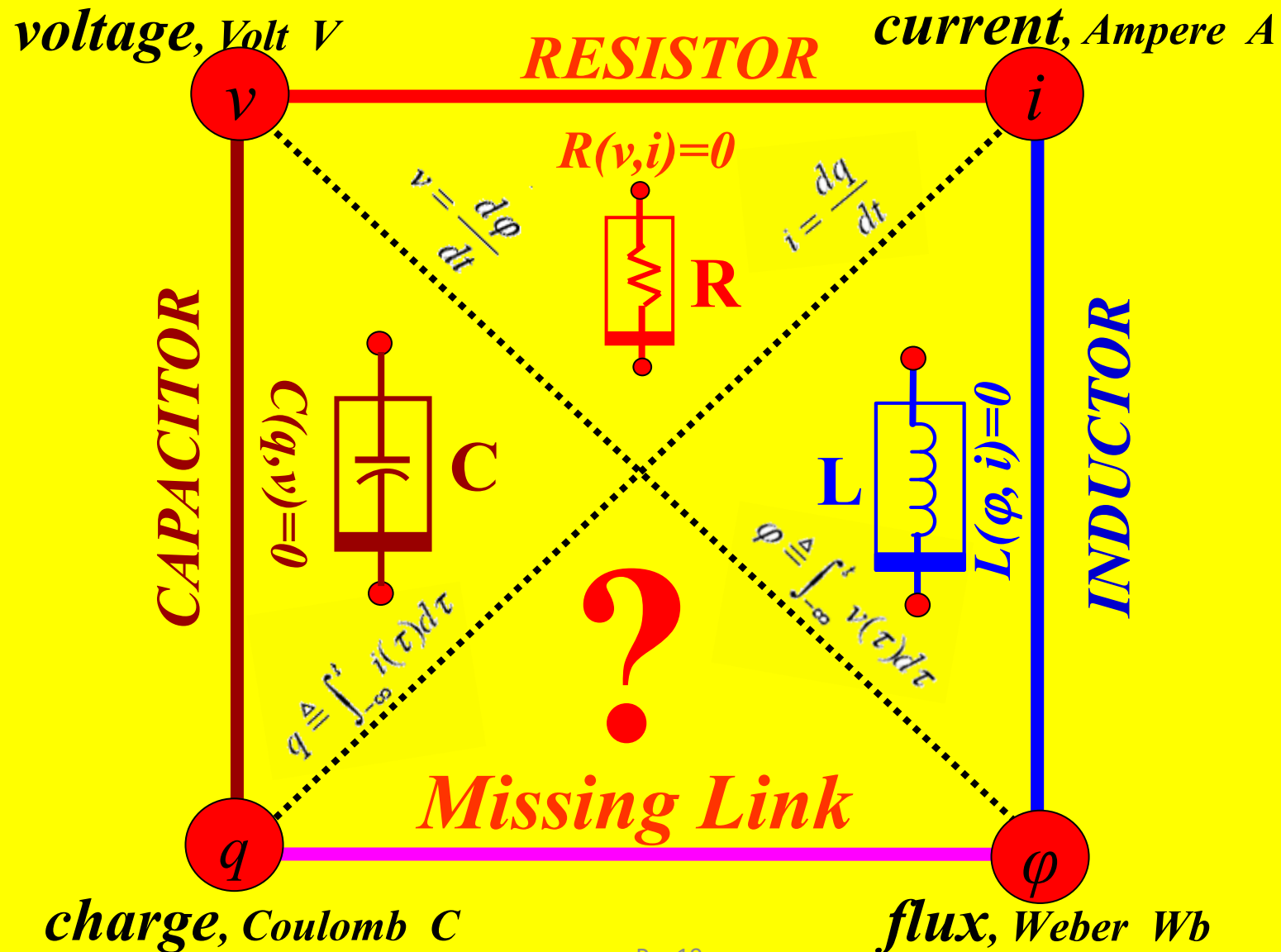


L

$L(\phi, i)=0$

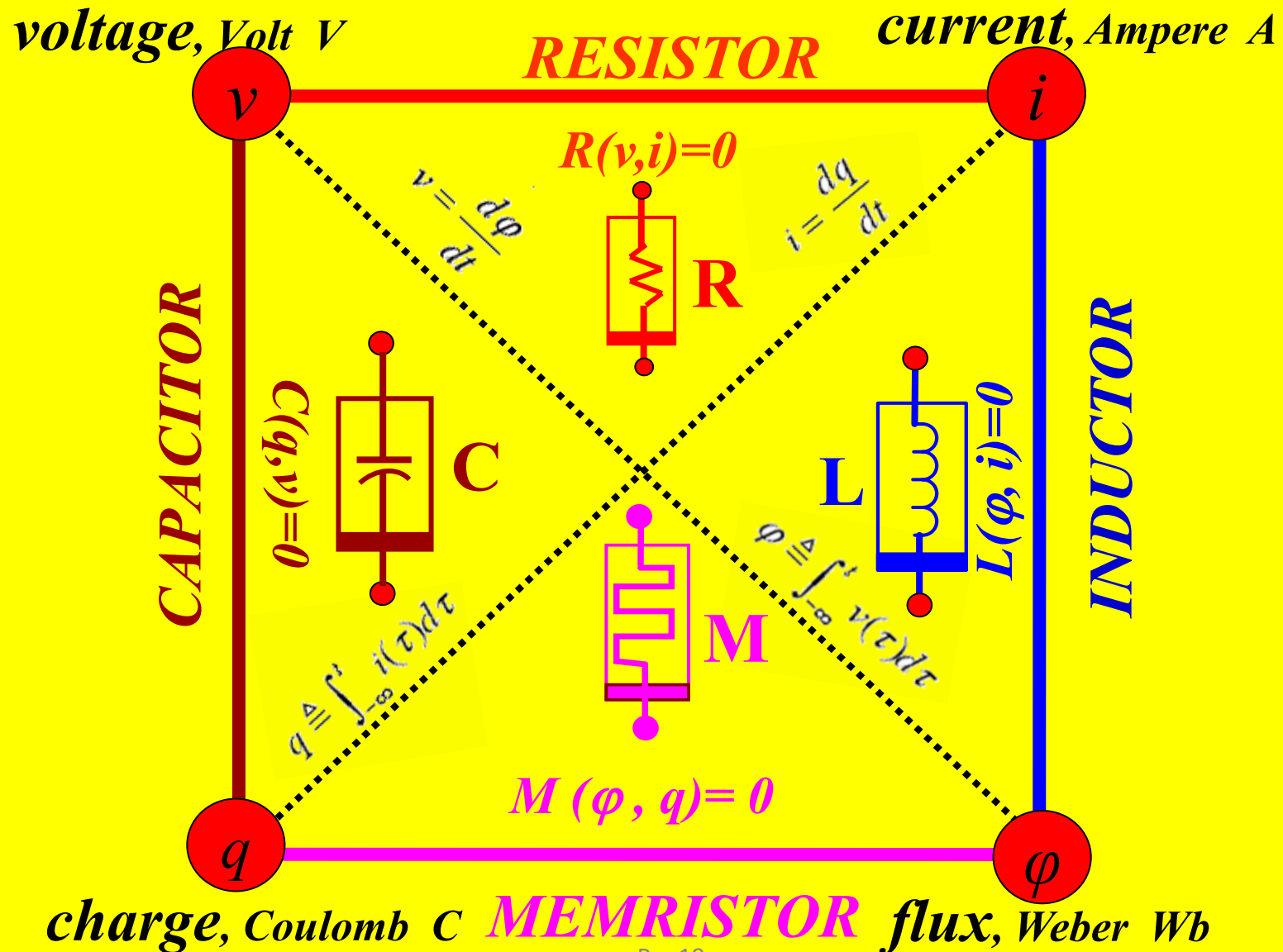


4 Basic Circuit Elements



*The missing
circuit element
is the
memristor !*

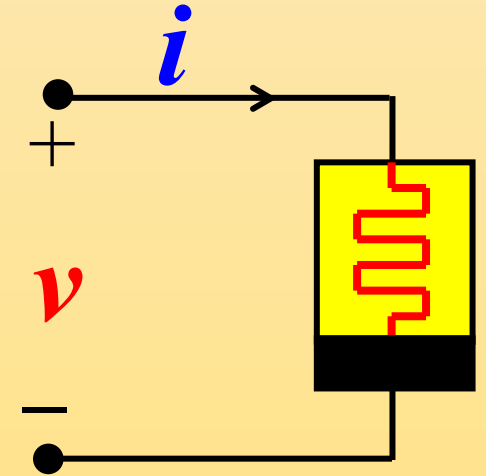
4 Basic Circuit Elements



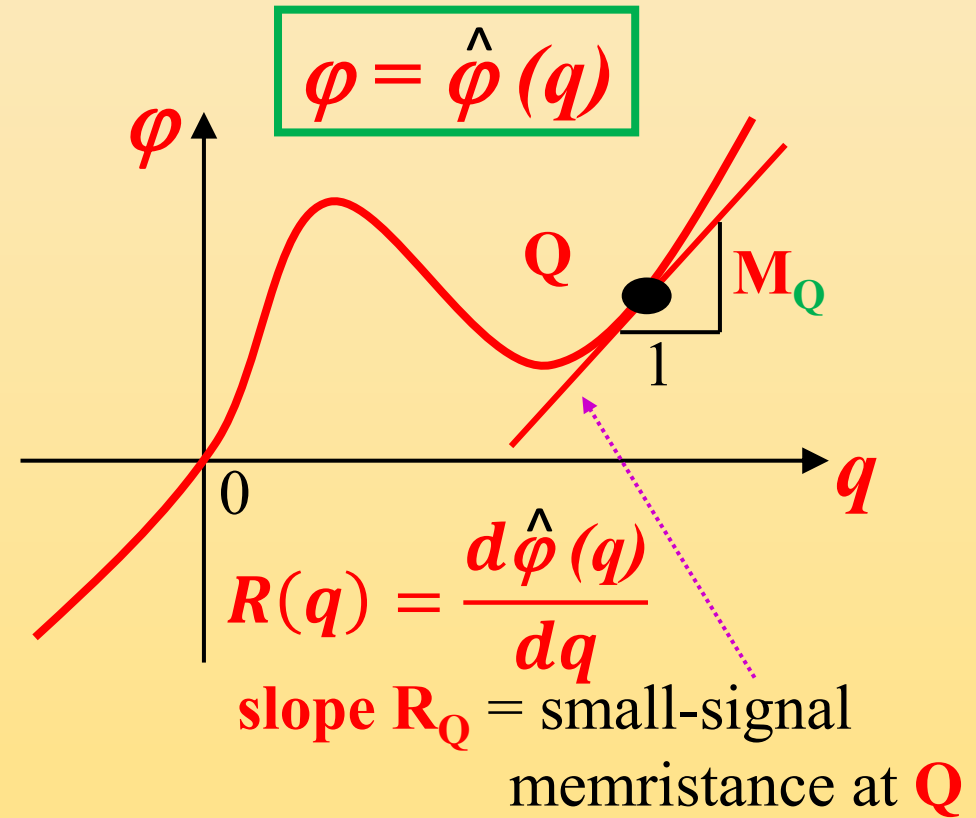
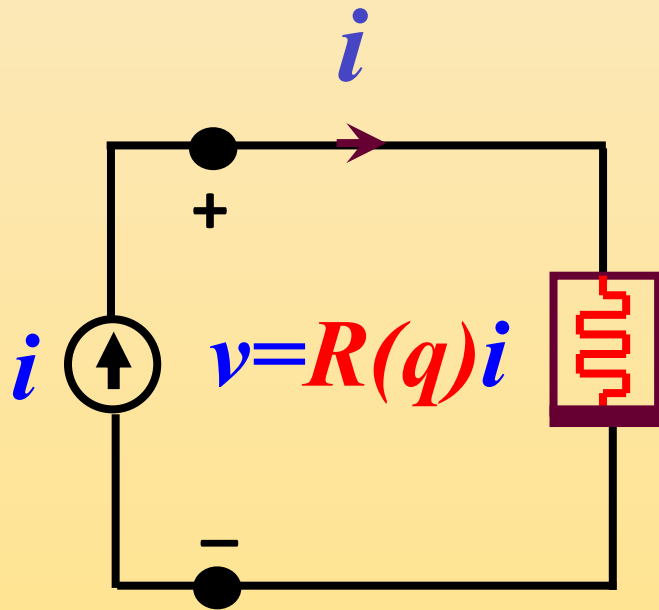
What is a *MEMRISTOR* ?

A **MEMRISTOR** is a 2-terminal device whose resistance can be tuned to some desired value by applying a voltage signal, and which **remembers** its resistance after the voltage source is disconnected.

It remembers its past!



What is a Memristor ?



At any point Q the memristor
obeys the **Ohm's Law**:

$$v = R(q) i$$

Memristor



is
not

a **switch !**

Memristor

is
a



continuum

Memristor
is an acronym for
Memory Resistor

Two *memristors*
suffice to compute
all Boolean functions

From:

E. Lehtonen, J.H. Poikonen, and M. Laiho

Electronic Letters, vol. 3, pp. 239, 2010.

Memristor—The Missing Circuit Element

LEON O. CHUA, SENIOR MEMBER, IEEE

From:

Leon O. Chua,

“Memristor - The Missing Circuit Element,”

IEEE Transaction on Circuit Theory,

Vol. CT-18, No. 5, pp. 507-519, 1971



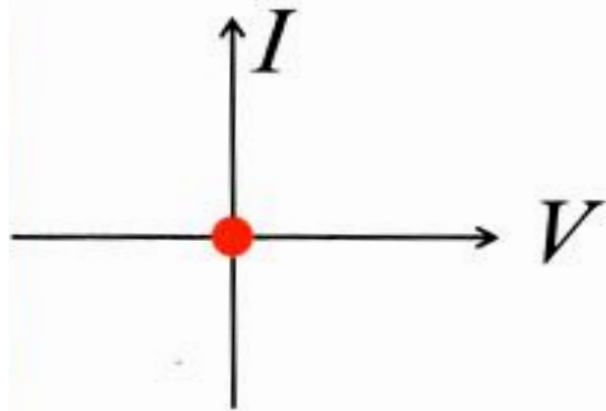
Nature **453**, 80–83 (2008). <https://doi.org/10.1038/nature06932>

The Missing Memristor Found

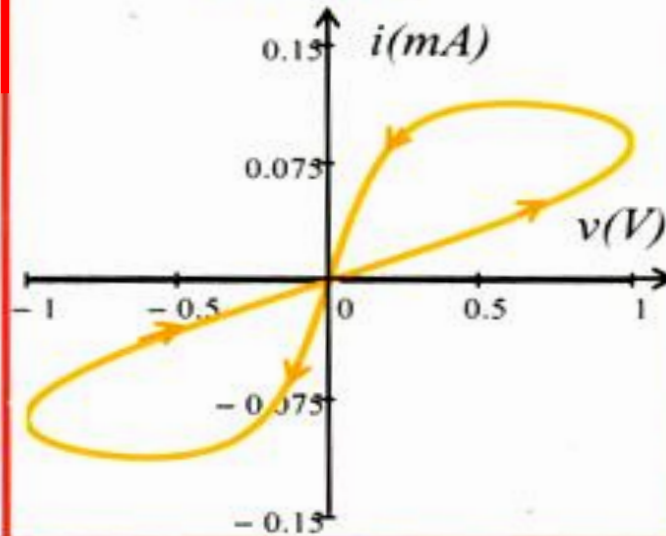
Dmitri B. Strukov, Gregory S. Snider, Duncan R. Stewart and R. Stanley Williams

HP Memristor

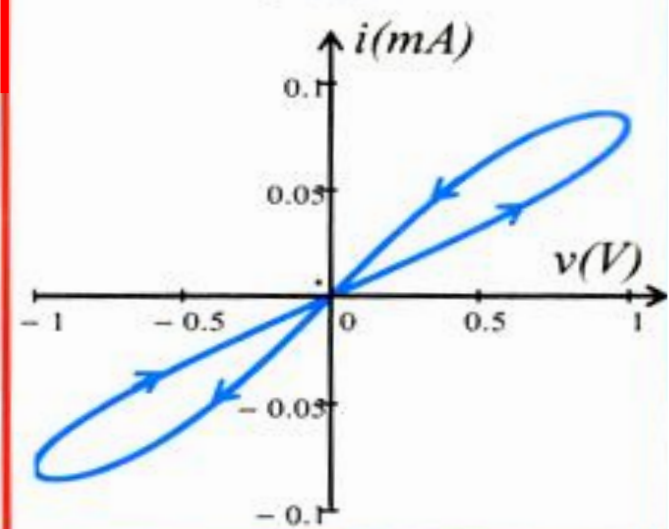
DC V-I Curve



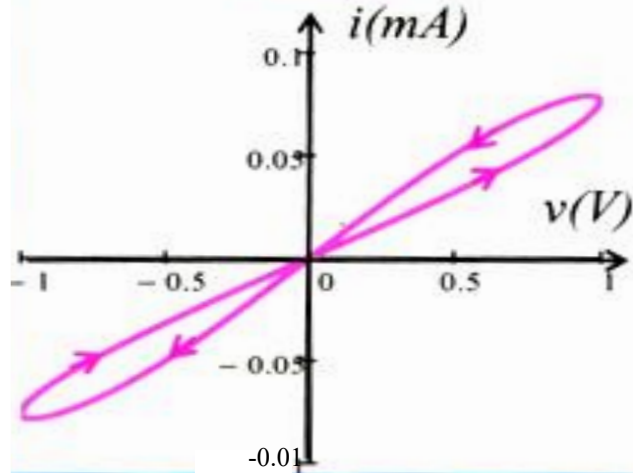
$f=0.4\text{Hz}$



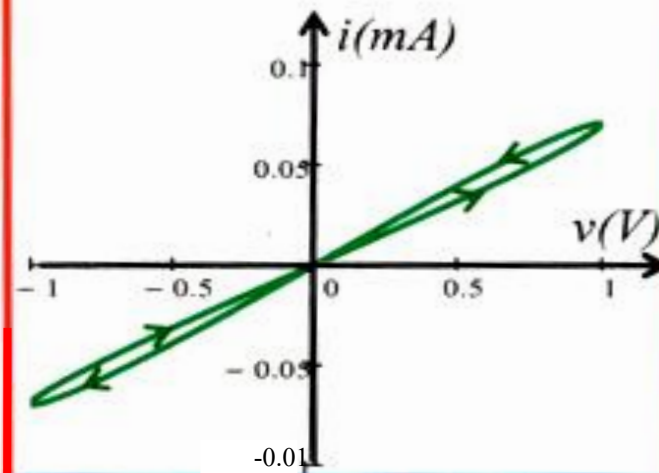
$f=0.5\text{Hz}$



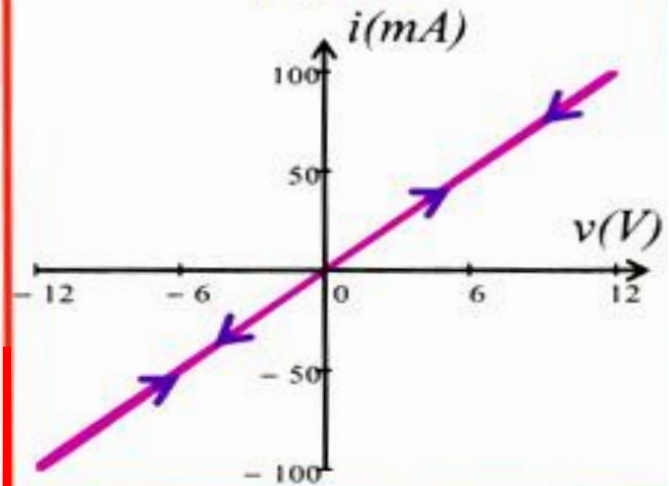
$f=0.6\text{Hz}$



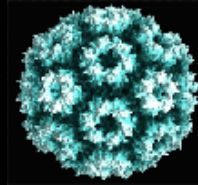
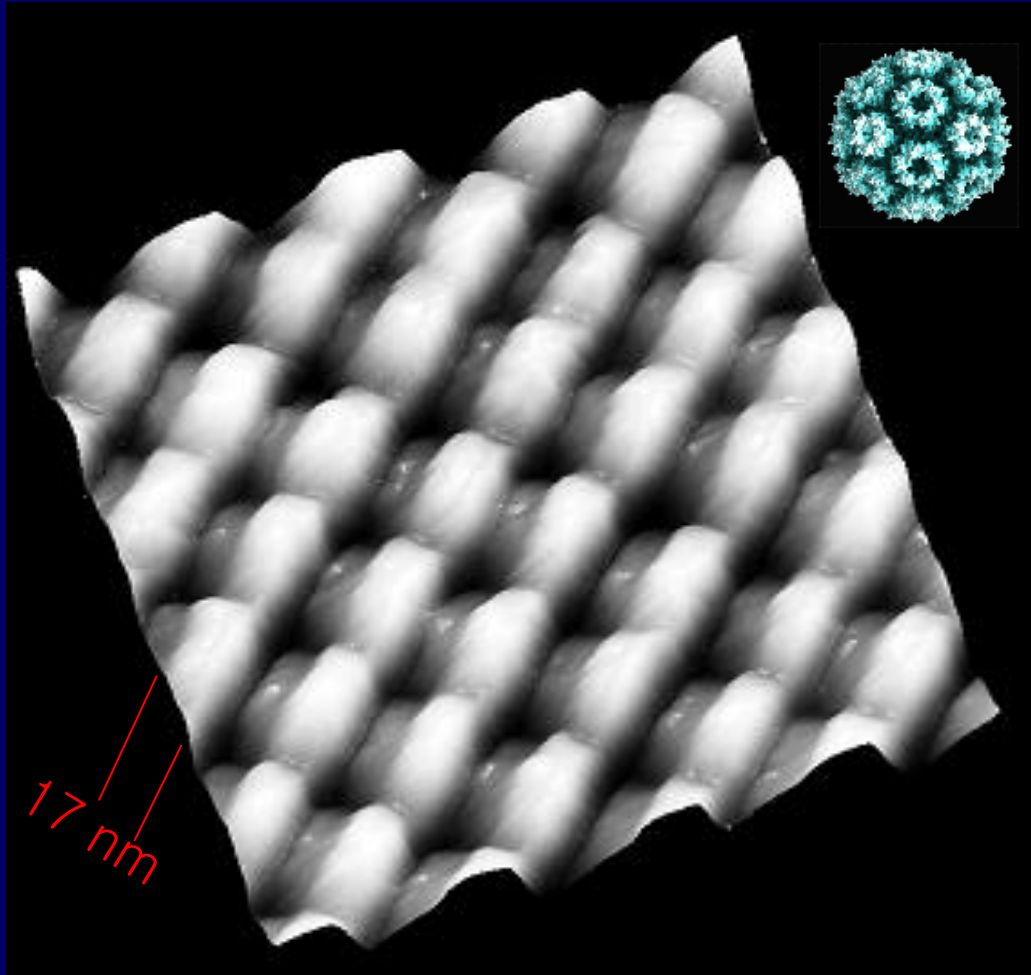
$f=1\text{Hz}$



$f=10\text{Hz}$



AFM image of memristor crossbar at 17nm

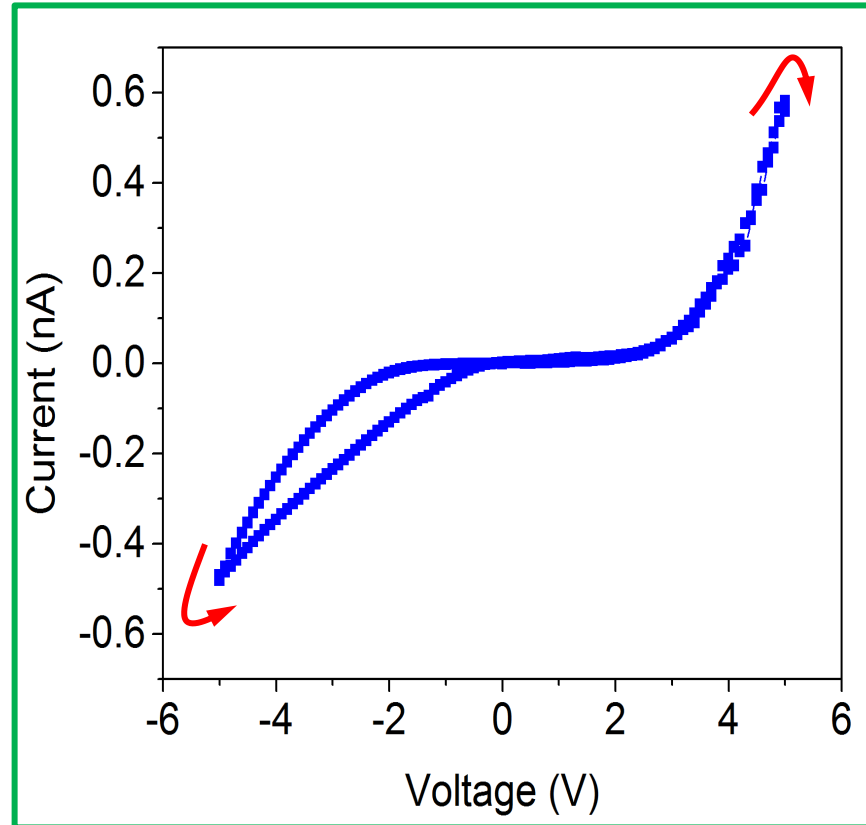
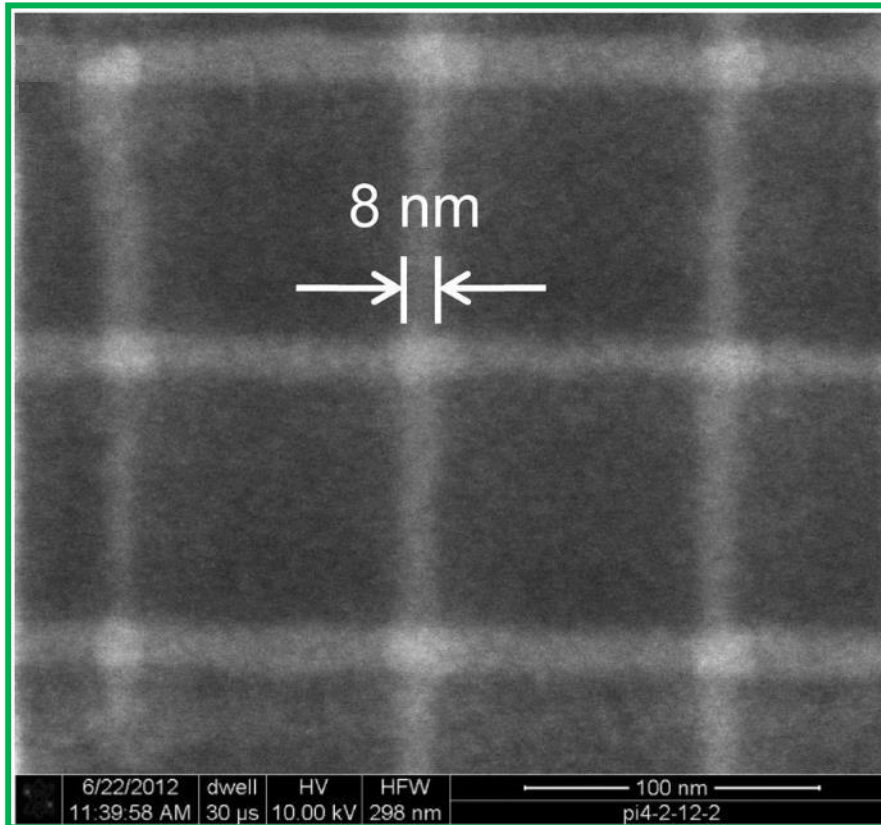


The smallest virus:
~25 nm in dia

Fabricated by
Imprint Lithography –
Can be stacked for
Higher area density

(cell density : 100 Gbit/cm²)

An 8 nm Memristor



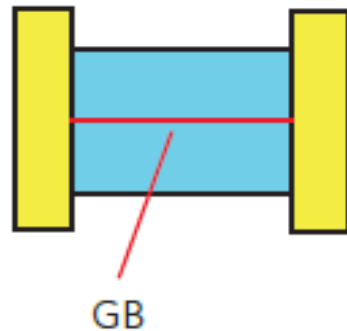
From:

S. Pi, P. Lin, Q. Xia, “Cross point arrays of 8 nm × 8 nm memristive devices fabricated with nano imprint lithography”, J. Vac. Sci. Technol. B 31, 06FA02-1 - 06FA02-6, 2013

Memristor

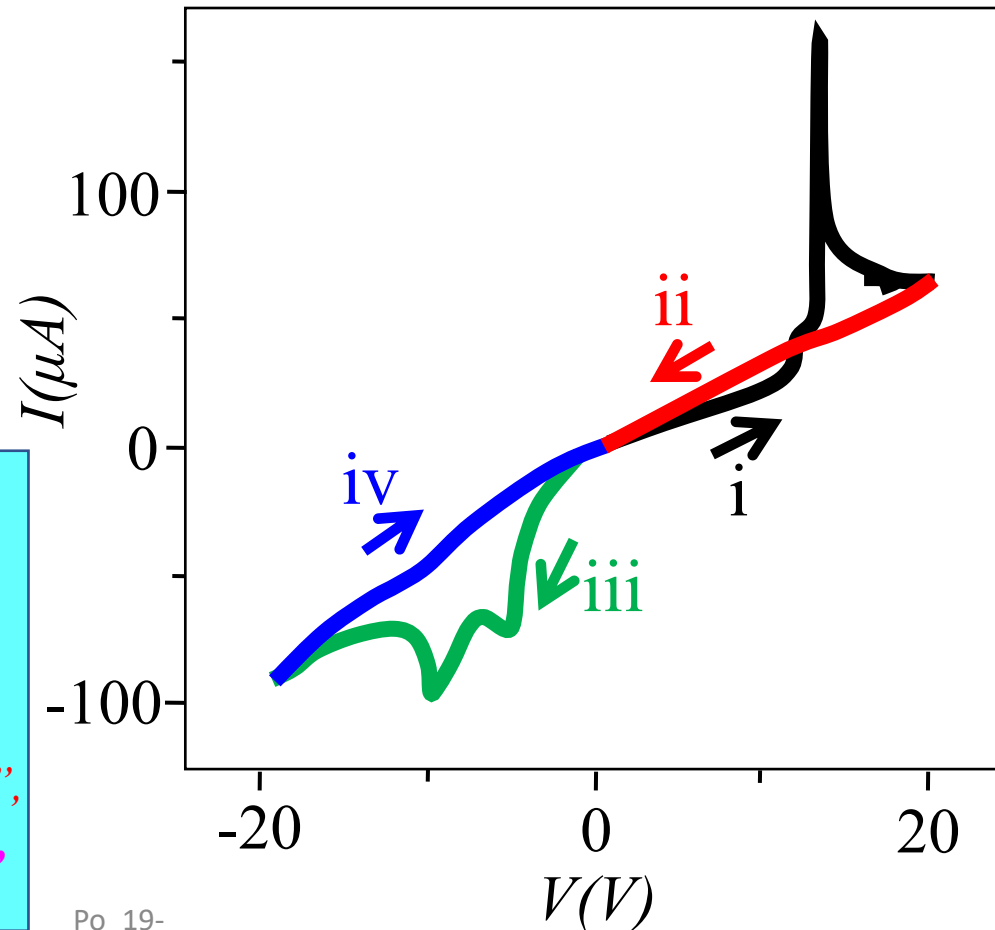
made from

A single Layer of the Molecule MoS_2

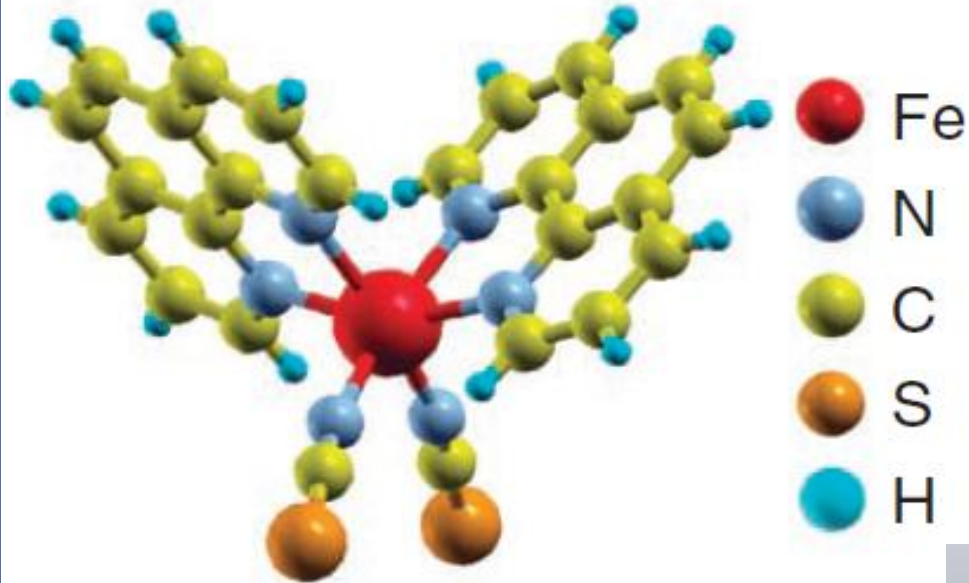


From:

*V. K. Sangwan, D. Jariwala, I. S. Kim,
K. S. Chen, T. J. Marks, L. J. Lauhon,
M. C. Hersam, "Gate-tunable
memristive phenomena mediated by
grain boundaries in single-layer MoS_2 ",
Nature Nanotechnology 10, p. 403-406,
2015*



One-Molecule Memristor



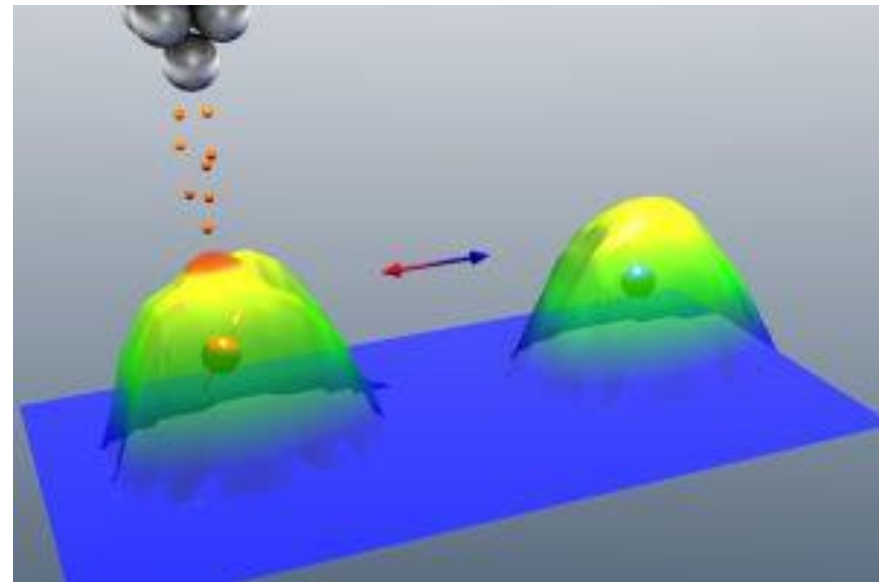
Toshio Miyamachi et al.

“Robust spin crossover and memristance across a single molecule”

Nature Communications 3, 938, p.1-6, 2012

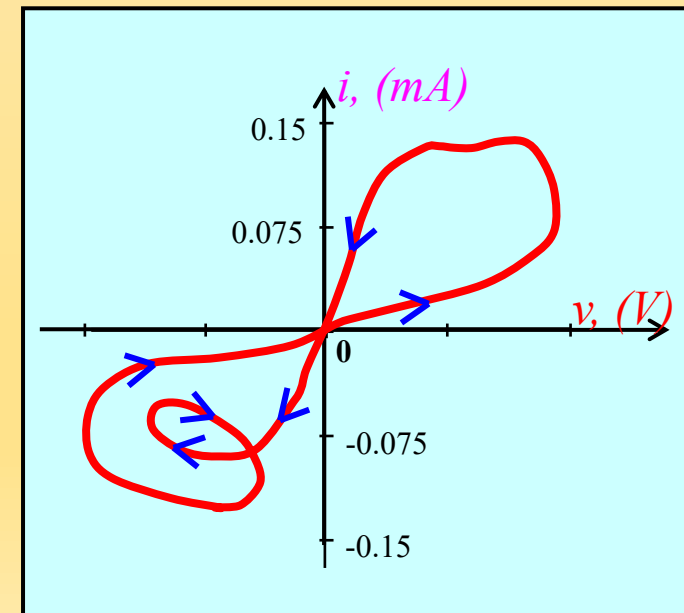
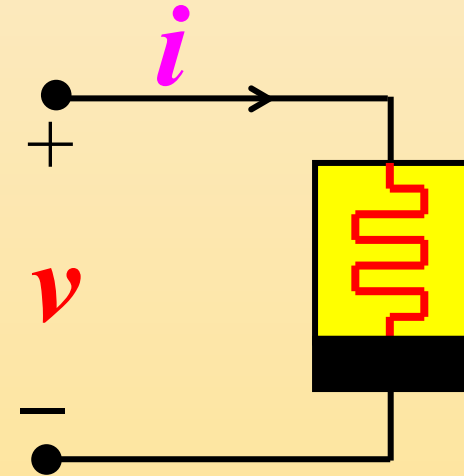
Nature: Molecule Changes Magnetism and Conductance

KIT Press Release No.116, July 2012
Karlsruhe Institute of Technology.



Experimental Definition of the **Memristor**

If it's Pinched,
it's a
Memristor



Some **Virtues** of *MEMRISTORS* ?

- *Memory* for free (i.e., no battery is needed to retain its memory)
- *Very High Speed* (nano seconds)
- *Very Low Power* (nano watts)
- *Very Small* (sub-nanometer)
- *Great Data Retention*
- *Great Endurance*
- *Compatibility with CMOS Technology*

Memristors Can Be Made from Many Different Materials

- **RRAM** *Memristors* (metal oxides TiO_2 , TaO_x , etc.)
- **Polymeric** *Memristors* (conducting polymers)
- **Ferroelectric** *Memristors* (Ferroelectric films)
- **Manganite** *Memristors* (Perovskite manganite)
- **Spintronic** *Memristors* (spin-transfer torque magnetic layers)

Memristors fabrication

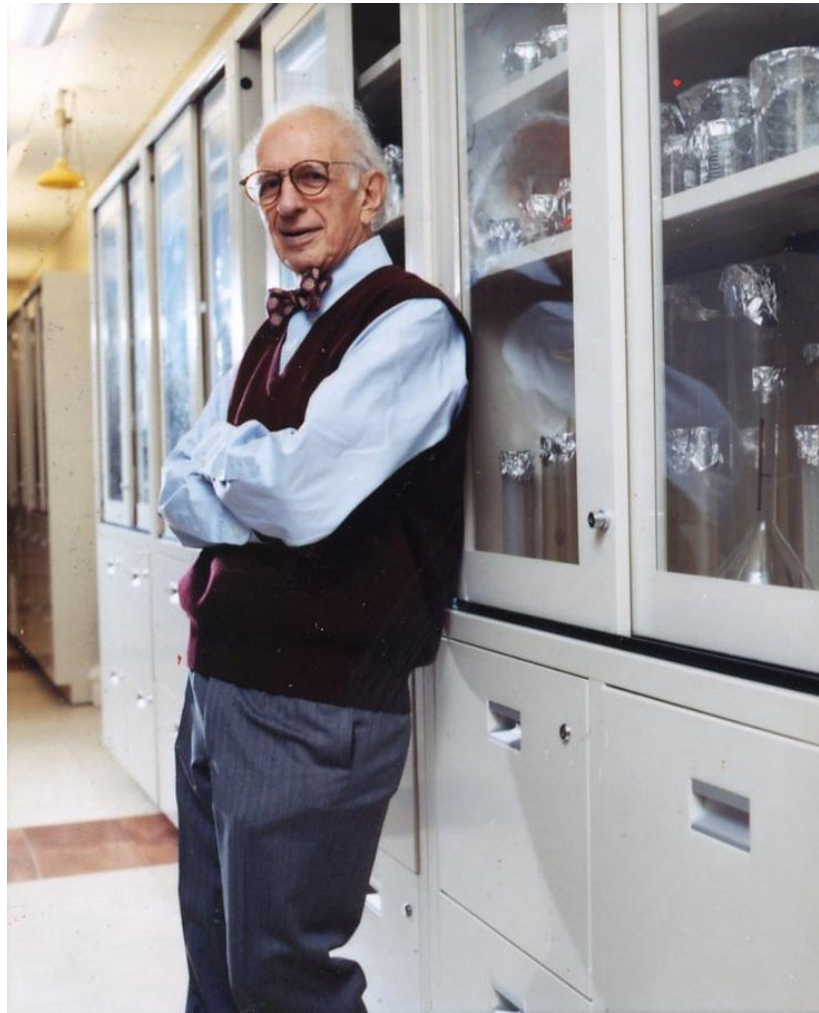
Everything **switches!**

I	II											III	IV	V	VI	VII	VIII
1 H																	2 He
3 Li	4 Be											5 B	6 C	7 N	8 O	9 F	10 Ne
11 Na	12 Mg											13 Al	14 Si	15 P	16 S	17 Cl	18 Ar
19 K	20 Ca	21 Sc	22 Ti	23 V	24 Cr	25 Mn	26 Fe	27 Co	28 Ni	29 Cu	30 Zn	31 Ga	32 Ge	33 As	34 Se	35 Br	36 Kr
37 Rb	38 Sr	39 Y	40 Zr	41 Nb	42 Mo	43 Tc	44 Ru	45 Rh	46 Pd	47 Ag	48 Cd	49 In	50 Sn	51 Sb	52 Te	53 I	54 Xe
55 Cs	56 Ba		72 Hf	73 Ta	74 W	75 Re	76 Os	77 Ir	78 Pt	79 Au	80 Hg	81 Tl	82 Pb	83 Bi	84 Po	85 At	86 Rn
87 Fr	88 Ra		104 Rf	105 Db	106 Sg	107 Bh	108 Hs	109 Mt	110 Ds	111 Rg	112 Cn	113 Uut	114 Fl	115 Uup	116 Lv	117 Uus	118 Uuo
119 Uun																	
			57 La	58 Ce	59 Pr	60 Nd	61 Pm	62 Sm	63 Eu	64 Gd	65 Tb	66 Dy	67 Ho	68 Er	69 Tm	70 Yb	71 Lu
			89 Ac	90 Th	91 Pa	92 U	93 Np	94 Pu	95 Am	96 Cm	97 Bk	98 Cf	99 Es	100 Fm	101 Md	102 No	103 Lr

binary oxides
 complex oxides
 other compounds

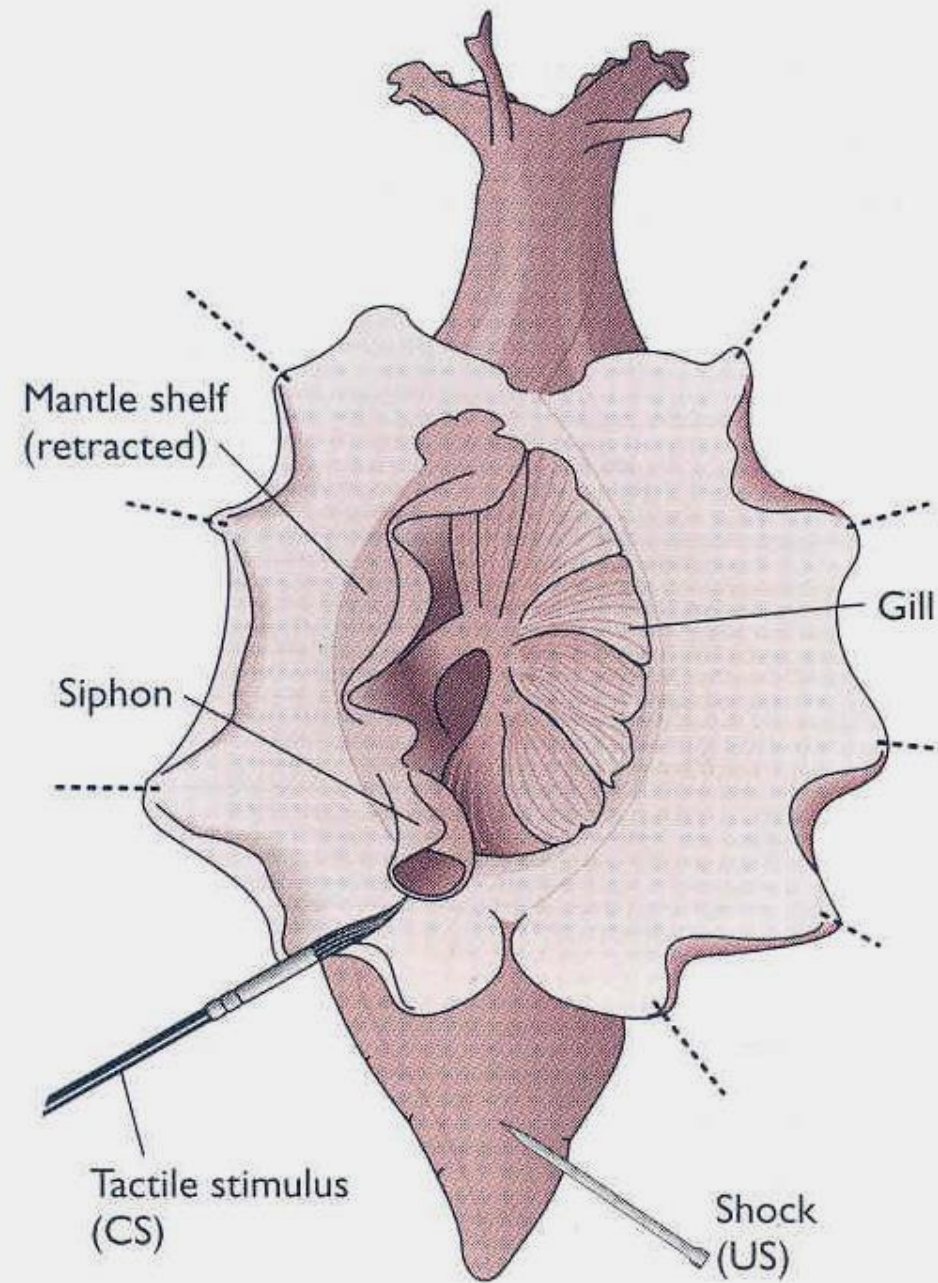
Aplysia honored with a Nobel Prize Medal





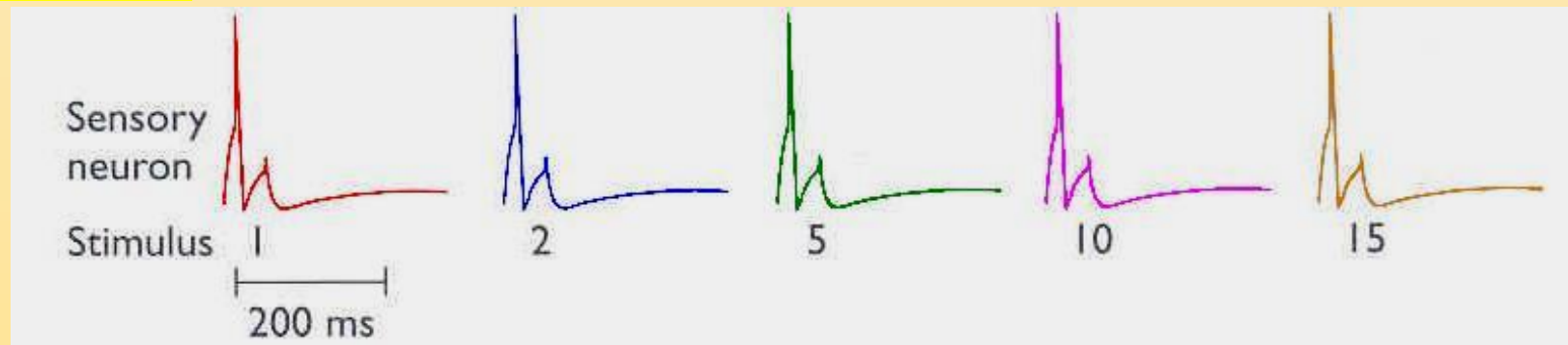
Eric R. Kandel

Nobel Prize in Physiology or Medicine, 2000.



Example of Déjà vu Response of the Aplysia

Excitation



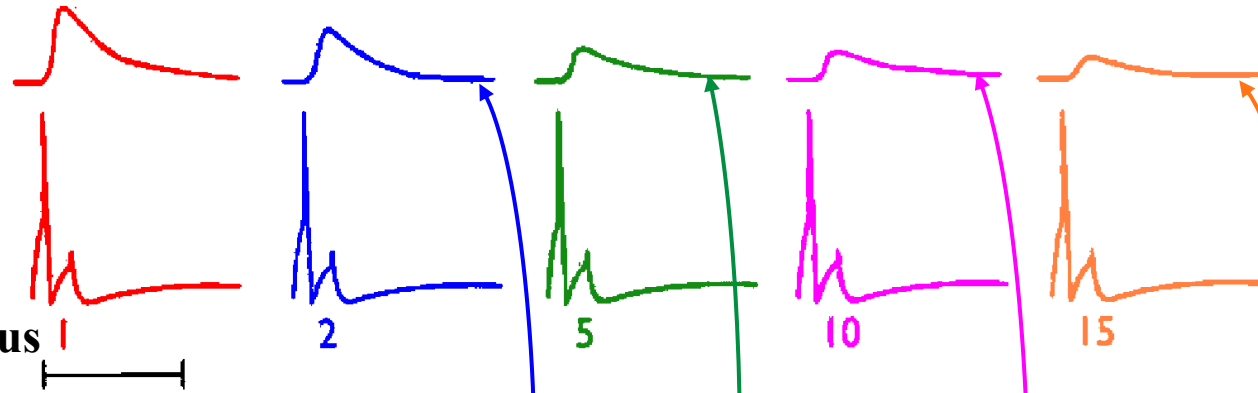
Response



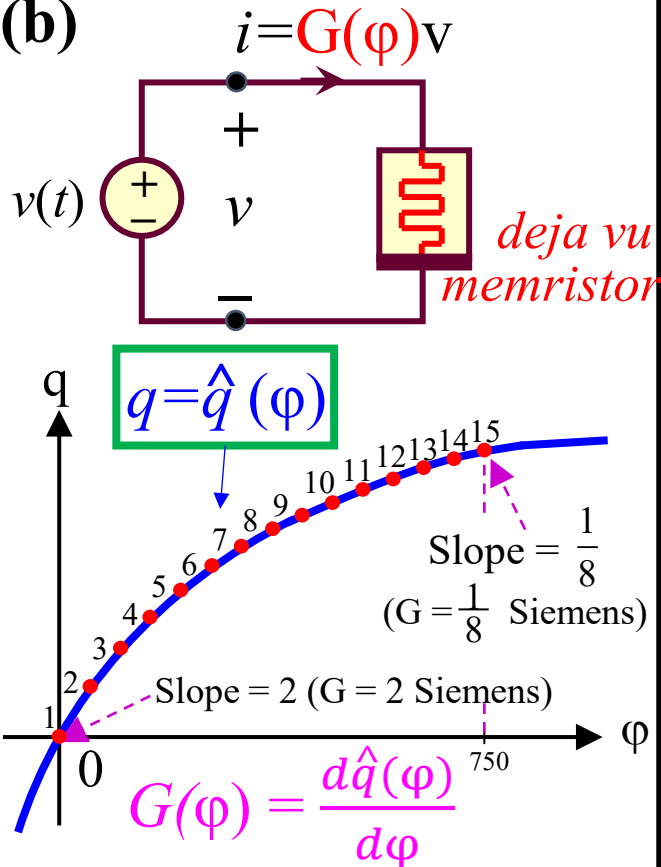
(a)

Sensory
Neuron

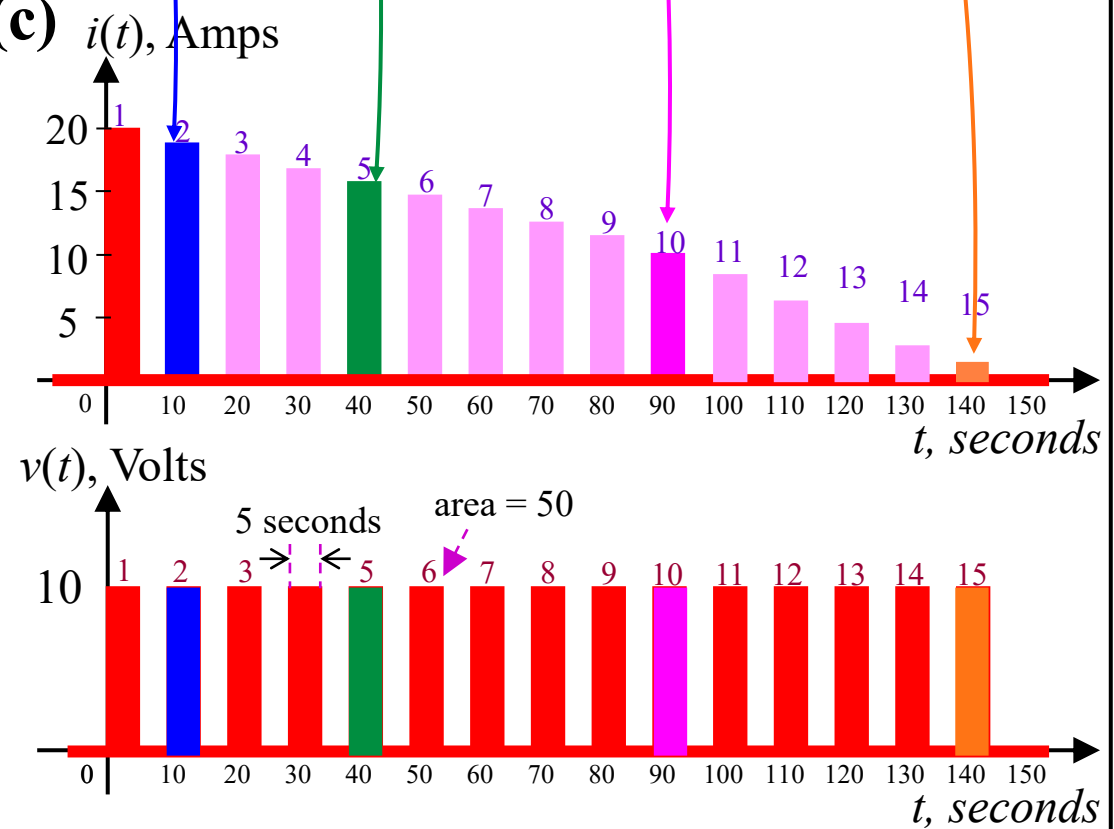
Stimulus

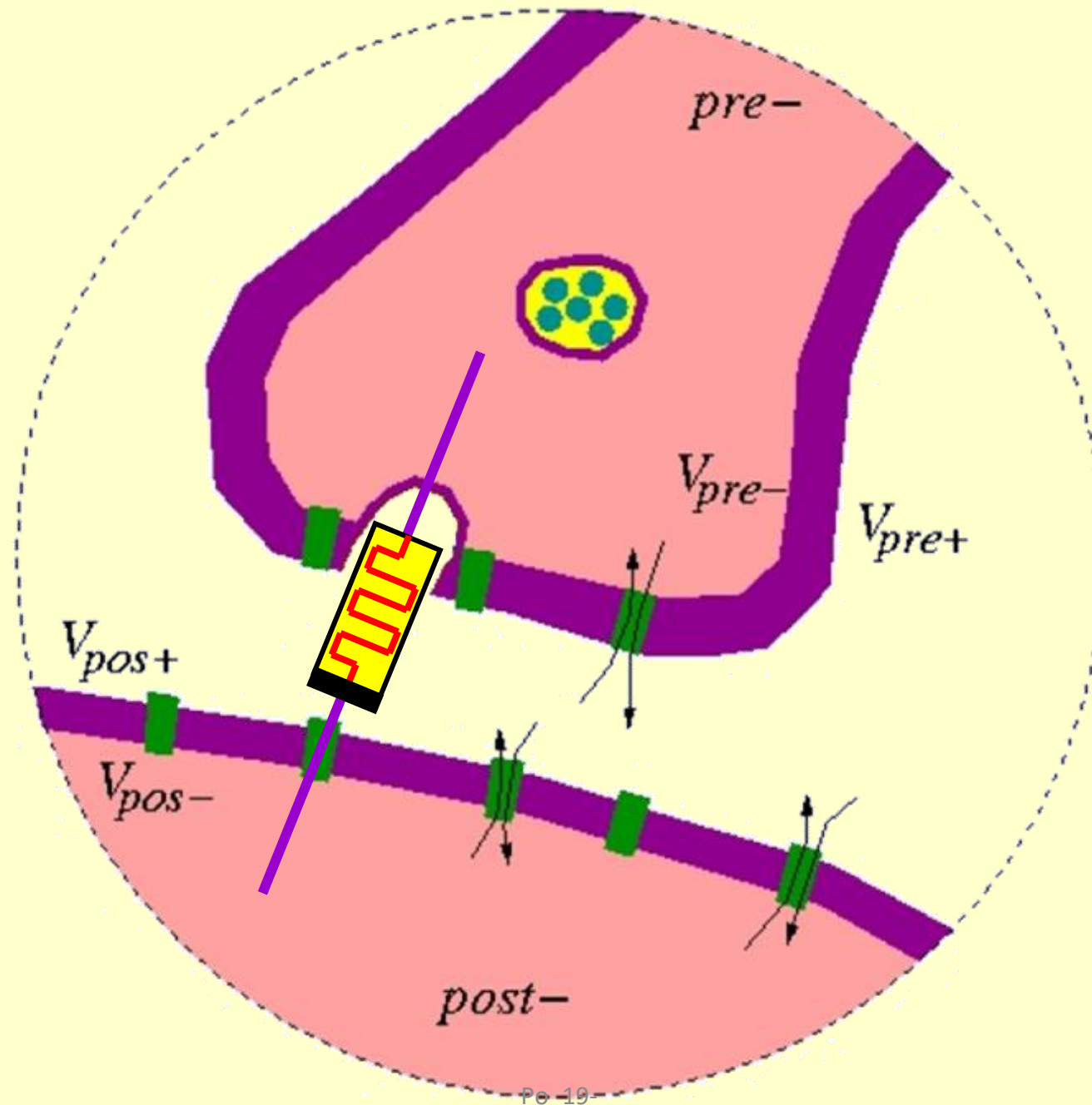


(b)



(c)

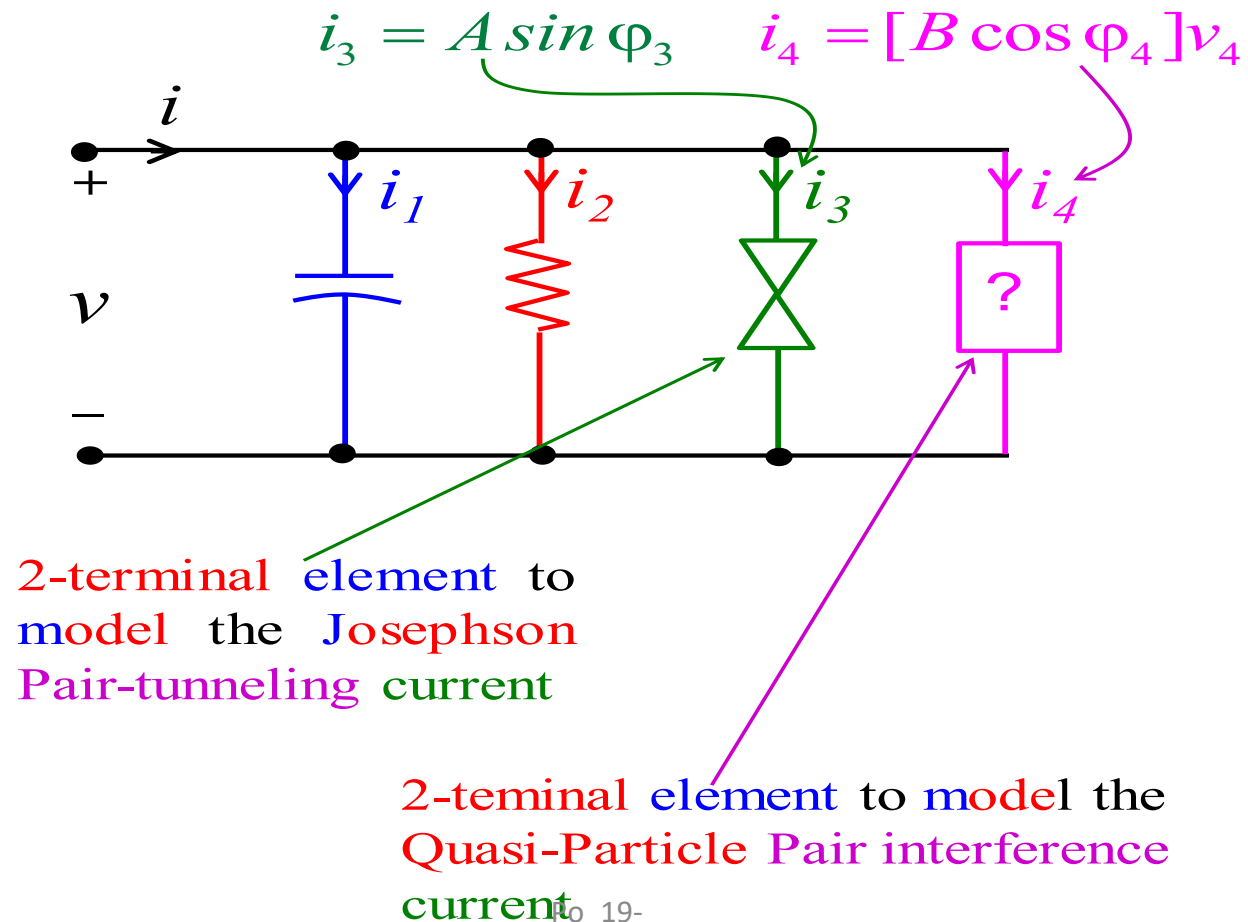




Synapses
are
Memristors!

Brian Josephson
1973 Nobel Prize in Physics:

JOSEPHSON JUNCTION CIRCUIT MODEL

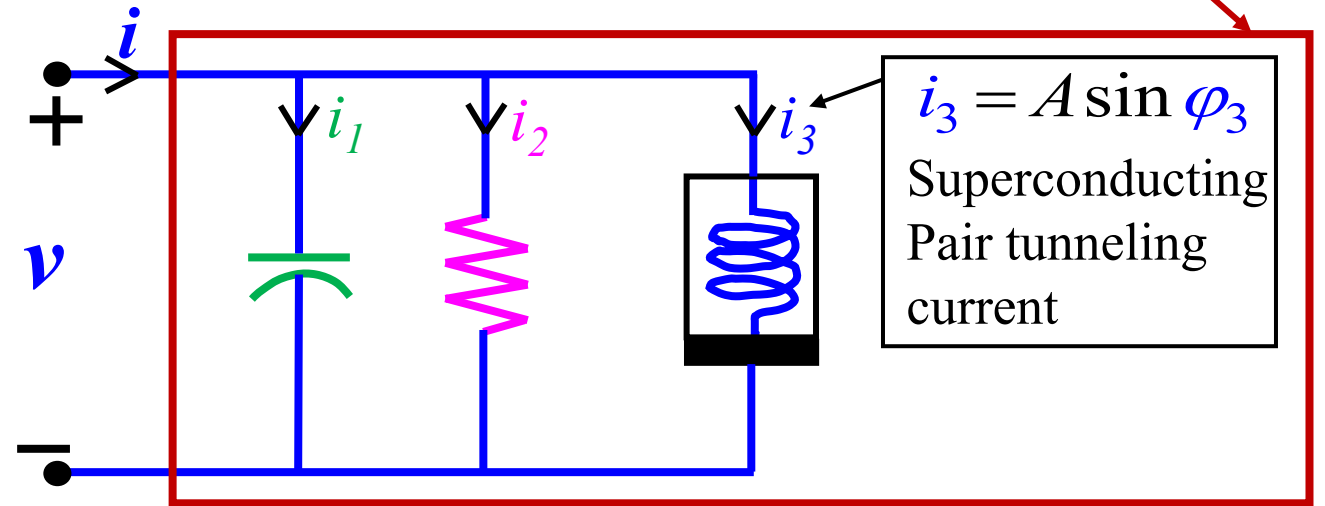




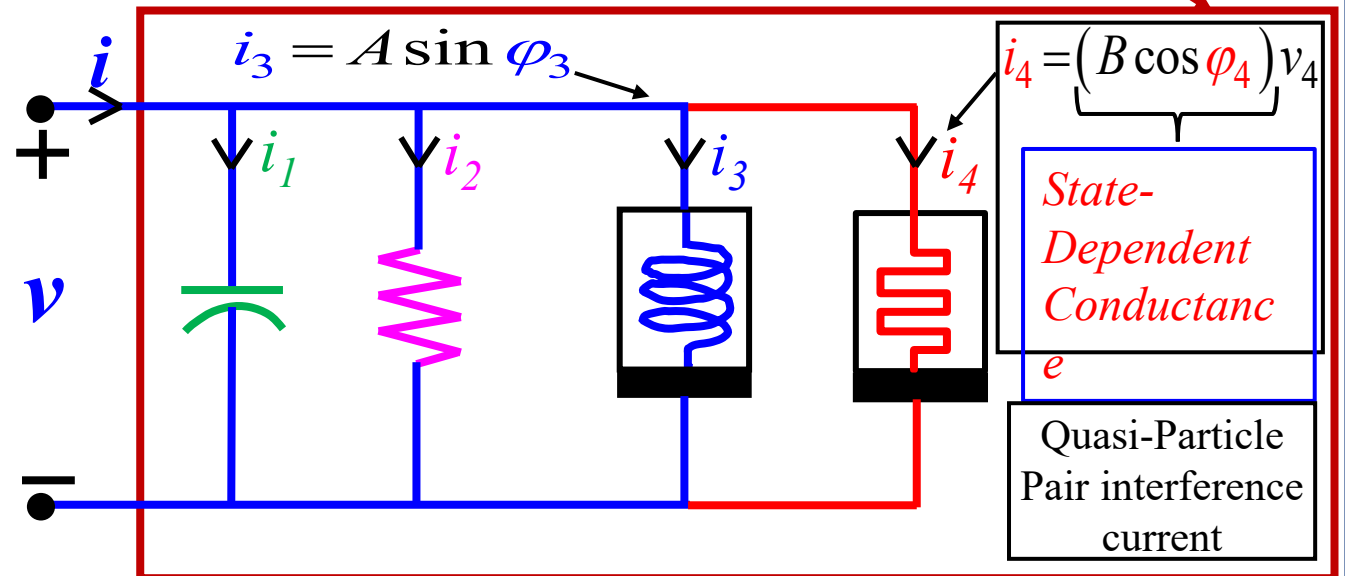
B.D. Josephson
Nobel Prize in Physics
(1973)

Po_19-92

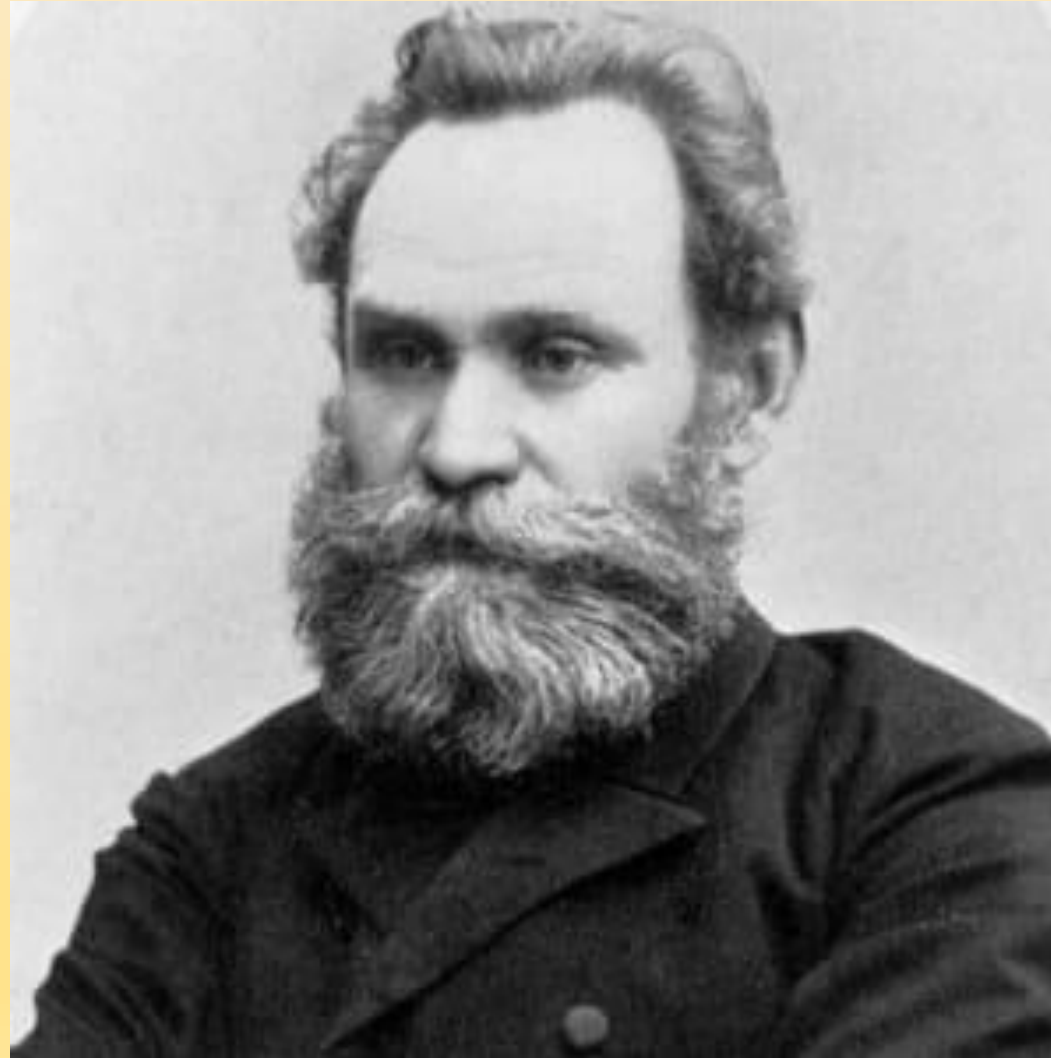
Original Josephson Junction Circuit Model [1962]



Improved Josephson Junction Circuit Model

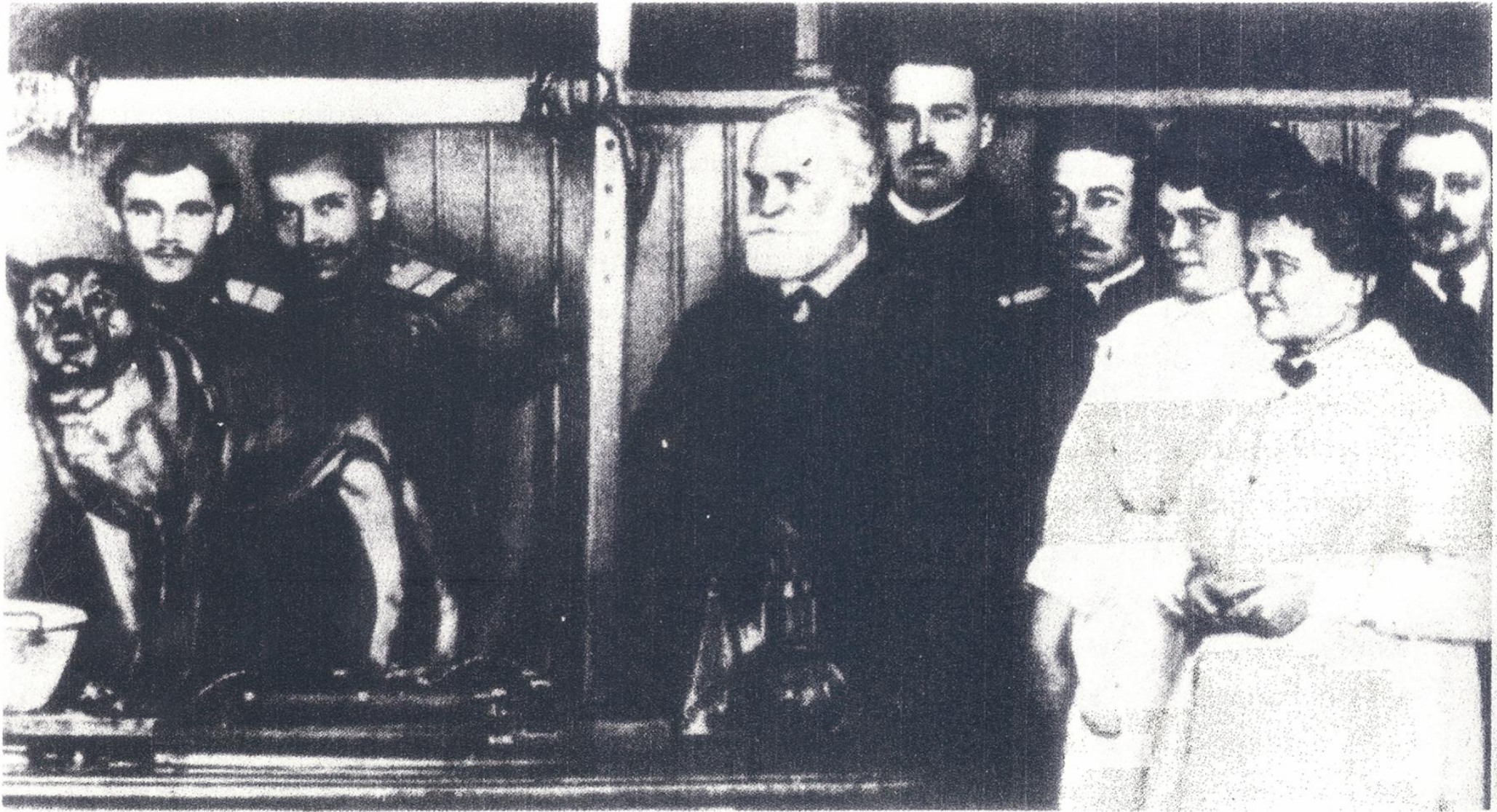


1904 Nobel Prize in Physiology or Medicine



Ivan Petrovich Pavlov (1849-1936)

Wuhan_201
9-159



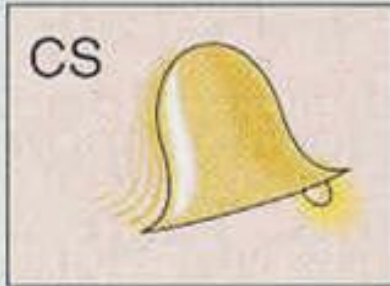
Pavlov, assistants, visitors, and dog. Pavlovian (classical) conditioning is the simplest associative learning paradigm.

Stimulus

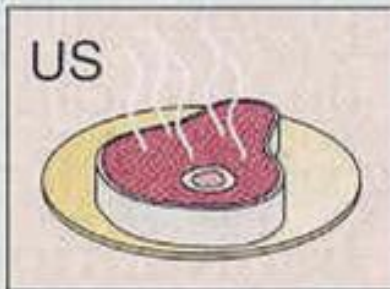
Response

Before conditioning

CS



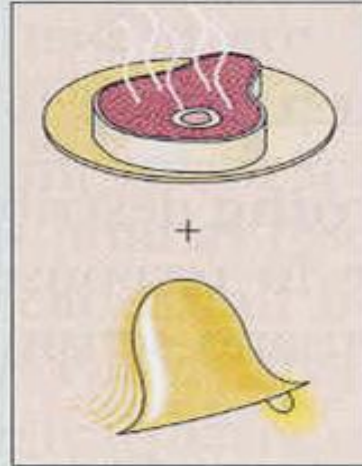
US



Stimulus

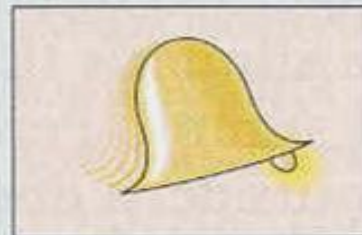
Response

Conditioning



After conditioning

CR



Associative Memory and Learning

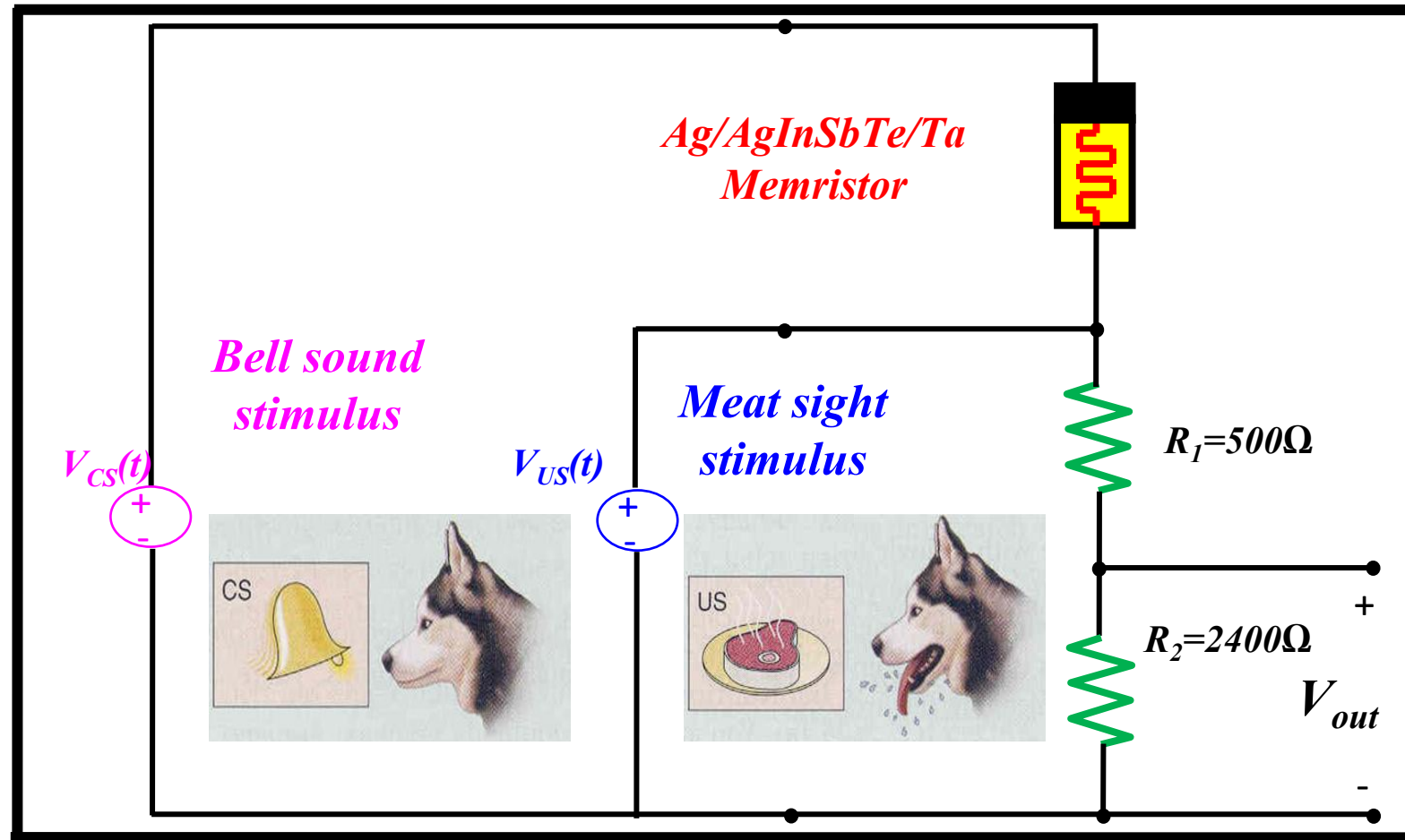
Associative memory and learning is defined as the ability to learn and remember the relationship between unrelated experiences or events.

Associative Learning

*In 350 BC,
Aristotle had already suggested that
learning involves an association of
ideas.*

Emulating Pavlov's Dog Associative Learning Phenomenon

Memristor Circuit for Emulating Pavlov's Dog



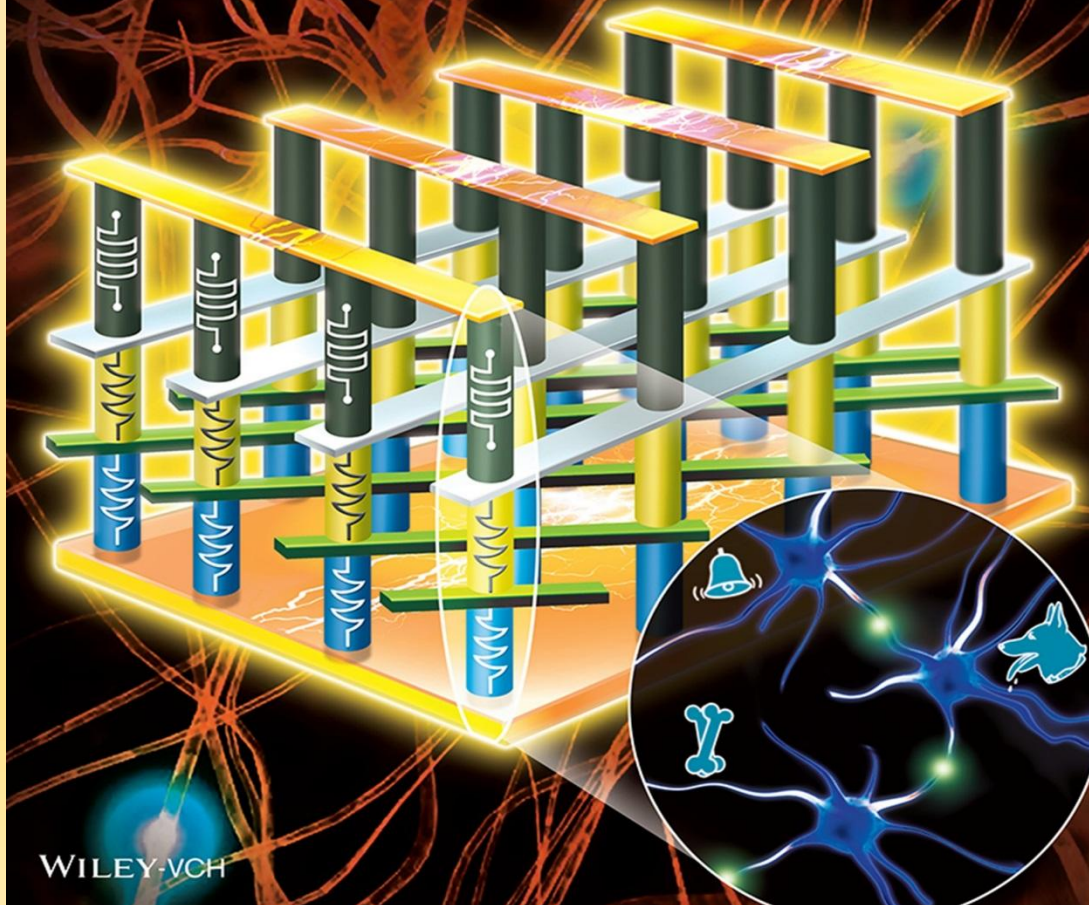
From:

Y. Li, L. Xu, Y. P. Zhong, Y. X. Zhou, S. J. Zhong, Y. Z. Hu, L. O. Chua, X. S. Miao

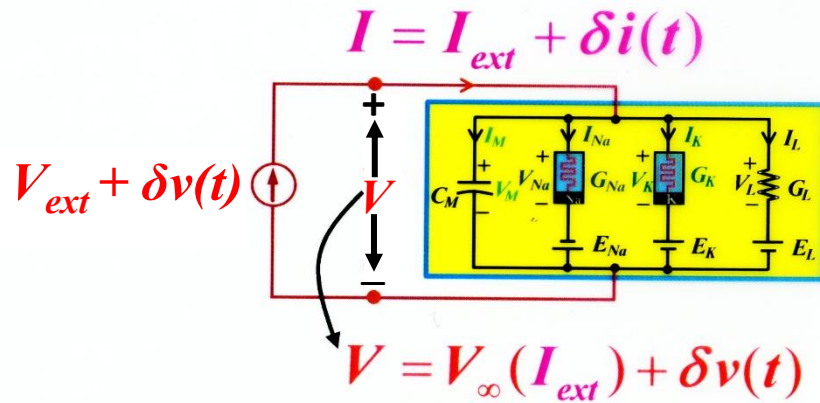
Associative Learning with Temporal Contiguity in a Memristive Circuit for Large-Scale Neuromorphic Networks

Adv. Electronic Materials, 1500125, p.1-8, 2015.

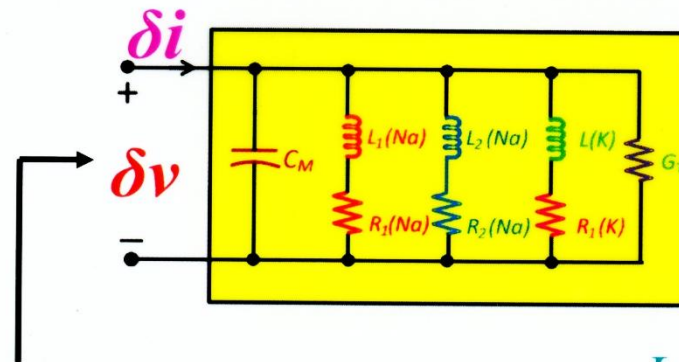
ADVANCED ELECTRONIC MATERIALS



Memristors are promising candidates for applications as artificial synapses in neuromorphic engineering. In article number 1500125, Xiang-Shui Miao and co-workers demonstrate **associative learning** and extinction functions in a compact circuit with only one **memristor** and two resistors. Based on the spike-timing-dependent plasticity rule, learning with temporal contiguity is realized, consistent with biological behaviors. Wuhan_2019-



Hodgkin-Huxley Axon Circuit



Small-Signal equivalent
circuit at DC
(equilibrium point) (V_{ext}, I_{∞})

Admittance: $Y(s ; V_{ext}) \triangleq \frac{L\{\delta i(t)\}}{L\{\delta v(t)\}}$

Complexity function:

$$Y(s ; V_{ext}) = \frac{b_4 s^4 + b_3 s^3 + b_2 s^2 + b_1 s + b_0}{a_3 s^3 + a_2 s^2 + a_1 s + a_0}$$

Main Theorem:

Eigenvalues of the Hodgkin-Huxley 4×4 Jacobian matrix $J_{HH}(I_{ext})$ are identical to the zeros of the complexity function $Y(s; V_{ext})$, i.e. the solution of the scalar polynomial equation.

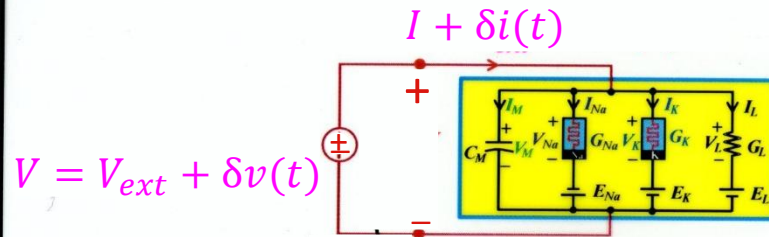
$$b_4 s^4 + b_3 s^3 + b_2 s^2 + b_1 s + b_0 = 0$$

A Glimpse

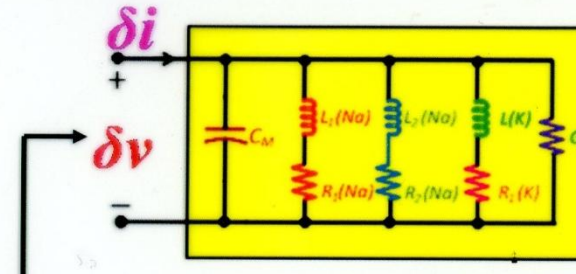
at

Edge of Chaos

Hodgkin-Huxley Axon Circuit



Small-Signal equivalent circuit at DC (equilibrium point) (V_{ext}, I_{∞})



Admittance: $Y(s; V_{ext}) \triangleq \frac{L\{\delta i(t)\}}{L\{\delta v(t)\}}$

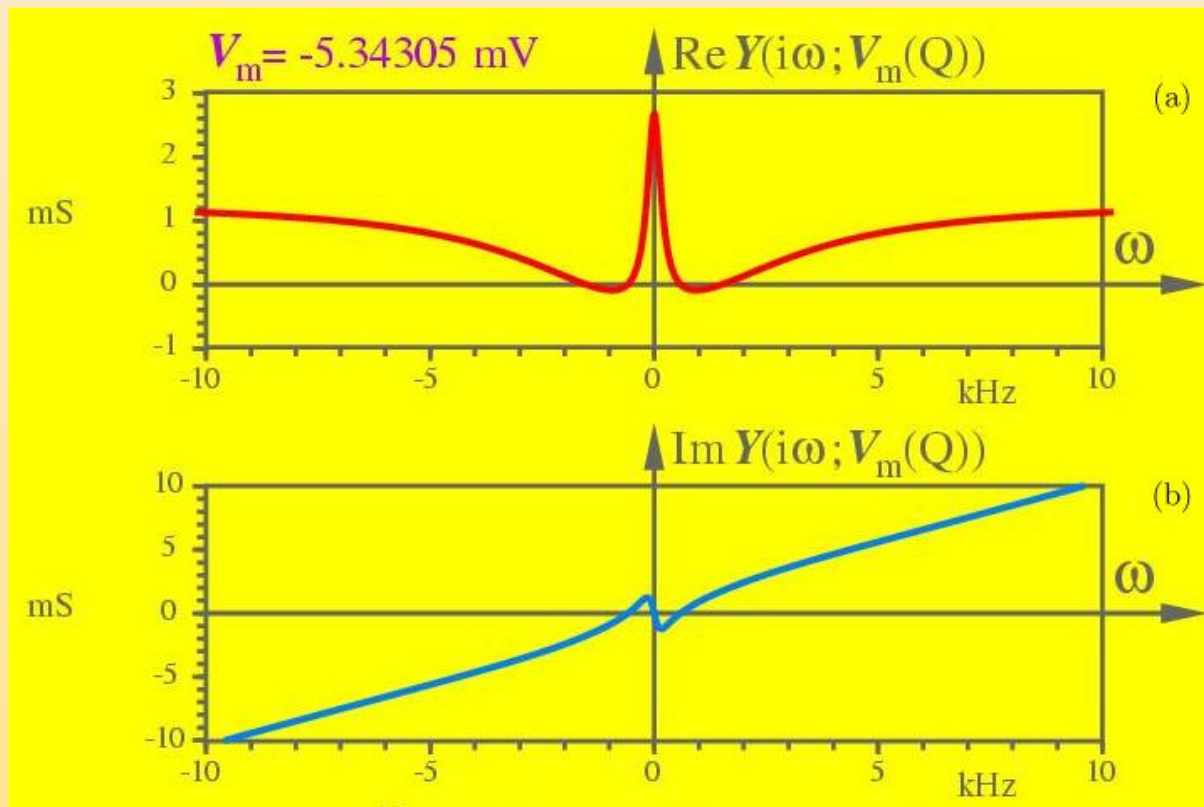
Hodgkin-Huxley Edge of Chaos Criterion:

The *Hodgkin-Huxley Neuron* under synaptic input V_{ext} is on the *edge of chaos* if, and only if,

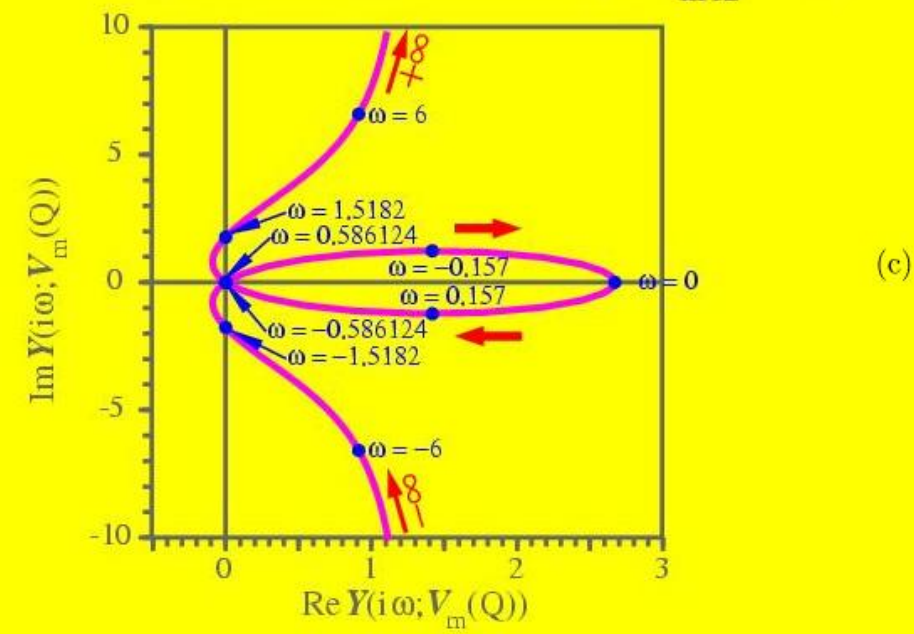
1. All zeros of $Y(s; V_{ext})$ are in the open left half plane,
i. e. $\text{Re } z_k(V_{ext}) < 0 \quad k = 1, 2, 3, 4$
2. $\text{Re } Y(i\omega; V_{ext}) < 0$ for at least one frequency $\omega = \omega_{ext}$

Edge of Chaos is Testable !

Testing
for
Edge of Chaos
Involves only
Linear Algebra



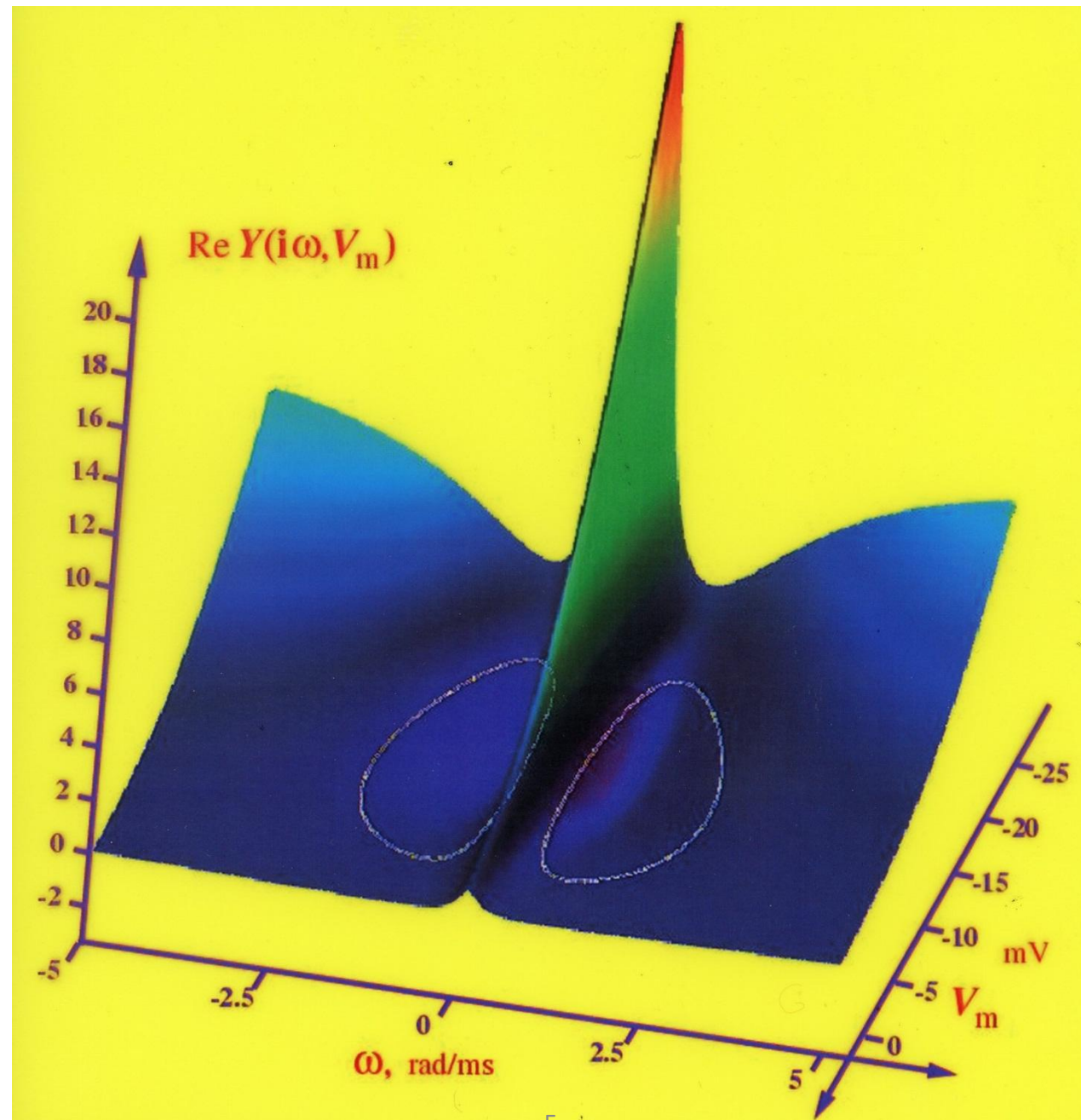
Nyquist Plot



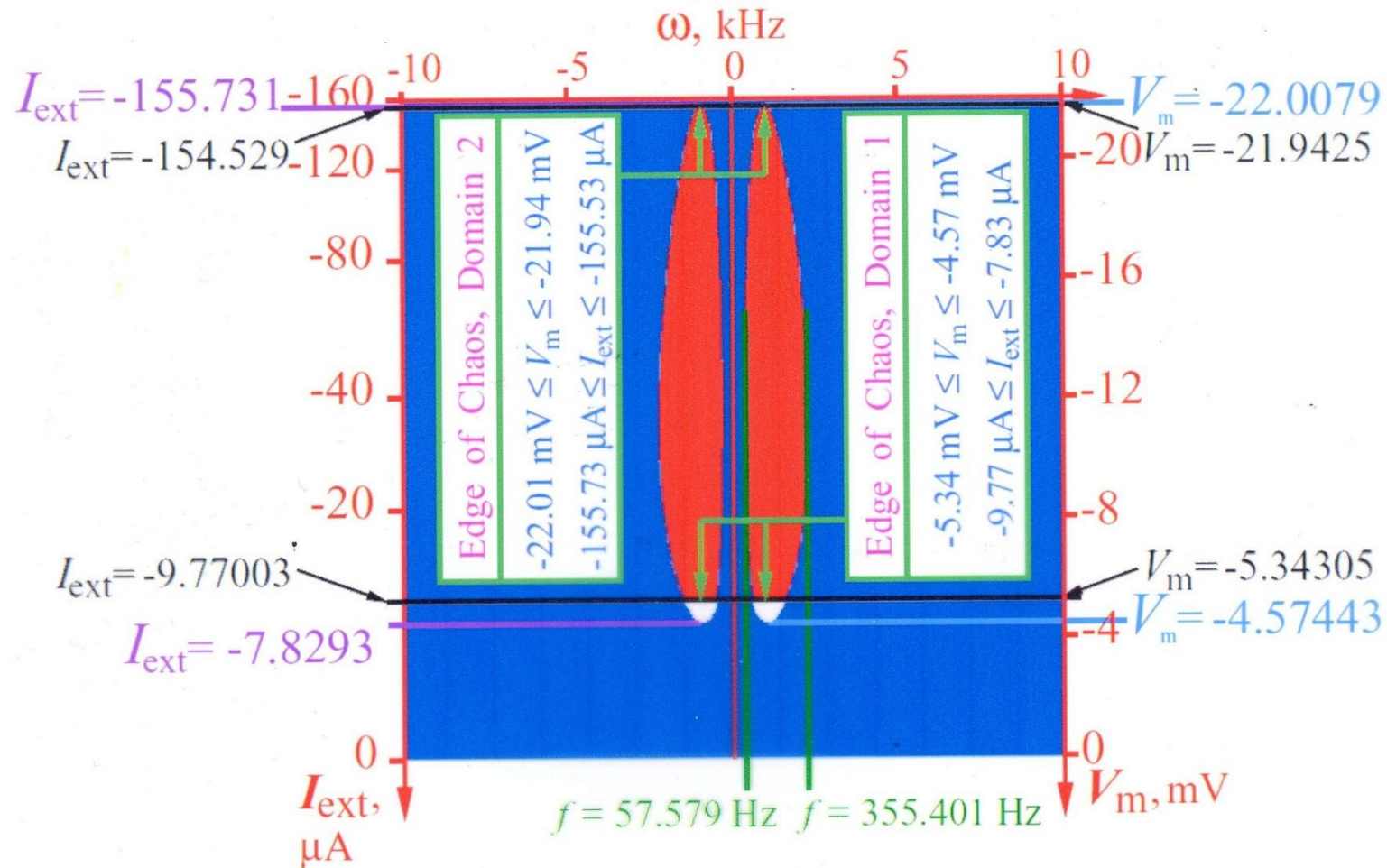
Test for *Edge of Chaos*

Hodgkin Huxley Model is on *Edge of Chaos* at $V = V_m$

if its **Nyquist Plot** intersects the **imaginary axis** $\text{Im } Y(i\omega)$



5-87



Range of Edge of Chaos :

$$-9.77 \mu\text{A} \leq I_{\text{ext}} \leq -7.83 \mu\text{A} \Rightarrow \Delta I_{\text{ext}} = 1.94 \mu\text{A}$$

$$-5.34 \text{ mV} \leq V_m \leq -4.57 \text{ mV} \Rightarrow \Delta V_m = 0.77 \text{ mV}$$

Goldilocks Zone for Action Potential

$$-9.77 \mu A \leq I_{ext} \leq -7.83 \mu A$$

$$-5.34 \text{ mV} \leq V_m \leq -4.57 \text{ mV}$$



Window For Life to be Possible

$$\Delta I < 2 \mu A$$

$$\Delta V < 1 \text{ mV}$$

Galvani's Irritability *Doctrine* is a manifestation of the *Edge of Chaos*, the *crown jewel* of the *Principle of Local Activity*

Galvani was well aware of the *doctrine of "irritability,"* one of the most important conceptual elaborations of **18th century physiology**.

According to *Galvani*, in the framework of *doctrine of irritability*, external electricity was acting as an *excitatory stimulus* (as a *trigger* we would say now) for the **contraction**, and *not as the direct "efficient" cause of the observed phenomena*.

How could it be that such a tiny electrical force would produce muscle contraction if it were not setting in motion some *internal force*, and **triggering** some extremely *mobile principle* existing in nerves which then *excites the action* of the nerve-muscular force?

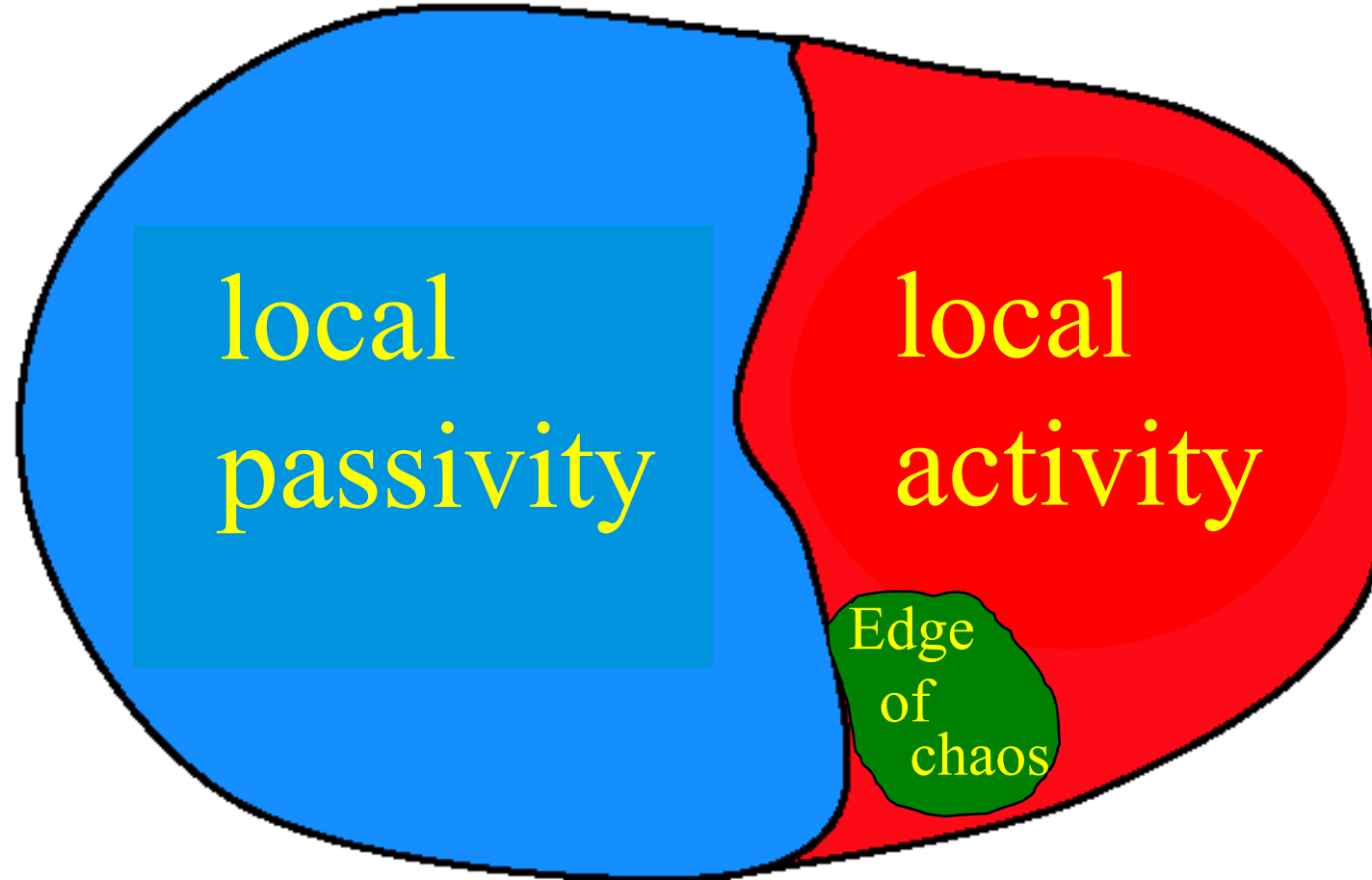
The Local Activity Principle

All 2-terminal electrical devices can be classified into either “*locally passive*”, or “*locally active*”. Occasionally, a tiny subset of “*locally active*” devices is blessed with an “*edge of chaos*” domain capable of exhibiting fascinating phenomena and attributes, such as “*learning*”, “*artificial intelligence*”, and even “*life*” itself.

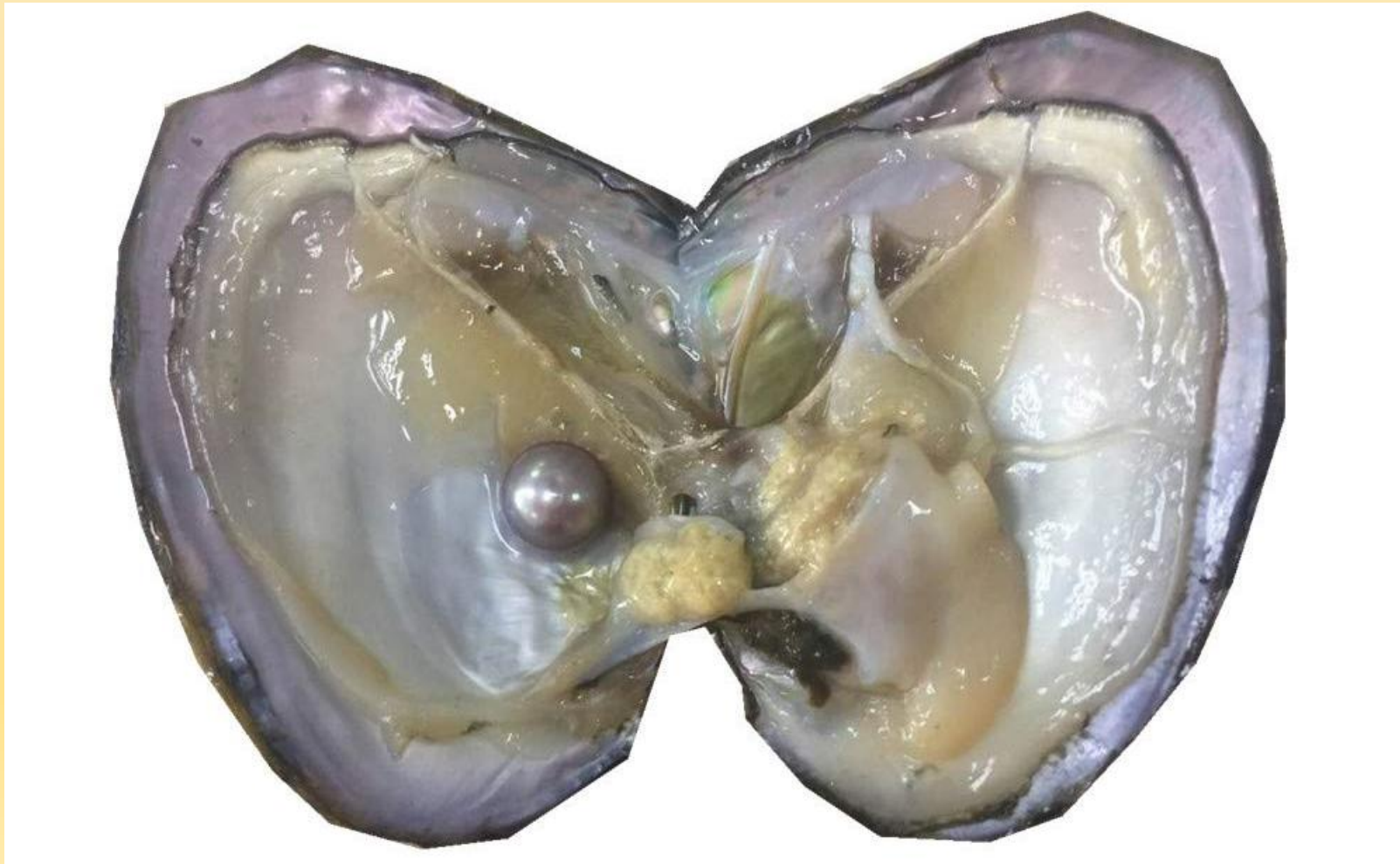
*Proof of the **Principle of Local Activity**
requires non-trivial applications of the following
mathematical results:*

- 1. **Nonlinear Monotone Operator in Function Space***
- 2. **High-dimensional Analog of the Mean-Value Theorem***
- 3. **LaSalle's Complete Stability Theorem***
- 4. **Poisson's Integral on half-plane***
- 5. **Theory of Positive-Real Function of a complex Variable***
- 6. **Tellegen's Theorem***

Local Activity Principle



The “*Edge of Chaos*” is the
“**Pearl**” amidst the
“*locally active domain*”



Local Activity Theorem

A *reaction-diffusion equation* with one diffusion coefficient is *Locally active* at $Q \Leftrightarrow$ any one of the following conditions holds :

1. $Z_Q(s)$ has a pole in $\text{Re } [s] > 0$.
2. $\text{Re } Z_Q(i\omega) < 0$ for at least one $\omega = \omega_0$.
3. $Z_Q(s)$ has a simple pole $s = i\omega_p$ on the imaginary axis i. e.,

$$Z_Q(s) = \frac{Z_1(s)}{(s - i\omega_p)^m Z_2(s)}, \quad m = 1$$

and

$$\lim_{s \rightarrow i\omega_p} (s - i\omega_p) Z_Q(s) < 0.$$

4. $Z_Q(s)$ has a multiple pole on the imaginary axis.

Physical Meaning of Locally Active Cells

*One can use a **locally-active cell** to amplify “**small**” **signals** at the expense of an external power supply.*

*For example, **neurons** in our brain maintain their local levels of organization by burning **glucose**. In fact, every “**living**” cell is a “**molecular**” amplifier.*

UK-57

Fundamental New Result
Edge of Chaos Theorem

*A stable equilibrium state Q can be destabilized
by coupling it to a passive environment, if, and
only if, Q is poised
on the *Edge of chaos**

Action Potential
is
Impossible
without
Edge of Chaos

Neurons
are poised near the
Edge of Chaos

Local Activity

is the

Origin of Complexity

Local Activity Test



Searching A Needle in Haystack

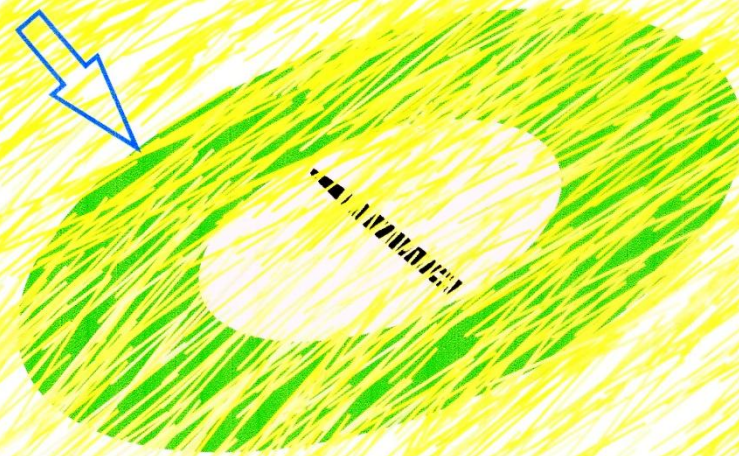
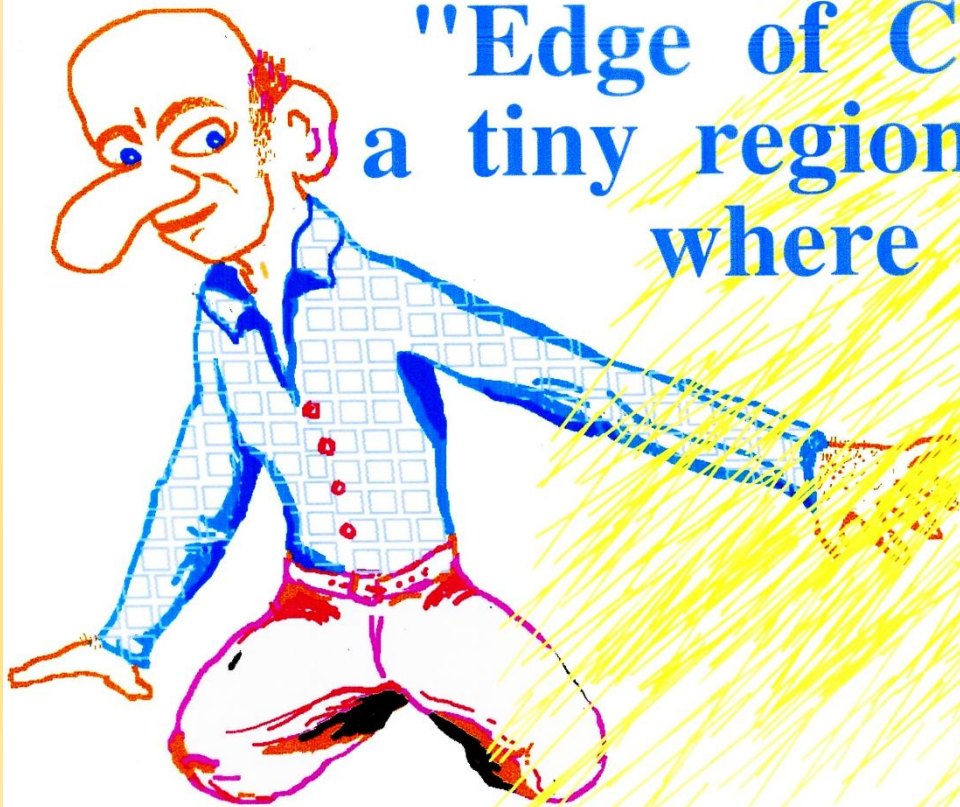
**"Local activity"
tells you
where "not to look"**



Searching A Needle in Haystack

"Local activity"
tells you
where "not to look"

"Edge of Chaos" identifies
a tiny region inside the haystack
where the needle lies



*Following Reversed Engineering Article has just appeared
online in*

NATURE ELECTRONICS

14 May, 2018

How We Predicted the Memristor

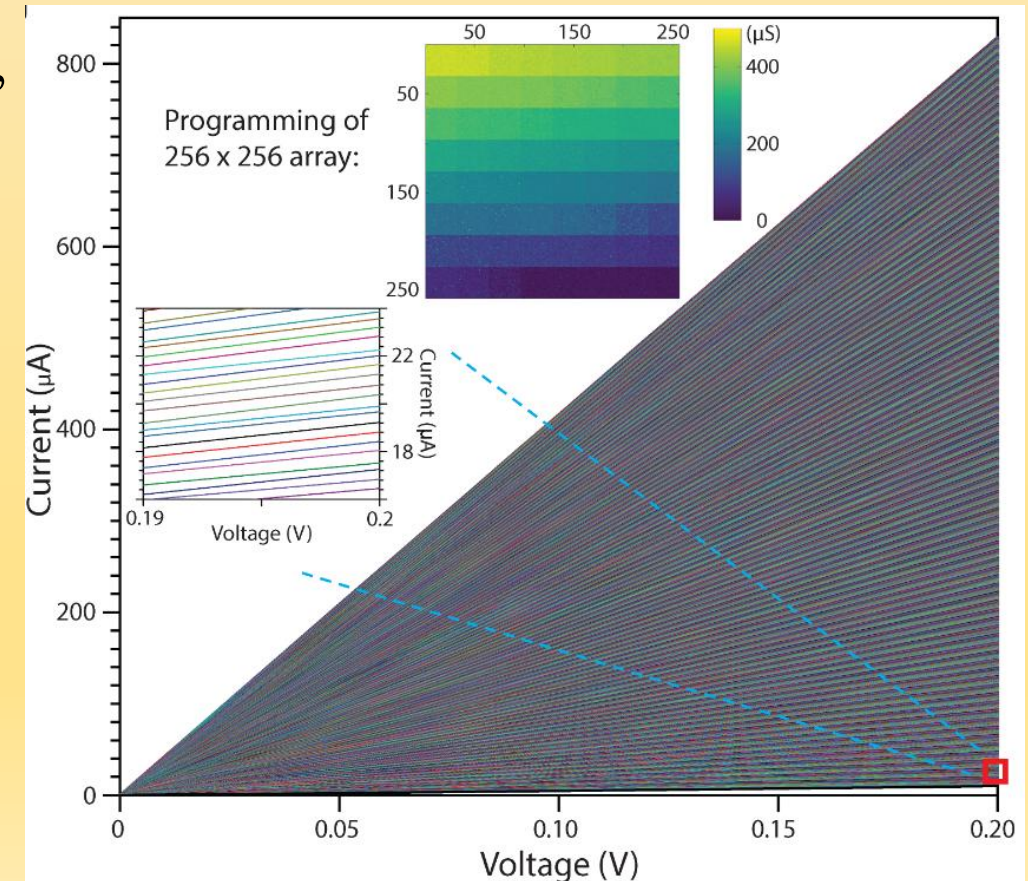
Leon O. Chua

Nature Vol. 615, No. 7954, pp.823-829, **2023**

**Thousands of conductance levels in memristors
monolithically integrated on CMOS**

Mingyi Rao, Hao Tang, Jiangbin Wu, Wenhao Song, Max Zhang, Wenbo Yin, Ye Zhuo, Fatemeh Kiani, Benjamin Chen, Xiangqi Jiang, Hefei Liu, Hung-Yu Chen, Rivu Midya, Fan Ye, Hao Jiang, Zhongrui Wang, Mingche Wu, Miao Hu, Han Wang, Qiangfei Xia, Ning Ge, Ju Li, J. Joshua Yang.

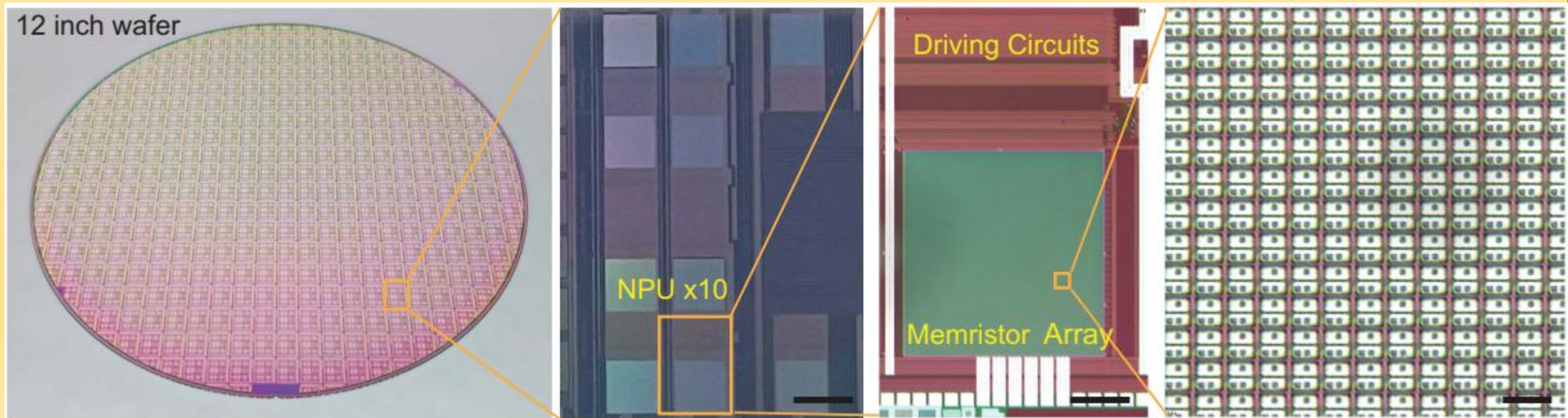
***2048 resistance levels
(11 bit/cell)
in one memristor***



Science: Vol. 383, No. 6685, 804-810, February 2024

Programming memristor arrays with arbitrarily high precision for analog computing

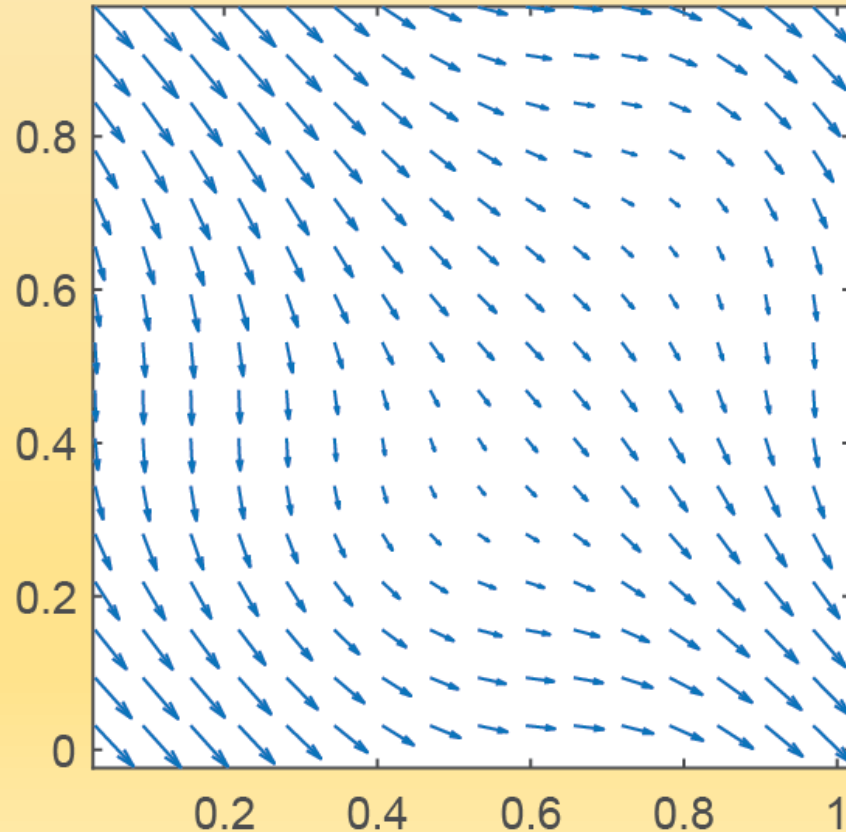
Wenhao Song, Mingyi Rao, Yunning Li, Can Li, Ye Zhuo, Fuxi Cai, Mingche Wu, Wenbo Yin, Zongze Li, Qiang Wei, Sangsoo Lee, Hengfang Zhu, Lei Gong, Mark Barnell, Qing Wu, Peter A. Beerel, Mike Shuo-Wei Chen, Ning Ge, Miao Hu, Qiangfei Xia, J. Joshua Yang



Programming Memristor Array to Solve PDE

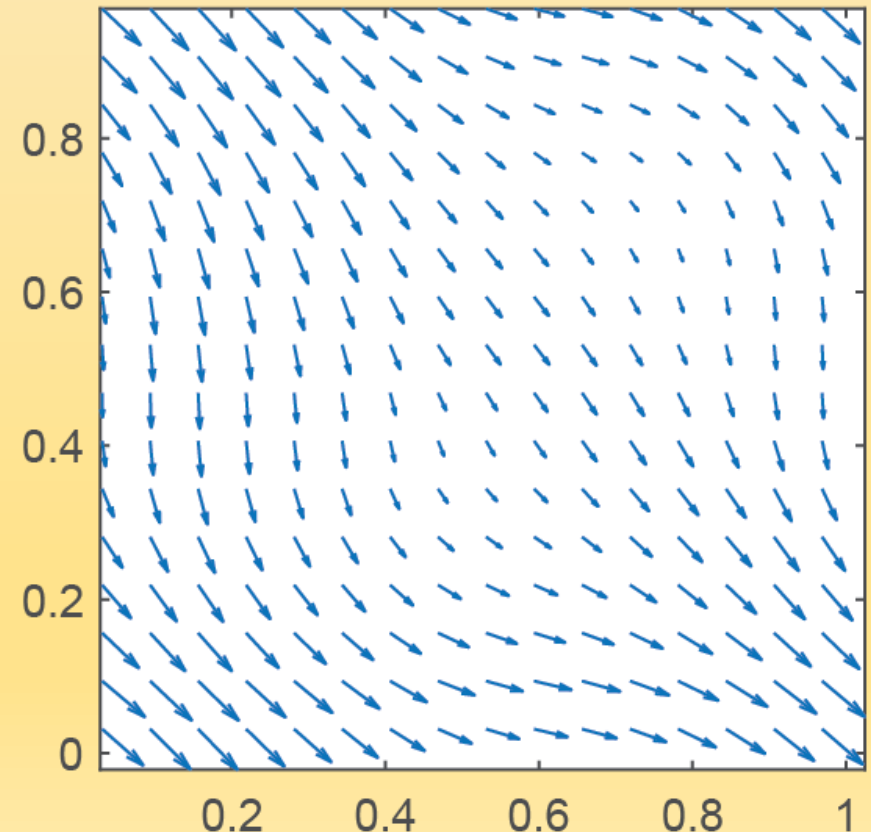
Example: Navier-Stokes Equations

Solved by MATLAB



J-C

Solved by MEMRISTOR



Nature: Vol. 640, 17 April, 2025

The Growing Memristor Industry

<https://doi.org/10.1038/s41586-025-08733-5>

Received: 21 November 2023

Accepted: 3 February 2025

Published online: 16 April 2025

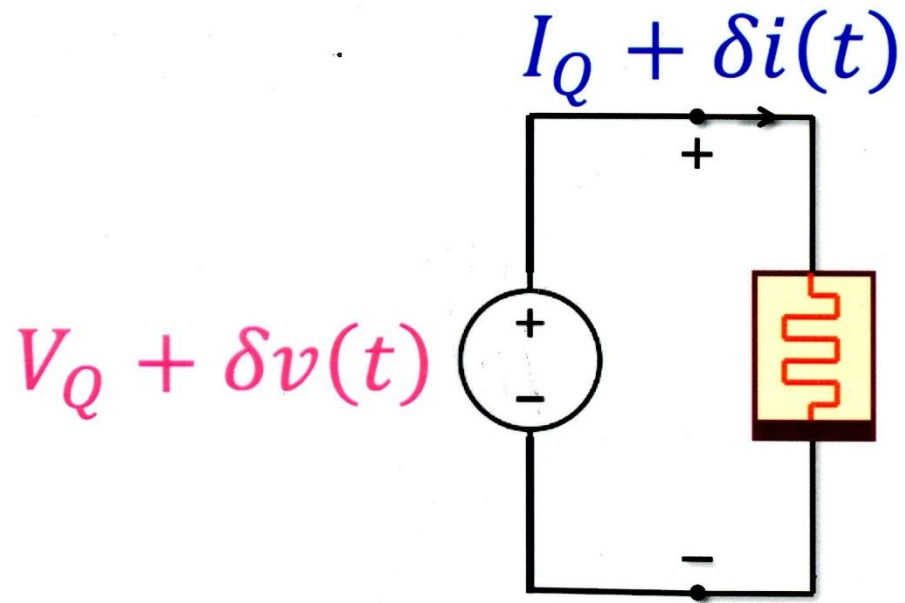


Check for updates

Mario Lanza^{1,2,3,15}✉, Sebastian Pazos^{4,15}, Fernando Aguirre⁵, Abu Sebastian⁶, Manuel Le Gallo⁶, Syed M. Alam⁷, Sumio Ikegawa⁷, J. Joshua Yang⁸, Elisa Vianello⁹, Meng-Fan Chang¹⁰, Gabriel Molas¹¹, Ishai Naveh¹¹, Daniele Ielmini¹², Ming Liu¹³ & Juan B. Roldan¹⁴

The semiconductor industry is experiencing an accelerated transformation to overcome the scaling limits of the transistor and to adapt to new requirements in terms of data storage and computation, especially driven by artificial intelligence applications and the Internet of Things. In this process, new materials, devices, integration strategies and system architectures are being developed and optimized. Among them, memristive devices and circuits—memristors are two-terminal memory devices that can also mimic some basic bioelectronic functions—offer a potential approach to create more compact, energy-efficient or better-performing systems. The memristor industry is growing quickly, raising abundant capital investment, creating new jobs and placing advanced products in the market. Here we analyse the status and prospects of the memristor industry, focusing on memristor-based products that are already commercially available, prototypes with a high technological readiness level that might affect the market in the near future, and discuss obstacles and pathways to their implementation.

Locally Active Memristor



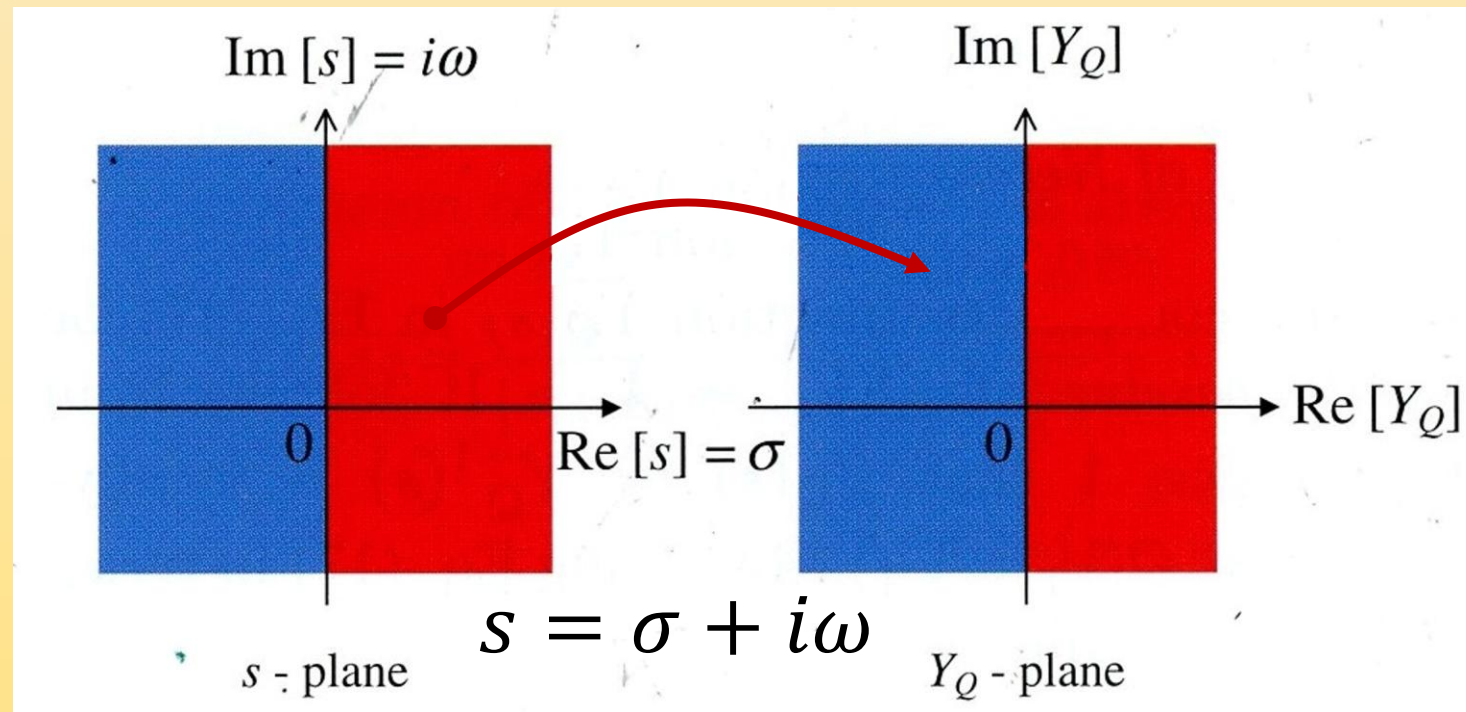
Definition:

A **memristor** is *locally active* at an Equilibrium Point $Q(V_Q, I_Q)$ *if* there exists an **admissible small-signal pair** $(\delta v(t), \delta i(t))$ such that

$$\int_0^T \delta v(\tau) \delta i(\tau) d\tau < 0$$

at some **finite** time T .

Not-Positive Real Function implies **N** is *locally-Active*



Small-Signal
Admittance

$$Y_Q(s) = \frac{I(s)}{V(s)}$$

at DC operating
point **Q**

World War II hero Alan Turing's saved papers sell for record price at UK auction



A treasure trove of wartime codebreaker Alan Turing's scientific papers has sold for more than three times their expected sale price... All after being saved from the shredder.

A collection of rare scientific papers written by mathematician, computer scientist and Second World War codebreaker Alan Turing has sold for a record £465,400 (€544,400) at auction in Lichfield, UK.

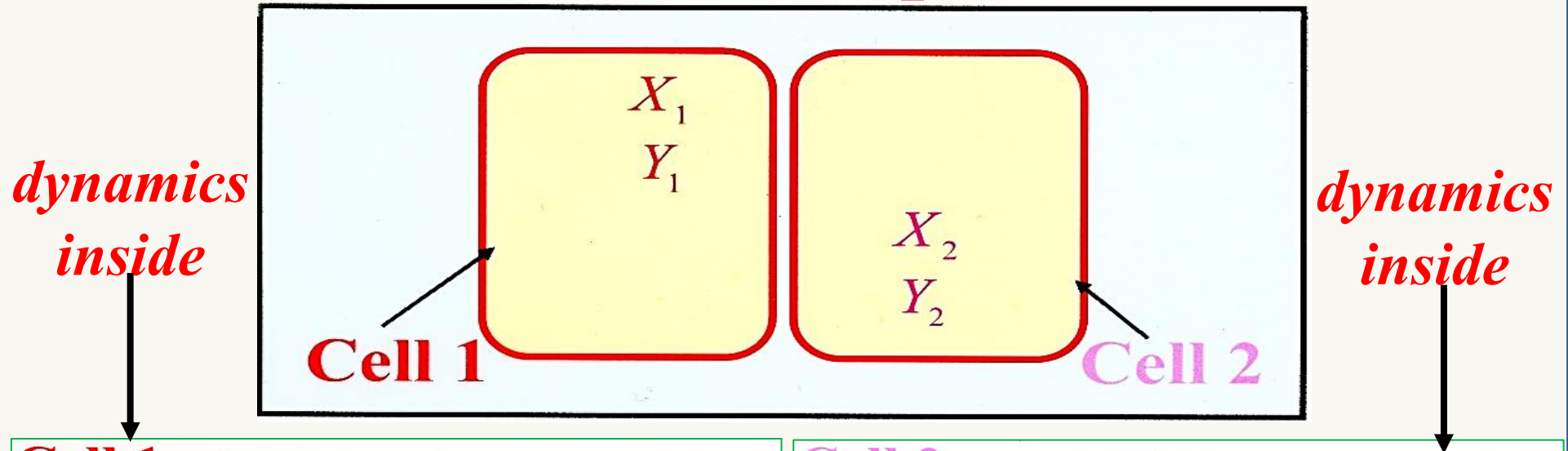
The documents were discovered in a loft at a property in Bermondsey, London, and were almost destroyed during a house clearance.

Some of the rare items that went under the hammer included a personal signed copy of Turing's 1938 PhD dissertation, "Systems of Logic Based on Ordinals", which sold for £110,500 (€129,200), as well as his paper "On Computable Numbers" - also known as "Turing's Proof" - which introduced the world to the idea of a universal computing machine in 1936.

The collection also included "The Chemical Basis Of Morphogenesis", which sold for £19,500 (€22,800). Dating from 1952, it is Turing's last major published work.

What is TURING INSTABILITY ?

2 Identical Uncoupled Cells



Cell 1

$$\dot{X}_1 = (5X_1 - 6Y_1 + 1)$$

$$\dot{Y}_1 = (6X_1 - 7Y_1 + 1)$$

Equilibrium Solution of Cell 1:

$$X_1 = 1, Y_1 = 1$$

Cell 2

$$\dot{X}_2 = (5X_2 - 6Y_2 + 1)$$

$$\dot{Y}_2 = (6X_2 - 7Y_2 + 1)$$

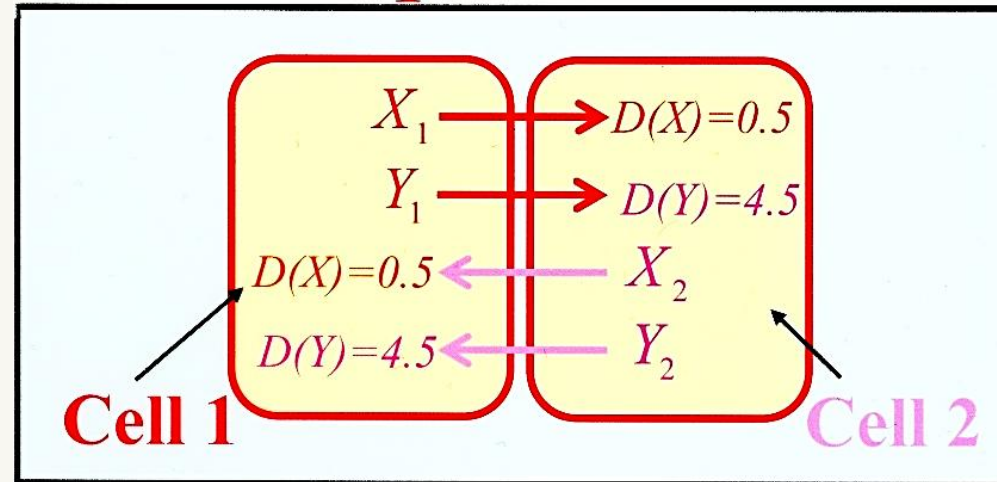
Equilibrium Solution of Cell 2:

$$X_2 = 1, Y_2 = 1$$

TURING INSTABILITY

Emerges from the Reaction-Diffusion Equations of 2 coupled identical cells

(X_1, Y_1)
diffuses to
Cell 2



(X_2, Y_2)
diffuses to
Cell 1

Diffusion is Dissipation !

Cell 1

$$\dot{X}_1 = (5X_1 - 6Y_1 + 1) + 0.5(X_2 - X_1)$$

$$\dot{Y}_1 = (6X_1 - 7Y_1 + 1) + 4.5(Y_2 - Y_1)$$

Cell 2

$$\dot{X}_2 = (5X_2 - 6Y_2 + 1) + 0.5(X_1 - X_2)$$

$$\dot{Y}_2 = (6X_2 - 7Y_2 + 1) + 4.5(Y_1 - Y_2)$$

TURING's Two-Cell Reaction-Diffusion Equations

$$\underbrace{\begin{bmatrix} 5 & -6 & 0 & 0 \\ 6 & -7 & 0 & 0 \\ 0 & 0 & 5 & -6 \\ 0 & 0 & 6 & -7 \end{bmatrix}}_{\mathbf{G}_1} + \underbrace{\begin{bmatrix} -0.5 & 0 & 0.5 & 0 \\ 0 & -4.5 & 0 & 4.5 \\ 0.5 & 0 & -0.5 & 0 \\ 0 & 4.5 & 0 & -4.5 \end{bmatrix}}_{\mathbf{G}_2} = \underbrace{\begin{bmatrix} 4.5 & -6 & 0.5 & 0 \\ 6 & -11.5 & 0 & 4.5 \\ 0.5 & 0 & 4.5 & -6 \\ 0 & 4.5 & 6 & -11.5 \end{bmatrix}}_{\mathbf{G} = \mathbf{G}_1 + \mathbf{G}_2}$$

Eigen Values of \mathbf{G}_1

\mathbf{G}_1
Stable
Matrix

$$\begin{aligned} \lambda_1(\mathbf{G}_1) &= -1 \\ \lambda_2(\mathbf{G}_1) &= -1 \\ \lambda_3(\mathbf{G}_1) &= -1 \\ \lambda_4(\mathbf{G}_1) &= -1 \end{aligned}$$

Eigen Values of \mathbf{G}_2

\mathbf{G}_2
Stable
Matrix

$$\begin{aligned} \lambda_1(\mathbf{G}_2) &= -9 \\ \lambda_2(\mathbf{G}_2) &= -1 \\ \lambda_3(\mathbf{G}_2) &= 0 \\ \lambda_4(\mathbf{G}_2) &= 0 \end{aligned}$$

A FOLK Theorem

Most engineers and mathematicians who are not specialists in matrix algebra believe in the following proposition:

Adding 2 stable matrices
gives another stable matrix.

TURING Instability: A shocking

TURING's Two-~~Phenomenon~~ Reaction-Diffusion Equations

$$\underbrace{\begin{bmatrix} 5 & -6 & 0 & 0 \\ 6 & -7 & 0 & 0 \\ 0 & 0 & 5 & -6 \\ 0 & 0 & 6 & -7 \end{bmatrix}}_{\mathbf{G}_1} + \underbrace{\begin{bmatrix} -0.5 & 0 & 0.5 & 0 \\ 0 & -4.5 & 0 & 4.5 \\ 0.5 & 0 & -0.5 & 0 \\ 0 & 4.5 & 0 & -4.5 \end{bmatrix}}_{\mathbf{G}_2} = \underbrace{\begin{bmatrix} 4.5 & -6 & 0.5 & 0 \\ 6 & -11.5 & 0 & 4.5 \\ 0.5 & 0 & 4.5 & -6 \\ 0 & 4.5 & 6 & -11.5 \end{bmatrix}}_{\mathbf{G} = \mathbf{G}_1 + \mathbf{G}_2}$$

Eigen Values of \mathbf{G}_1

\mathbf{G}_1
Stable
Matrix

$$\begin{aligned} \lambda_1(\mathbf{G}_1) &= -1 \\ \lambda_2(\mathbf{G}_1) &= -1 \\ \lambda_3(\mathbf{G}_1) &= -1 \\ \lambda_4(\mathbf{G}_1) &= -1 \end{aligned}$$

Eigen Values of \mathbf{G}_2

\mathbf{G}_2
Stable
Matrix

$$\begin{aligned} \lambda_1(\mathbf{G}_2) &= -9 \\ \lambda_2(\mathbf{G}_2) &= -1 \\ \lambda_3(\mathbf{G}_2) &= 0 \\ \lambda_4(\mathbf{G}_2) &= 0 \end{aligned}$$

Eigen Values of \mathbf{G}

\mathbf{G}
Unstable
Matrix

$$\begin{aligned} \lambda_1(\mathbf{G}) &= -14 \\ \lambda_2(\mathbf{G}) &= -1 \\ \lambda_3(\mathbf{G}) &= -1 \\ \lambda_4(\mathbf{G}) &= +2 \end{aligned}$$

What Special Property
must G_1 possess
to make
destabilization
possible ?

Mem-16

Various forms of *Turing's equations*, or *reaction-diffusion equations* have appeared in one form or another in many works and fields. However, any sort of systematic understanding or analysis seems far away. Before one can expect any general understanding, many examples will have to be thought through, both on the mathematical side and on the experimental side.

A MATHEMATICAL MODEL OF TWO CELLS
VIA TURING'S EQUATION
BY
S. SMALE

Edge of Chaos is Sine Qua Non for Turing Instability

A. Ascoli *Senior Member, IEEE*, A.S. Demirkol, R. Tetzlaff *Senior Member, IEEE*, and L. Chua *Fellow, IEEE*

Abstract—Diffusion-driven instabilities with pattern formation may occur in a network of identical, regularly-spaced, and resistively-coupled cells if and only if the uncoupled cell is poised on a locally-active and stable operating point in the Edge of Chaos domain. This manuscript presents the simplest ever-reported two-cell neural network, combining together only 7 two-terminal components, namely 2 batteries, 3 resistors, and 2 volatile NbO_x memristive threshold switches from NaMLab, and subject to diffusion-driven instabilities with the concurrent emergence of Turing patterns. Very remarkably, this is the first time an homogeneous cellular medium, with no other dynamic element than 2 locally-active memristors, hence the attribute *all-memristor* coined to address it in this paper, is found to support complex phenomena. The destabilization of the homogeneous solution occurs in this second-order two-cell array if and only if the uncoupled cell circuit parameters are chosen from the Edge of Chaos domain. A deep circuit- and system-theoretic investigation, including linearization analysis and phase portrait investigation, provides a comprehensive picture for the local and global dynamics of the bio-inspired network, revealing how a theory-assisted approach may guide circuit design with inherently non-linear memristive devices.

Index Terms—Smale Paradox, Turing Instability, Prigogine Symmetry-Breaking, Destabilization of the Homogeneous, Emergent Phenomena, Pattern Formation, Two-Cell Reaction-Diffusion System, Cellular Nonlinear/Neural/Nanoscale Network, Bio-Inspired Memristor Oscillator, NaMLab NbO_x Memristor, Threshold Switch, Theory of Local Activity, Edge of Chaos Principle

implementation of computing paradigms, which could resolve the current limitations of purely-CMOS electronic systems. Moreover, some volatile devices, called threshold switches in the device physics community, are blessed with the extraordinary capability to act as local energy sources under suitable polarization, a property which earns them the attribute of *locally-active* [7]. Since, similarly as the sodium and potassium ion channels in neuronal axon membranes [8], devices of this kind feature a negative differential resistance (NDR) branch along their DC current-voltage characteristic, a signature for their small-signal amplification capability, their adoption in circuit design allows to design more reliable electronic neuron implementations [9], and to develop artificial cellular neural networks, which, processing information according to biological principles, enable the solution of challenging computational problems more efficiently than current software- or hardware-based solutions [10]. Recently [11], the adoption of locally-active NbO_x memristors [12] from NaMLab [13] has enabled to develop the simplest ever-reported reaction-diffusion network, employing only four dynamic states to reproduce the counterintuitive phenomenon, which mesmerised the American luminary Stephen Smale in 1974 [14], as he reported how two biological cells, mathematically-dead on their own, were unexpectedly found to pulse indefinitely under diffusive coupling. As a follow-up study, this manuscript

Turing's
instability via
dissipation phenomenon
originates from the
Edge of Chaos

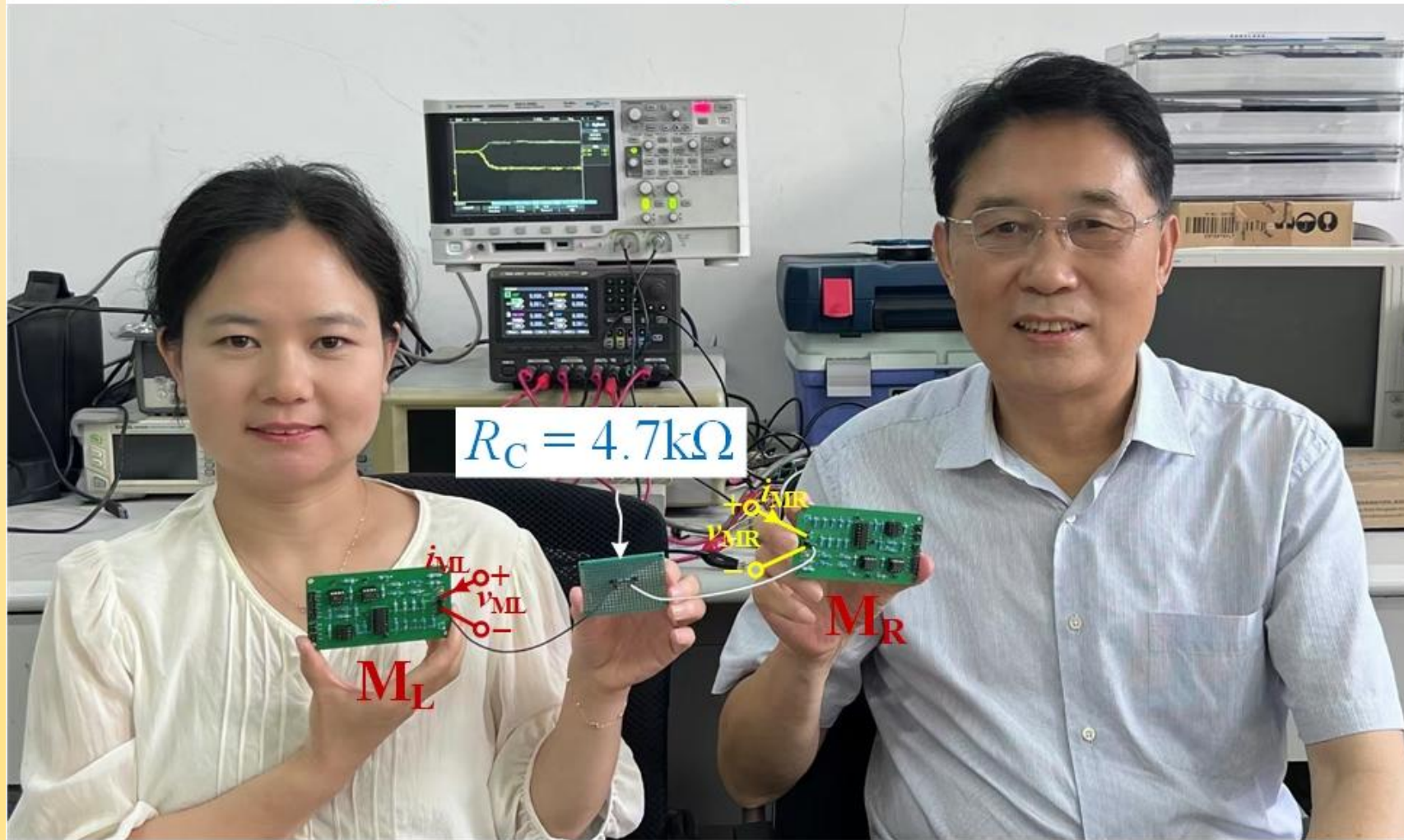
Chemists have pointed out to me that interpreting the reaction **R** to be an "**open system**" makes the model more acceptable.

A MATHEMATICAL MODEL OF TWO CELLS
VIA TURING'S EQUATION
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Chemists have pointed out to me that interpreting the reaction **R** to be an "**open system**" makes the model more acceptable.

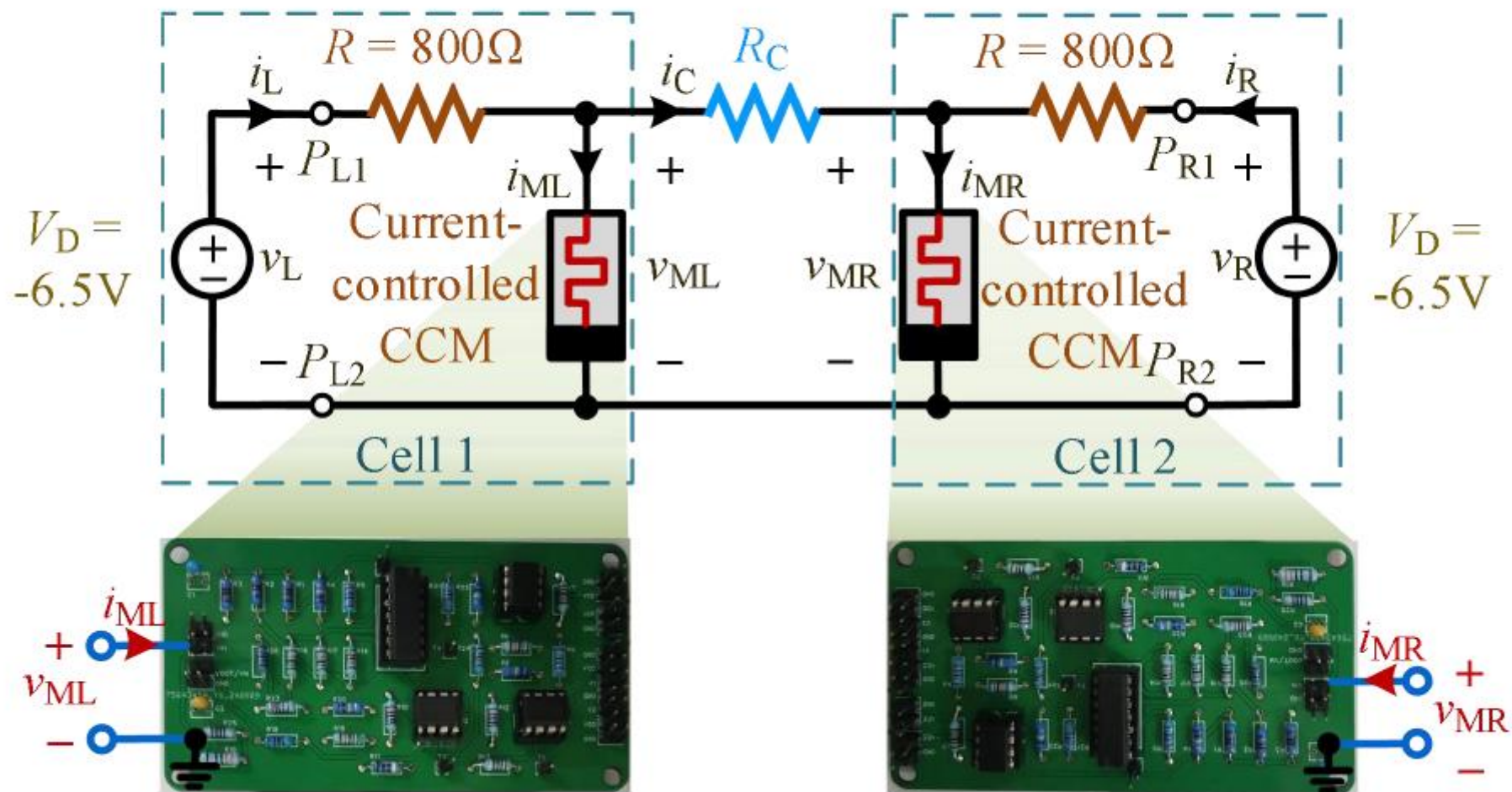
In my Talk today, **R** is a
locally-active System
operating on
Edge of Chaos

Turing Instability Demonstration



Guangyi Wang (right) and Peipei Jin (left)

Hardware Demonstration of Turing Instability



Current-Controlled Chua Corsage Memristor(CCM)

State-Dependent
Ohm's Law

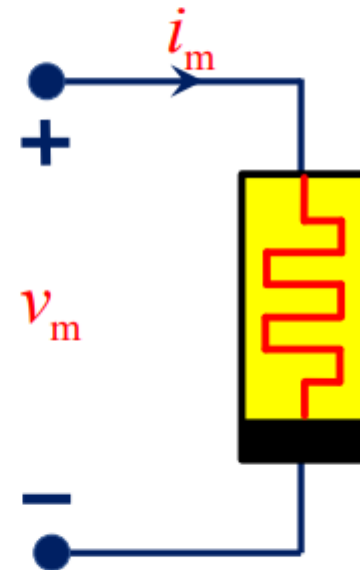
$$v_m = (20x^2)i_m$$

State Equation

$$\frac{dx}{dt} = h(x, i_m)$$

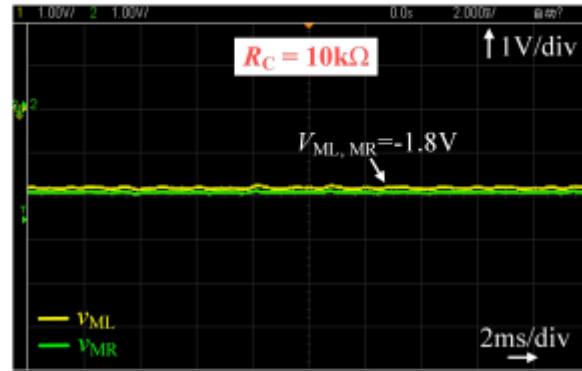
where

$$h(x, i_m) = 10^4 (30 - x + |x - 20| - |x - 40| + 10^3 i_m)$$

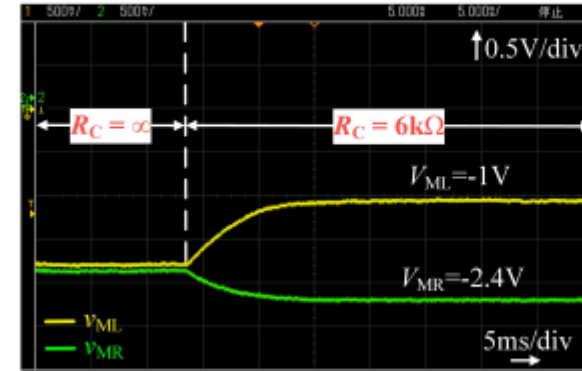


Experimental Results of Turing Instability

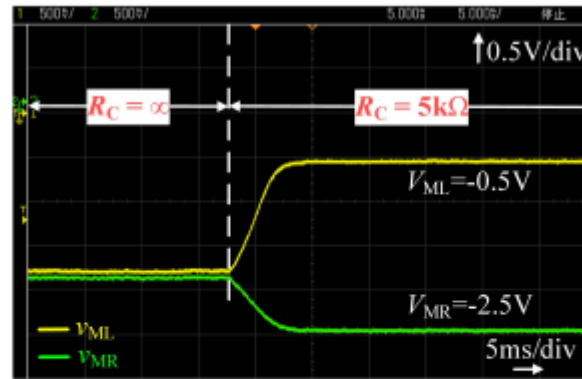
$R_C = 10\text{k}\Omega > 7.1\text{k}\Omega$



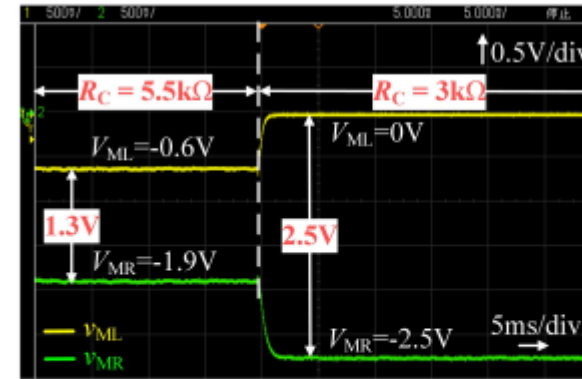
$R_C = 6\text{k}\Omega < 7.1\text{k}\Omega$



$R_C = 5\text{k}\Omega < 7.1\text{k}\Omega$



$R_C = 5.5\text{k}\Omega \rightarrow 3\text{k}\Omega$



Live hardware demonstration of Turing Instability. Experimentally measured memristor voltages $v_{ML}(t)$ and $v_{MR}(t)$ of the left and right cells in the coupled circuit with $R_C = 10\text{k}\Omega$, $R_C = \infty \rightarrow 6\text{k}\Omega$, $R_C = \infty \rightarrow 5\text{k}\Omega$, and $R_C = 5.5\text{k}\Omega \rightarrow 3\text{k}\Omega$.

There is a paradoxical aspect to the example. One has two dead (mathematically dead) cells interacting by a diffusion process which has a tendency in itself to equalize the concentrations. Yet in interaction, a state continues to pulse indefinitely.

A MATHEMATICAL MODEL OF TWO CELLS
VIA TURING'S EQUATION
BY
S. SMALE

Edge of Chaos Theory Resolves Smale Paradox

Alon Ascoli^{ID}, *Member, IEEE*, Ahmet Samil Demirkol^{ID}, Ronald Tetzlaff, *Senior Member, IEEE*,
and Leon Chua^{ID}, *Life Fellow, IEEE*

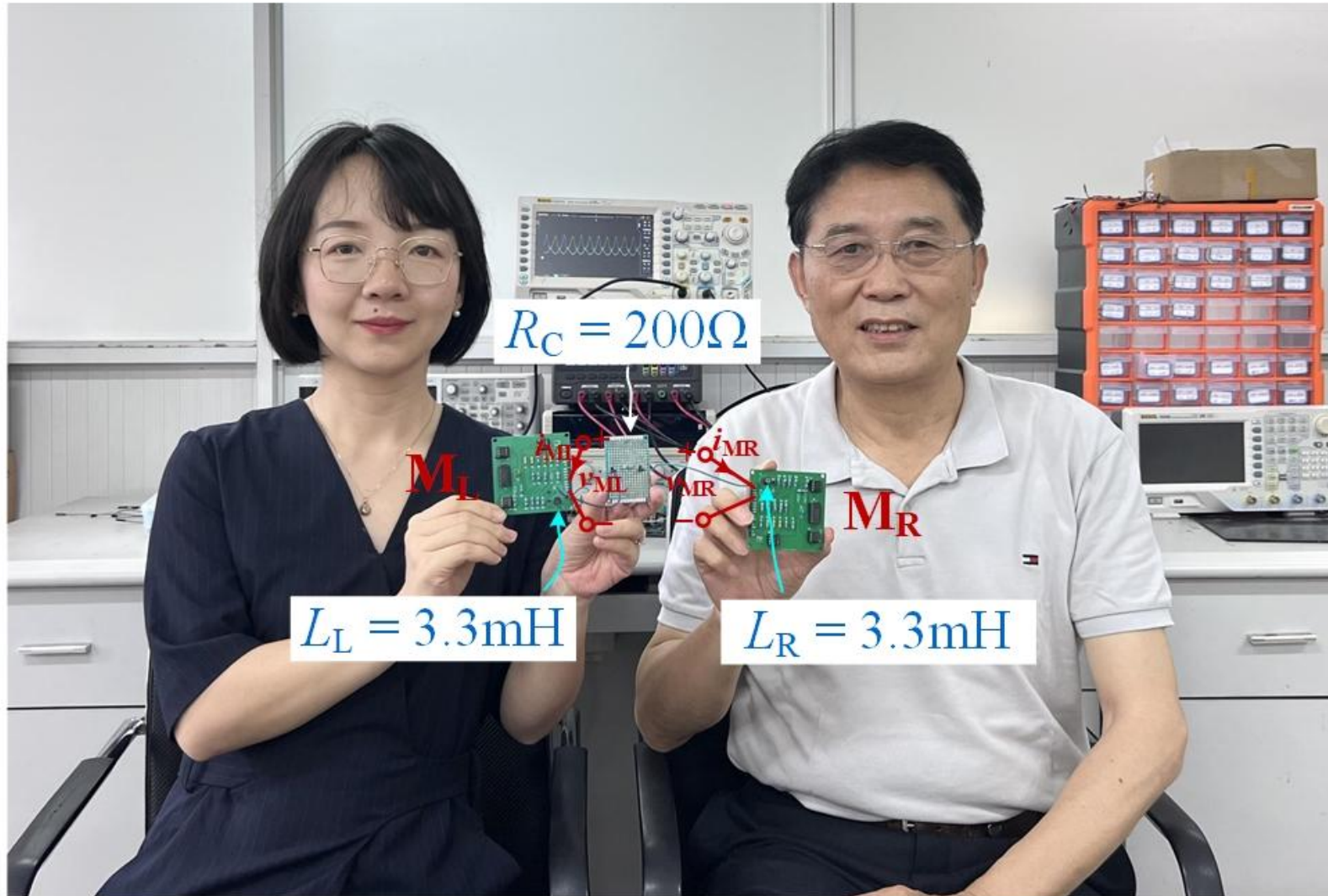
Abstract—No isolated system may ever support complexity. Emergent phenomena may however appear in an open system, if, as established by the Edge of Chaos theory, some of its constitutive elements feature the capability to amplify infinitesimal fluctuations in energy, provided an external source supplies them with a sufficient amount of DC power, which is known to be a signature for locally-active behaviour. In particular, complex behaviours, including static and dynamic pattern formation, may emerge in arrays of identical diffusively-coupled cells, if and only if the basic unit is poised on a particular sub-domain of the Local Activity regime, referred to as Edge of Chaos, within which a quiet state hides in fact a high degree of excitability. Here we show, for the first time, that these counterintuitive phenomena may emerge in a basic memristor cellular neural network, consisting of two identical diffusively-coupled second-order cells. The proposed bio-inspired array represents the simplest ever-reported open system, which reproduces the shocking phenomenon, reported by Smale in 1974, when, while studying a model from cellular biology, he observed two identical reaction cells, “mathematically dead” on their own, pulsating together upon diffusive coupling. Impressively, the bio-inspired two-cell reaction-diffusion network contains only nine circuit elements, specifically two DC voltage sources, three linear resistors, two linear capacitors, and two functional niobium oxide (NbO) memristors from NaMLab. Applying the theory of Local Activity to an accurate model of the memristor oscillator, a comprehensive picture for its local and global dynamics may be drawn, providing a systematic method to tune the design parameters of the two-cell array to enable diffusion-driven instabilities therein.

Index Terms—Smale paradox, turing instability, Prigogine symmetry-breaking, destabilization of the homogeneous, emergent phenomena, pattern formation, two-cell reaction-diffusion system, cellular nonlinear/neural/nanoscale network, bio-inspired memristor oscillator, NaMLab niobium oxide (NbO) memristor, theory of Local Activity, Edge of Chaos principle.

lies within a particular sub-domain of the Local Activity regime, referred to as *Edge of Chaos*. A neuronal axon in either the brain [3] or the heart [4], for example, may be modelled as a one-dimensional array of identical diffusively-coupled Hodgkin-Huxley cells. The propagation of an all-or-nothing spike, known as action potential, through this transmission line – an emergent phenomenon “par excellence” [5] – may occur when the single Hodgkin-Huxley cell, hosting biological memristors, namely the sodium and potassium ion channels [6], which are capable to amplify infinitesimal fluctuations in energy under suitable bias conditions, is poised near the Edge of Chaos.

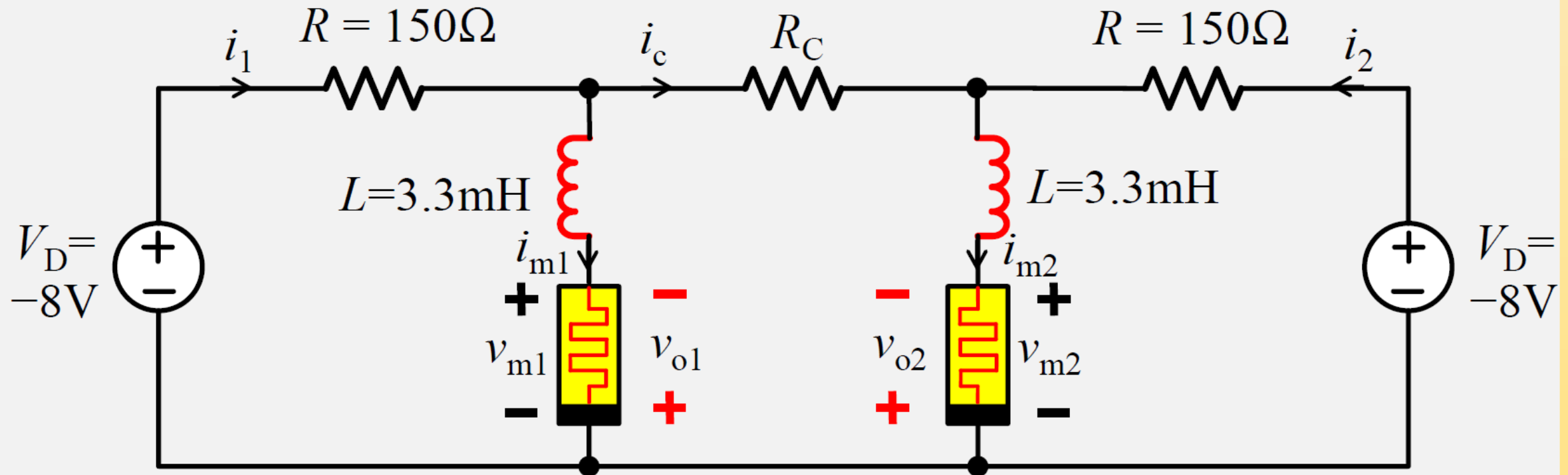
The upcoming end of Moore’s era calls for the development of new materials, processing paradigms, and computing architectures, which may allow to keep the integrated circuit performance growth in the years to follow, despite further CMOS transistor size reduction will no longer be feasible. The disruptive memristive nano-technologies ([7], [8]) offer the key to open up unprecedented opportunities in circuit design. Besides their extensive adoption in crossbar configuration for storing data [9] or for accelerating machine learning algorithms [10], which is of great appeal to the electronics industry, especially for artificial intelligence applications [11], non-volatile (volatile) memristor physical realisations may also be employed to design plausible electronic implementations of biological synapses [12] (neurons [13], [14]). Gaining a deep insight into the nonlinear dynamics of these memristors [15] is a fundamental preliminary step toward the ultimate goal to synthesise neuromorphic circuits capable to resemble as close as possible biological neural networks [16]. A compre-

Smale Paradox Demonstration



Guangyi Wang (right) and Yan Liang (left)

Coupled Circuit to Demonstrate Smale paradox

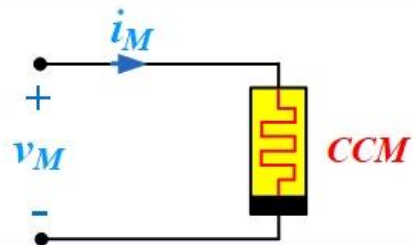


Chua Corsage Memristor (CCM)

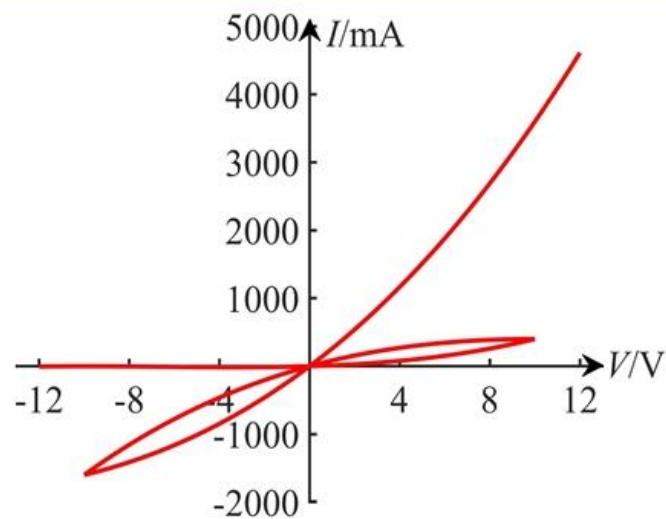
State-Dependent
Ohm's Law $i_M = 10^{-4} x^2 v_M$

State Equation $\frac{dx}{dt} = g(x, v_M)$

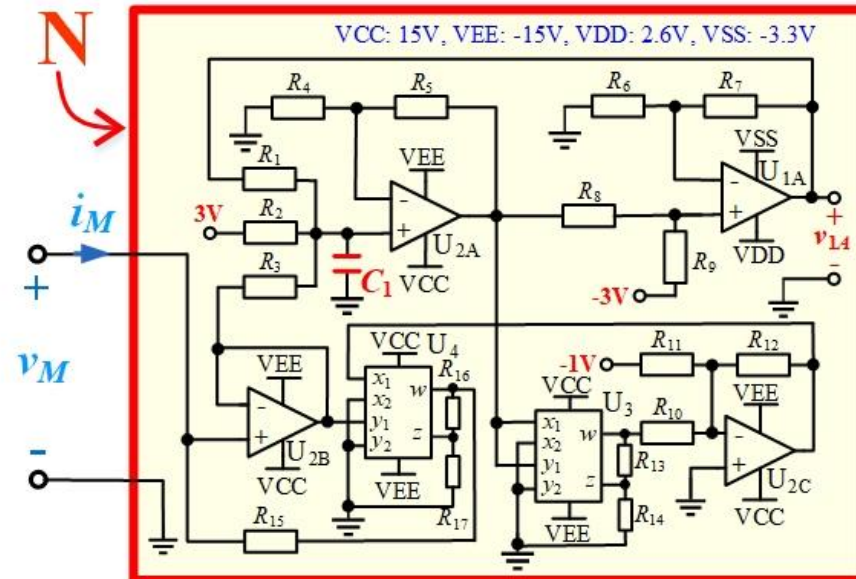
where $g(x, v_M)$ is defined by Eq. (60b)



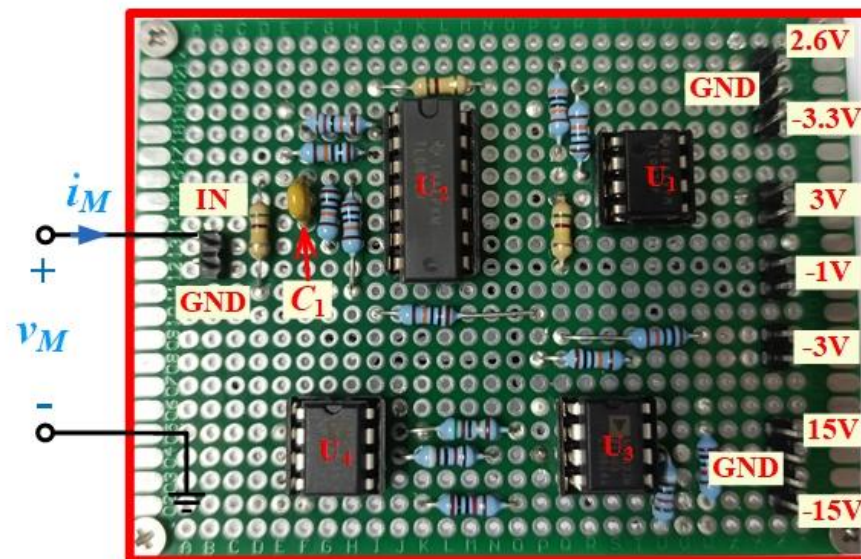
(a) Equations defining CCM



(b) DC V - I curve calculated from (a)

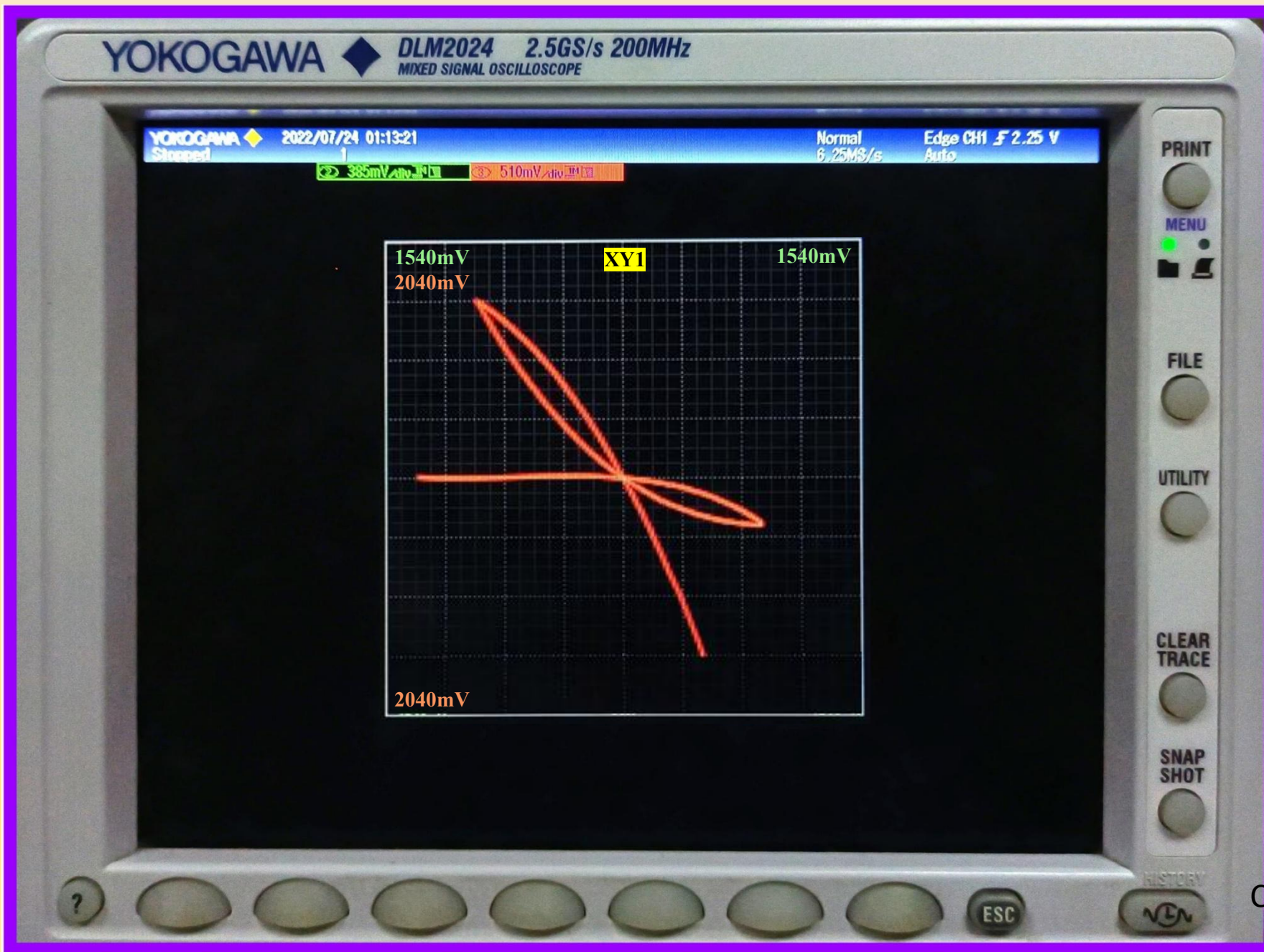


(c) Schematic diagram



(d) Circuit board implementing (c)

CCM-IJBC-16



CCM-IJBC-20

Corsage Ribbon decorating dress of *Marie Antoinette*



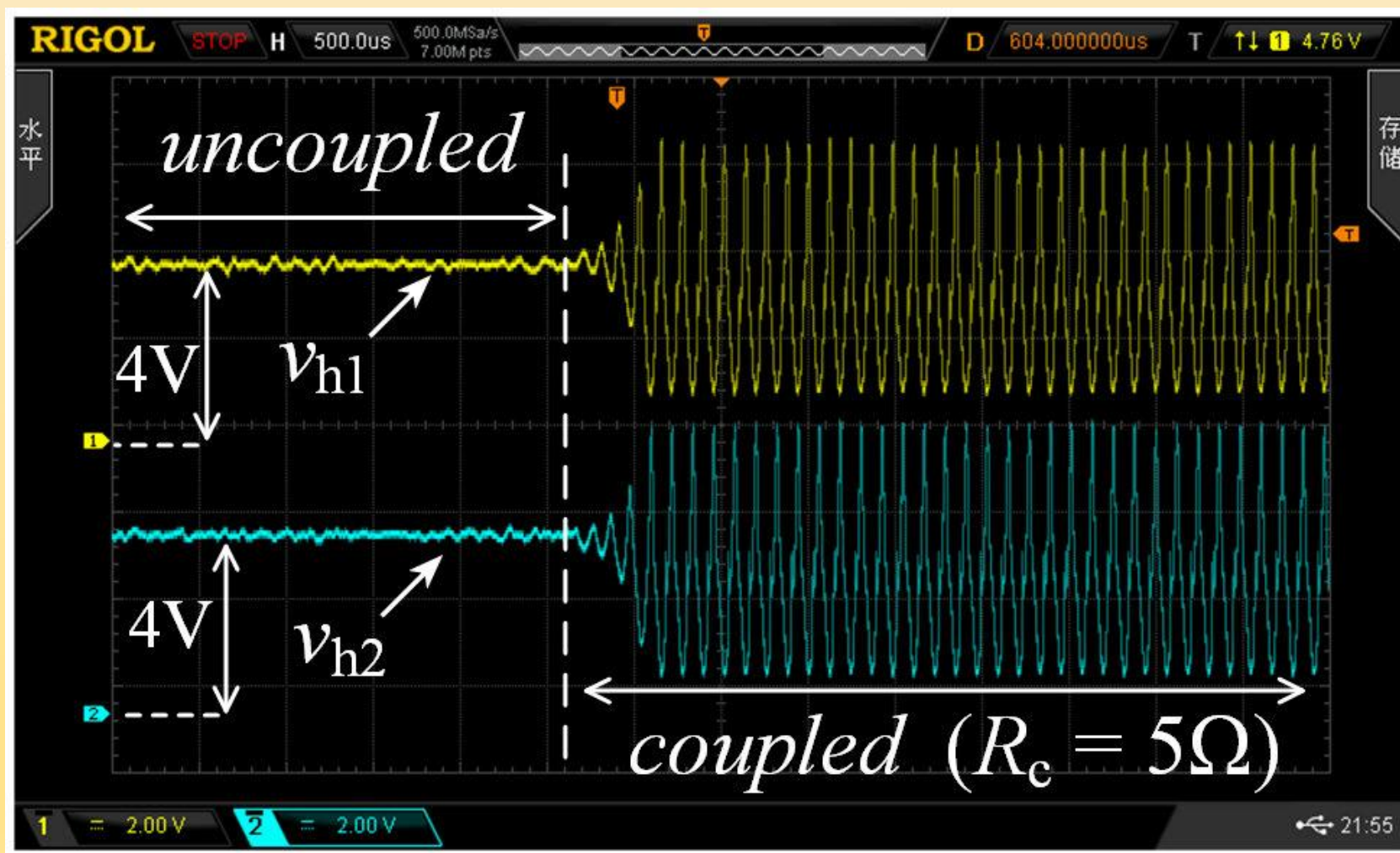
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A Poor Man's Memristor

Chua Corsage Memristor

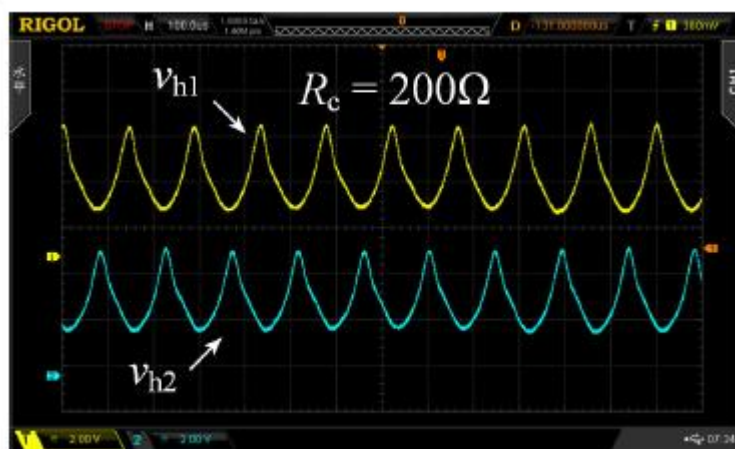
can be built
for less than

\$10 !

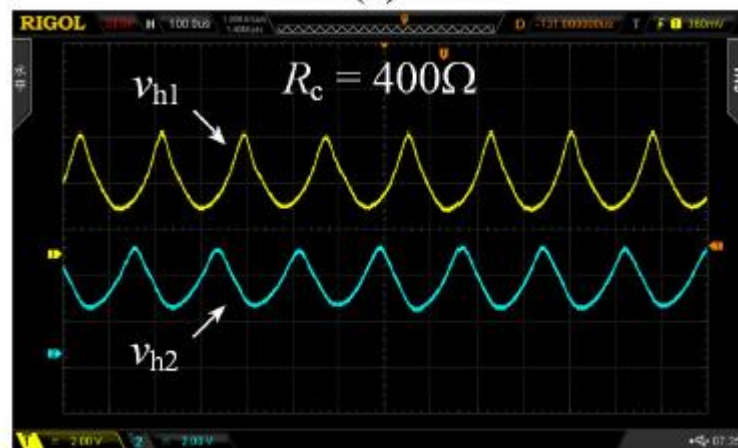




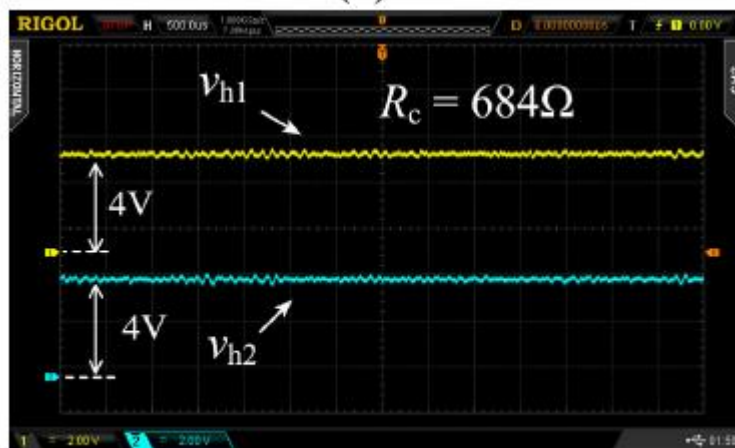
(a)



(b)



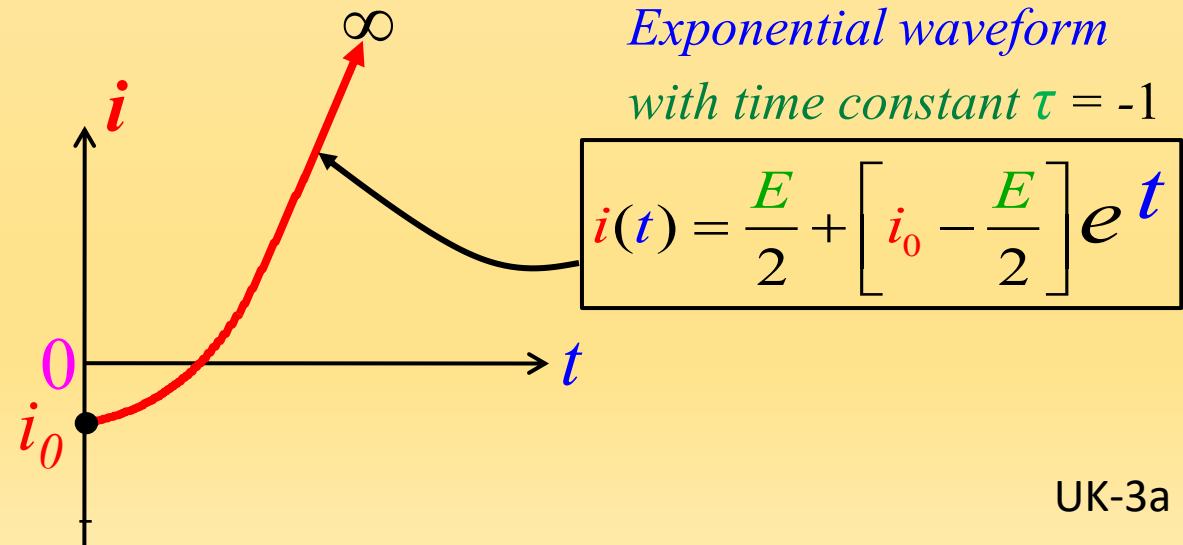
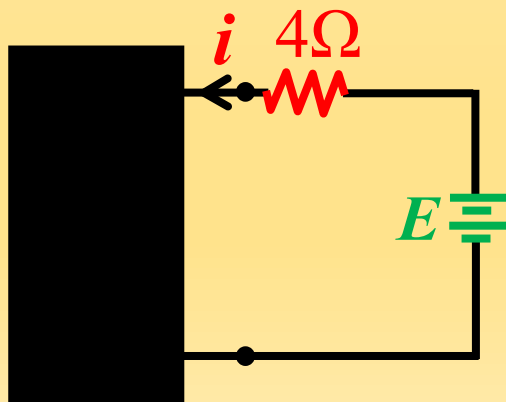
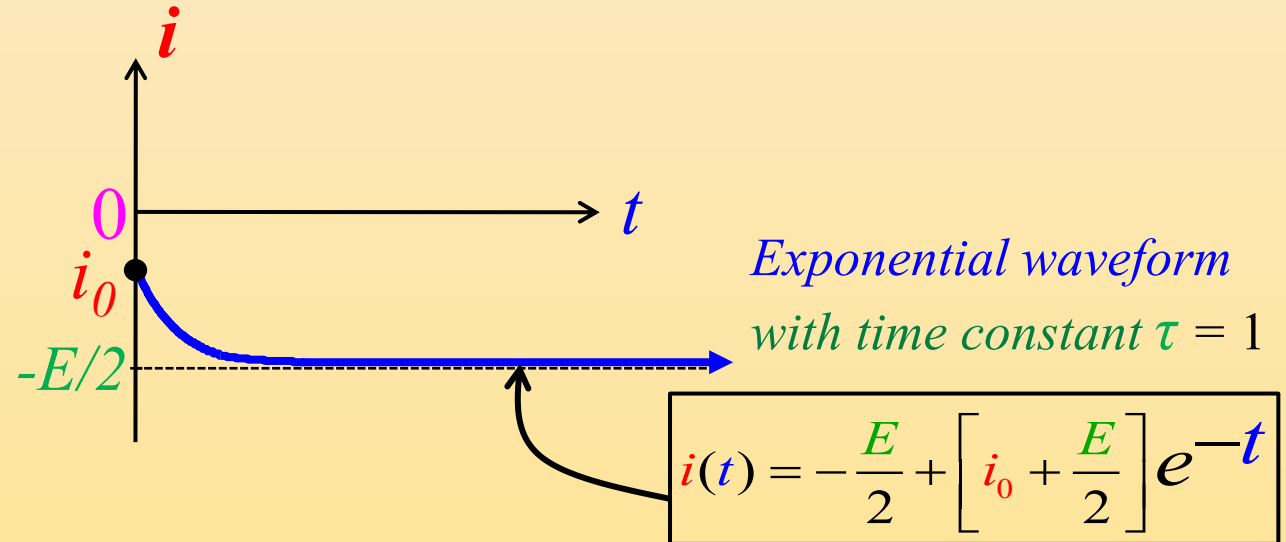
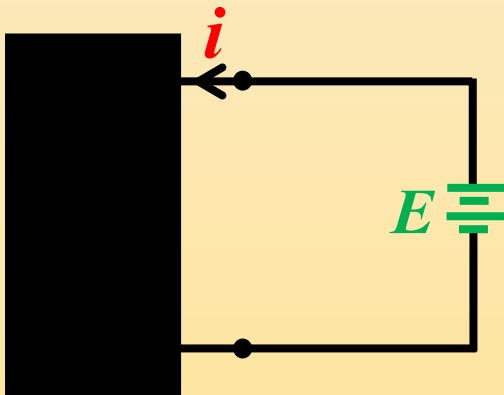
(c)



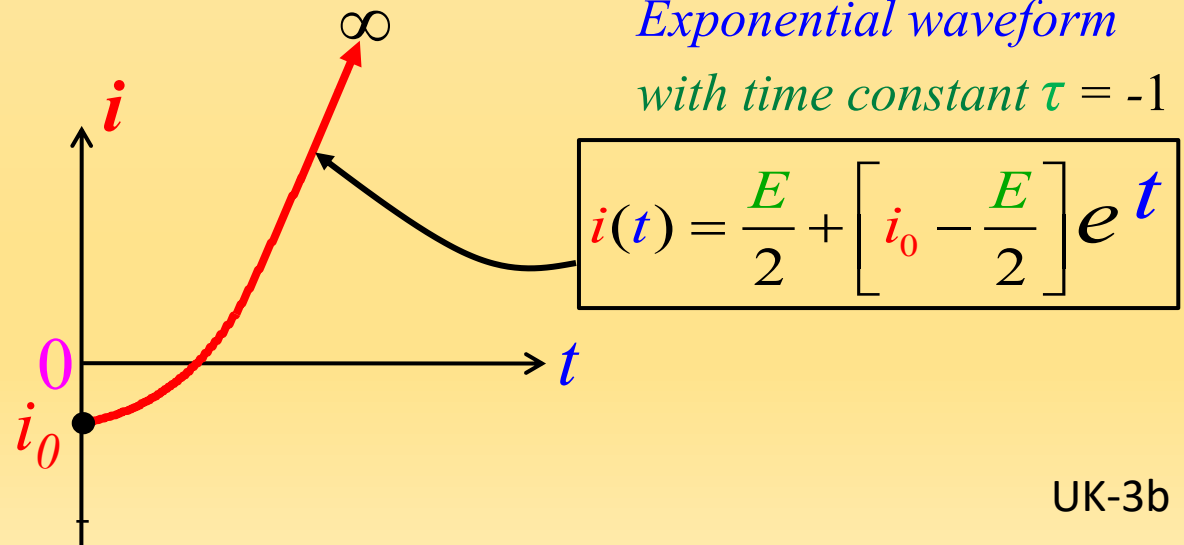
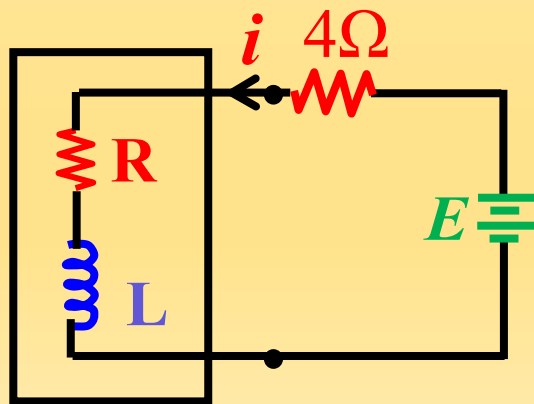
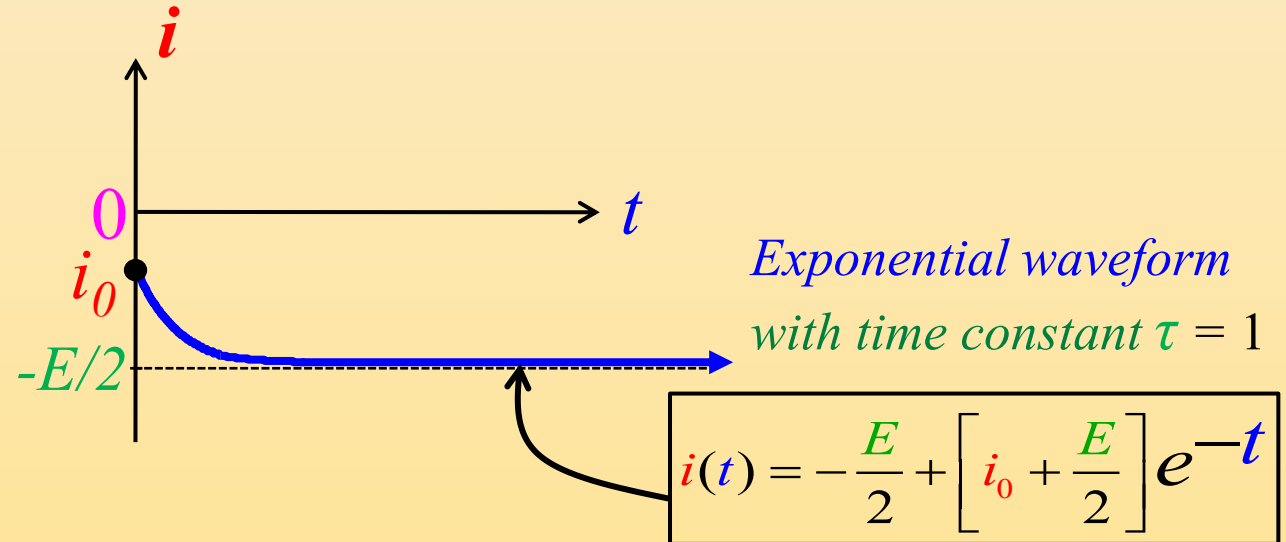
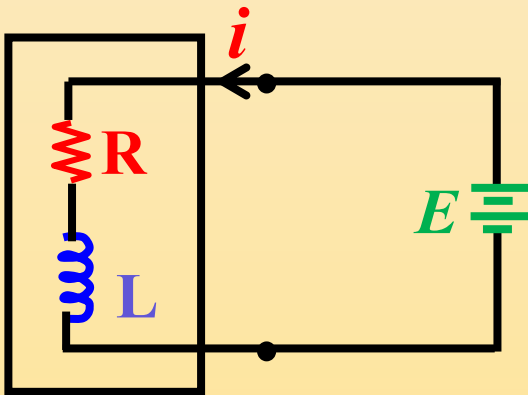
(d)

Chua's Riddle

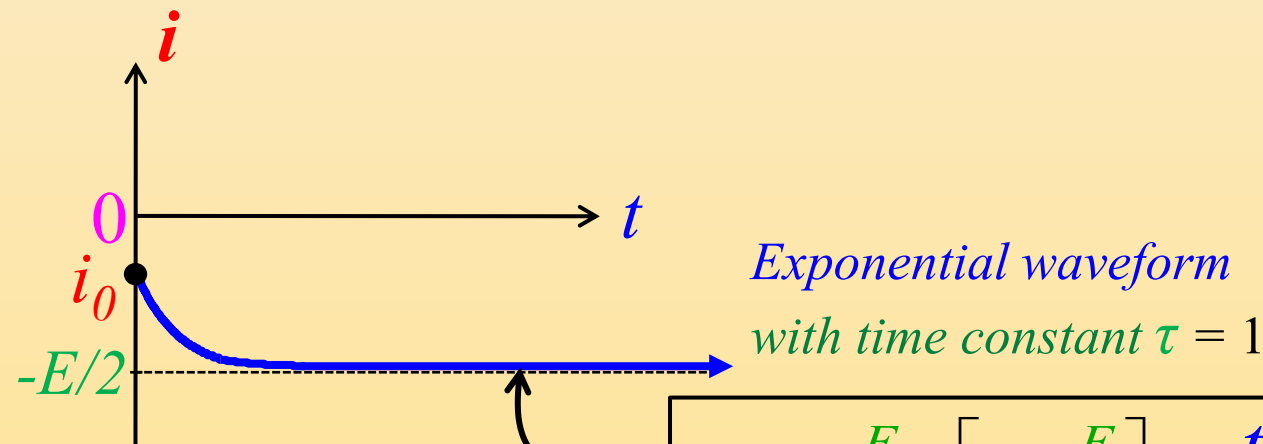
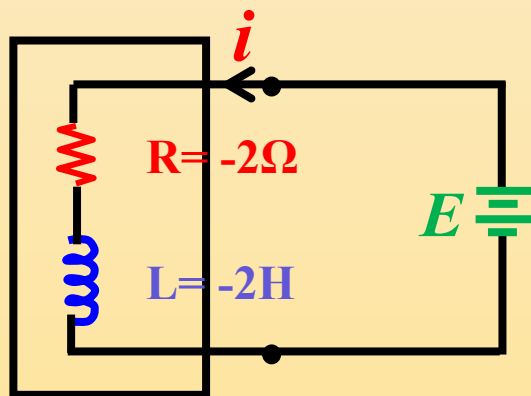
Chua's Riddle



Chua's Riddle



Chua's Riddle

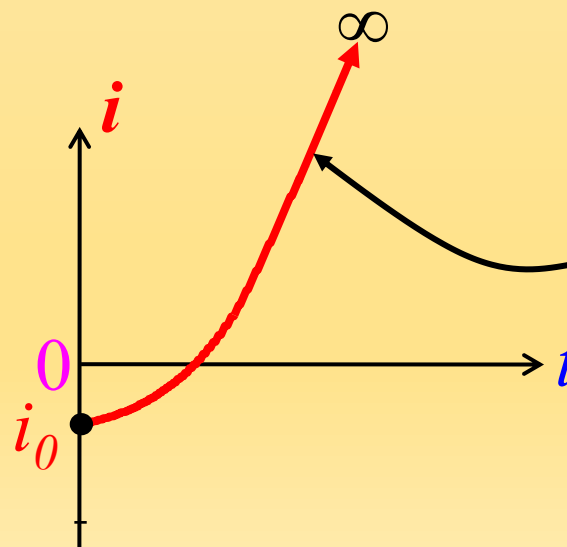
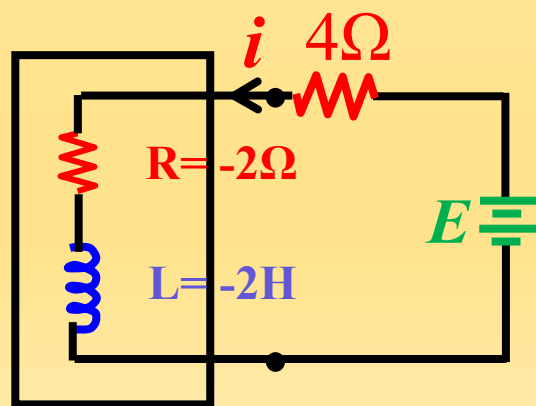


Exponential waveform
with time constant $\tau = 1$

$$i(t) = -\frac{E}{2} + \left[i_0 + \frac{E}{2} \right] e^{-t}$$

Exponential waveform
with time constant $\tau = -1$

$$i(t) = \frac{E}{2} + \left[i_0 - \frac{E}{2} \right] e^{t}$$



A New type of Computer



Geoffrey Hinton

“We are going to do what I call **mortal computation** where the knowledge that the system has learned and the hardware are inseparable. If we do that we can use very low power analog computation; you can have **trillion-way parallelism** *using things like memristors...*”

Great great grandson
of **George Boole**
Logician and **Inventor** of
Boolean Algebra

Geoffrey Hinton

Lecture presented at
NeurIPS, Dec. 2022

2024 Nobel Prize in Physics