

Lecture 3

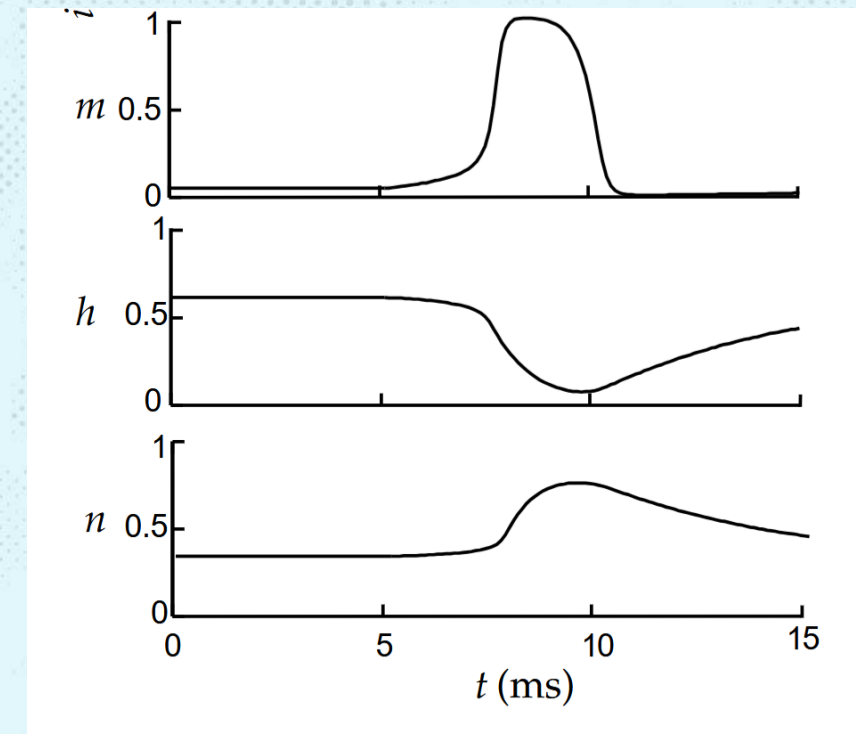
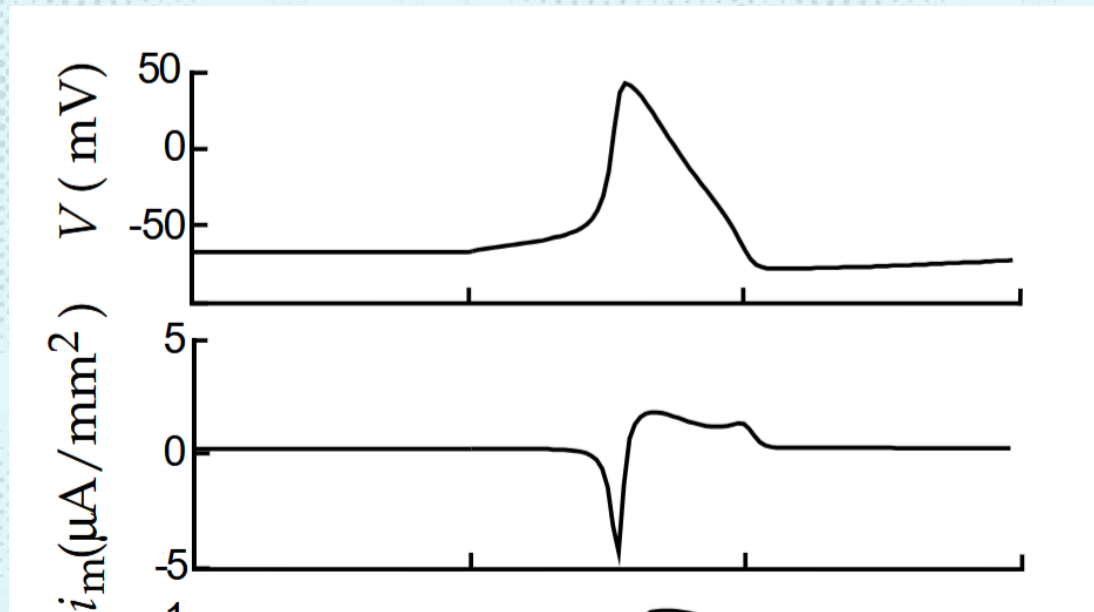
The Hodgkin-Huxley model

Dayan & Abbott, 2001

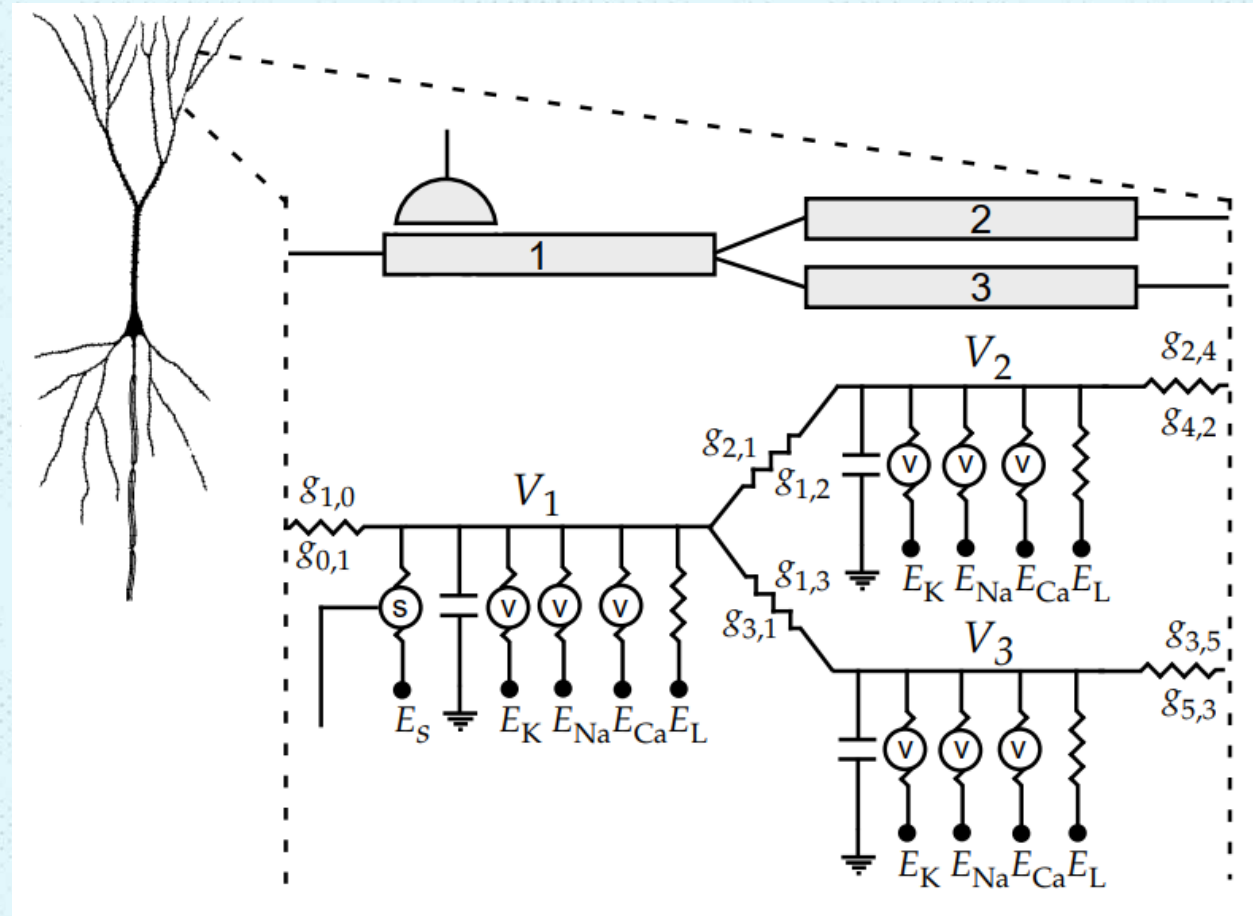
- Model of action potential generation

$$i_m = \bar{g}_L (V - E_L) + \bar{g}_K n^4 (V - E_K) + \bar{g}_{Na} m^3 h (V - E_{Na})$$

$$c_m \frac{dV}{dt} = -i_m + \frac{I_e}{A}.$$



Action potential propagation



Multicompartment models

- Non-branching compartment:

Connections to
neighbouring
compartments

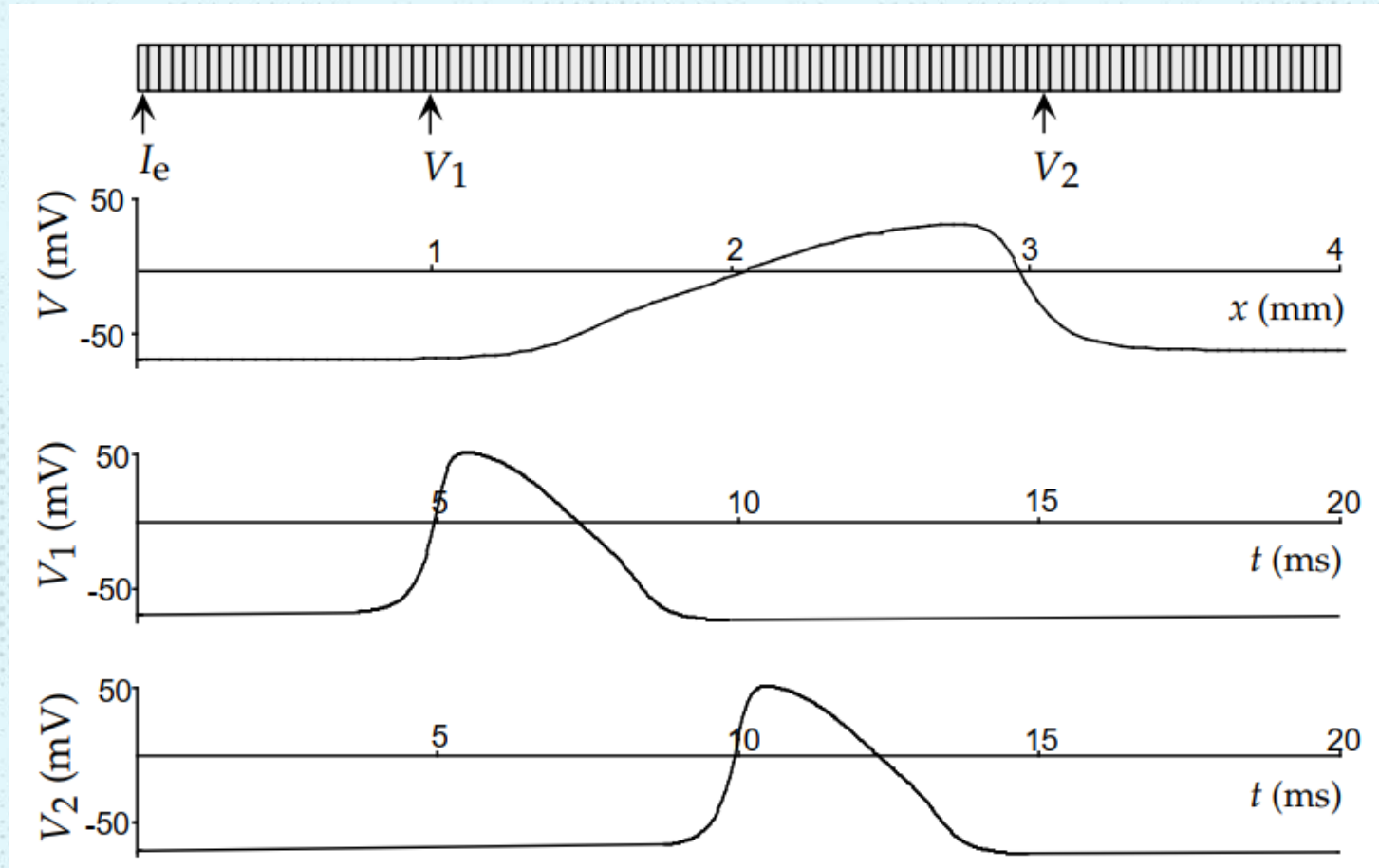
$$c_m \frac{dV_\mu}{dt} = -i_m^\mu + \frac{I_e^\mu}{A_\mu} + g_{\mu,\mu+1}(V_{\mu+1} - V_\mu) + g_{\mu,\mu-1}(V_{\mu-1} - V_\mu)$$

- Coupling constant:

$$g_{\mu,\mu'} = \frac{a_\mu a_{\mu'}^2}{r_L L_\mu (L_\mu a_{\mu'}^2 + L_{\mu'} a_\mu^2)}$$

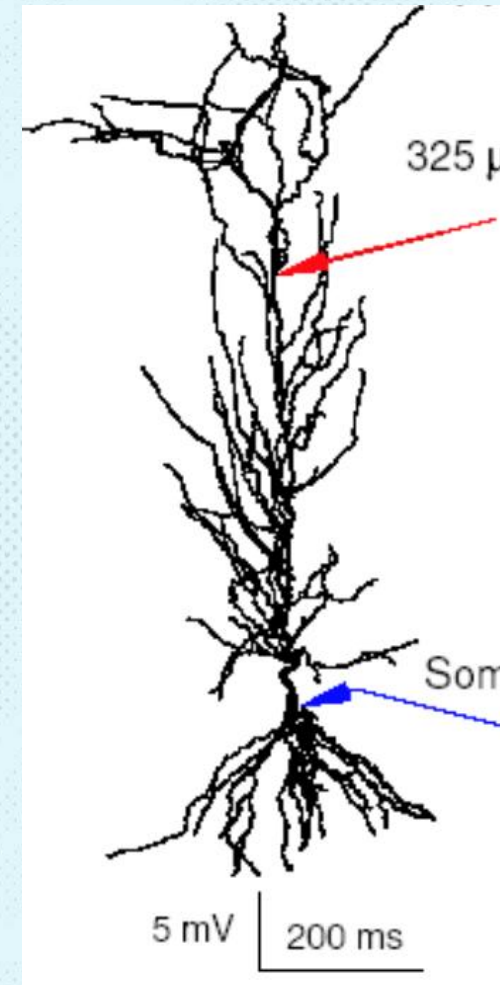
- Solving: Numerical integration!

Action potential propagation



Example 1: CA1 pyramidal cell

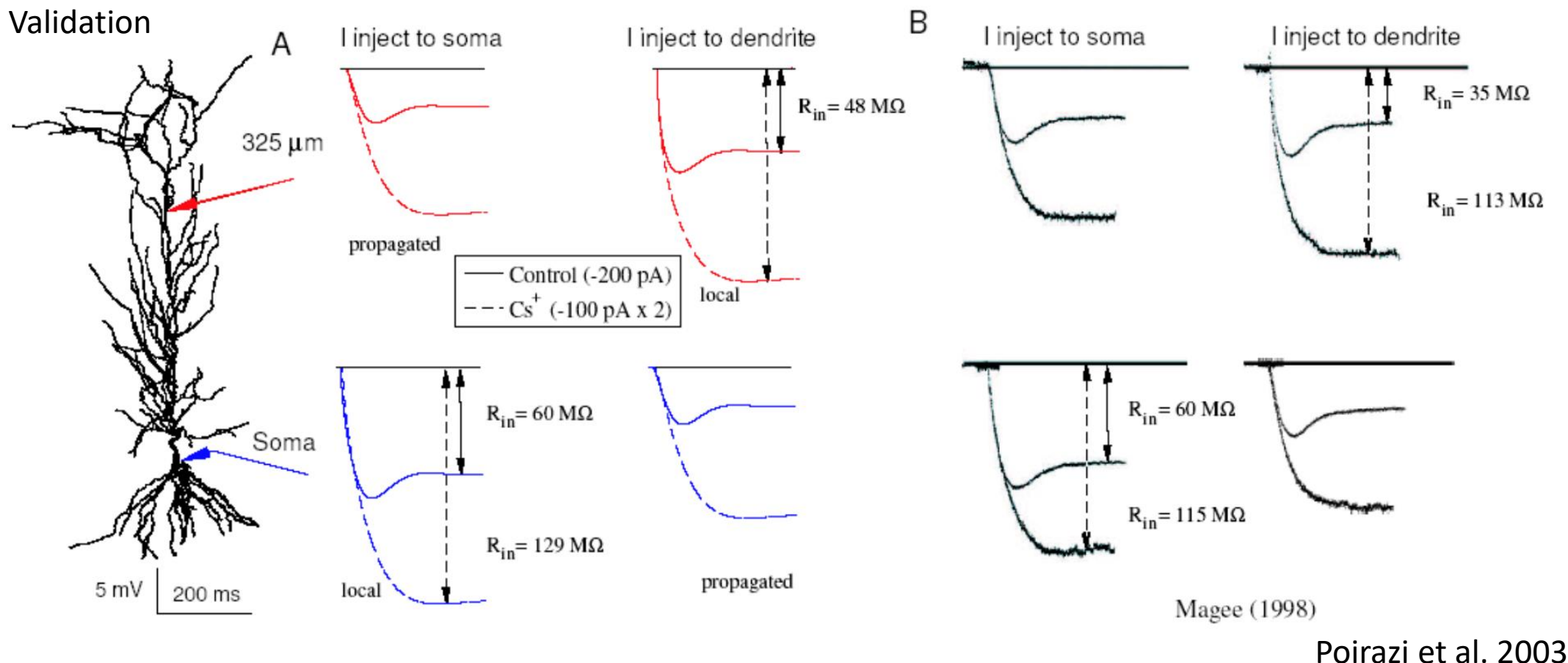
- Goal: understand dendritic integration in pyramidal cells
- Experiments are very difficult, replace real cells with model
- 21 types of ion channels + synapses
- Data source: primary literature on ion channels, synapses, channel densities
- Models inference: Hand-tuning!



Poirazi et al. 2003

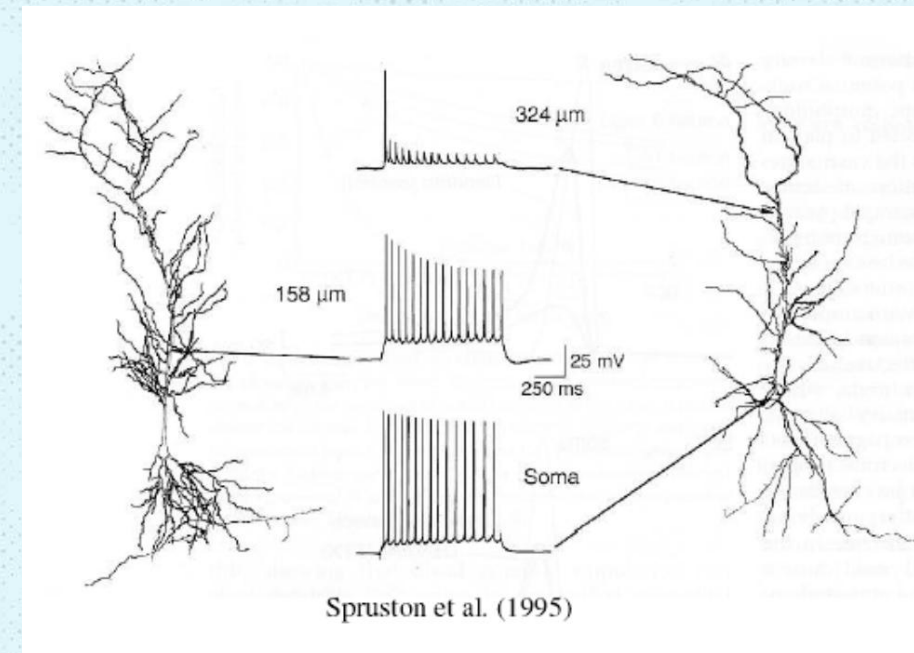
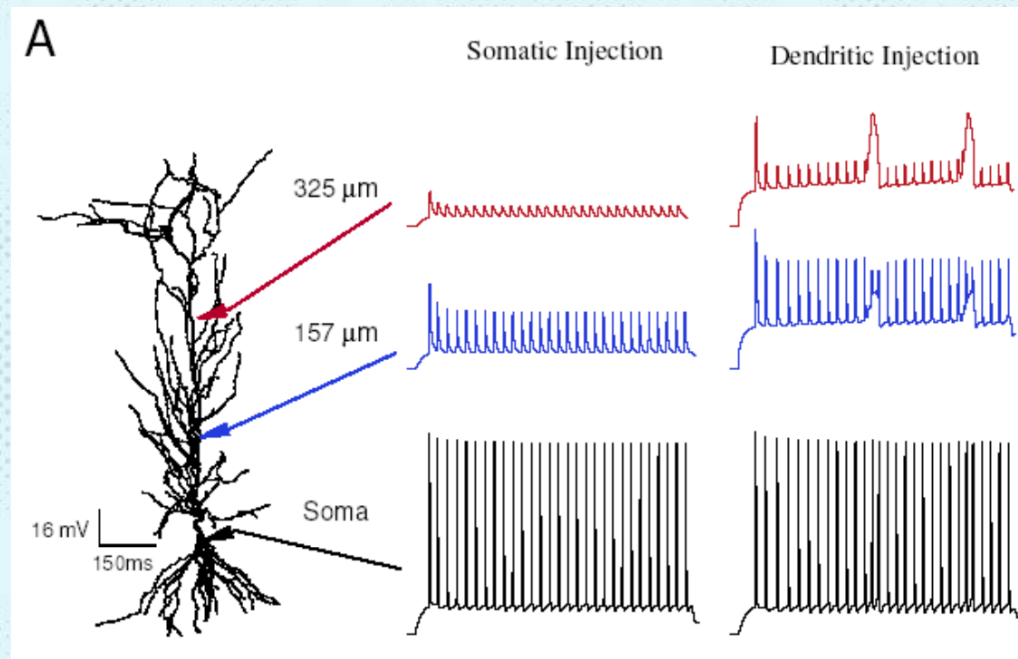
Example 1: CA1 pyramidal cell

Validation



Example 1: CA1 pyramidal cell

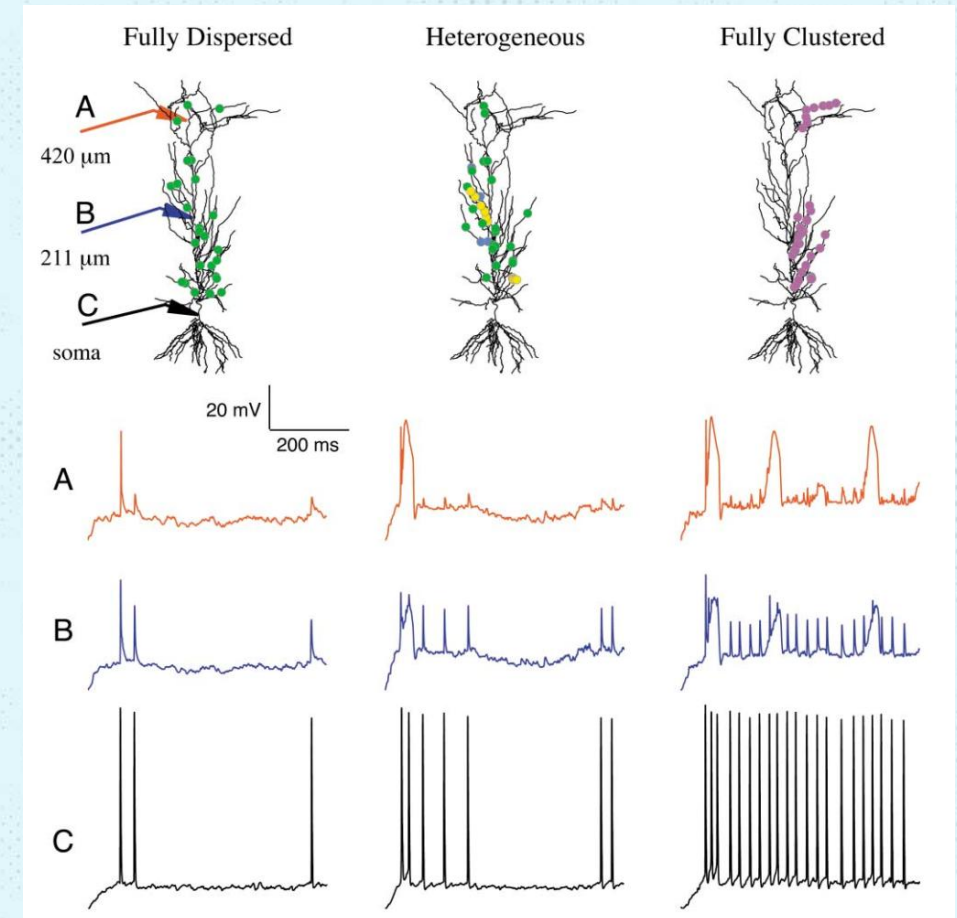
Validation



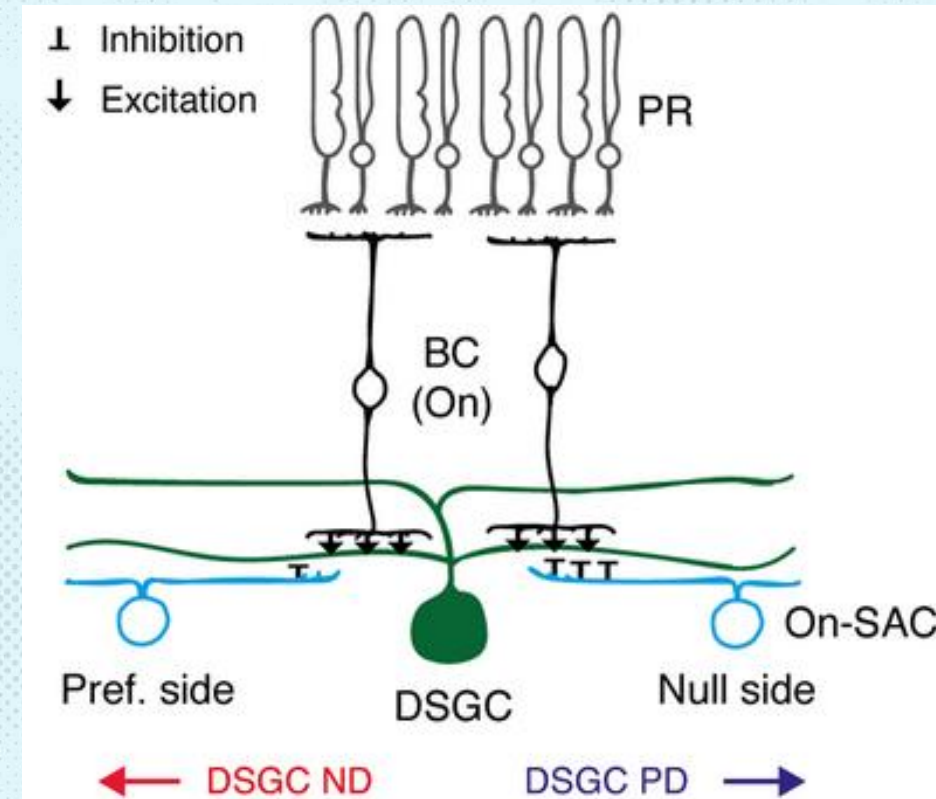
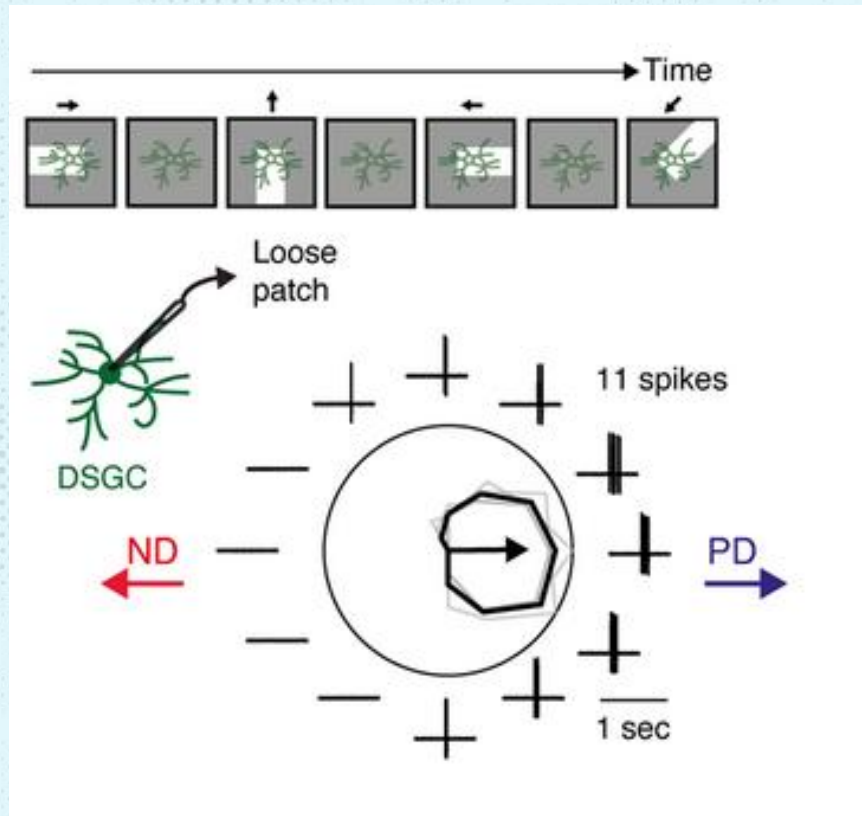
Poirazi et al. 2003

Example 1: CA1 pyramidal cell

- Placement of inputs affects propagation through the cell even with identical total input
- Dispersed: Minimal effect
- Clustered: Activates dendritic calcium spikes and leads to effective propagation



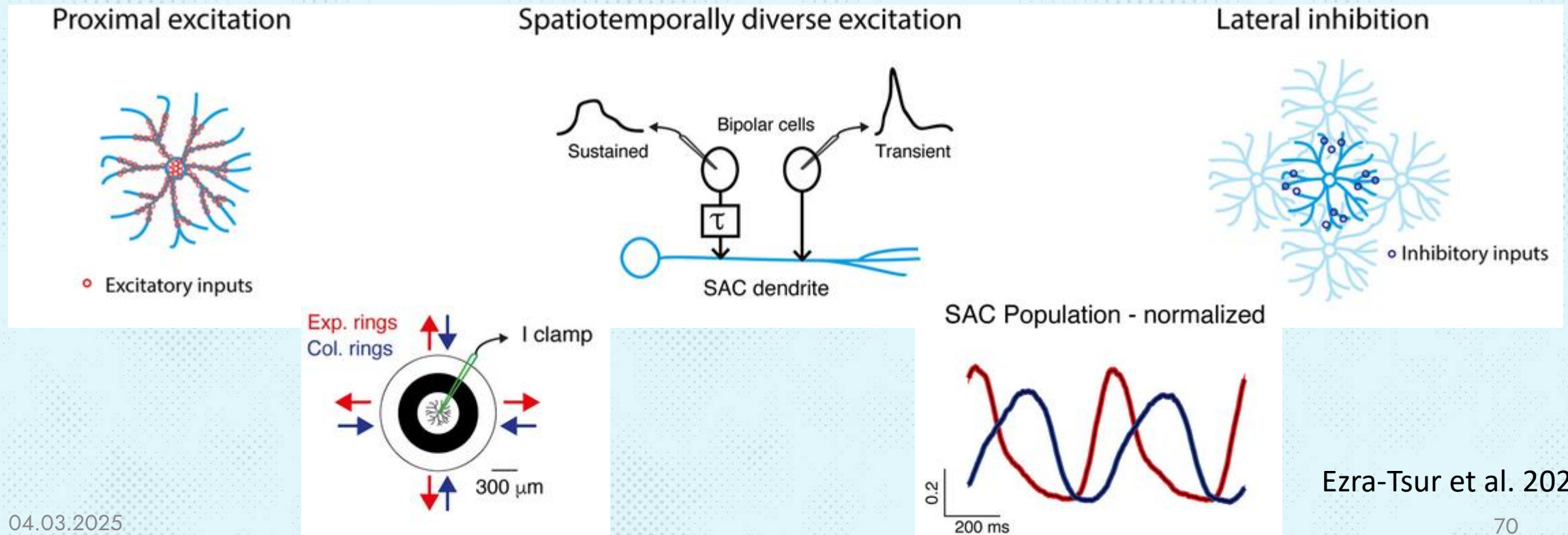
Example 2: Direction selectivity in the retina



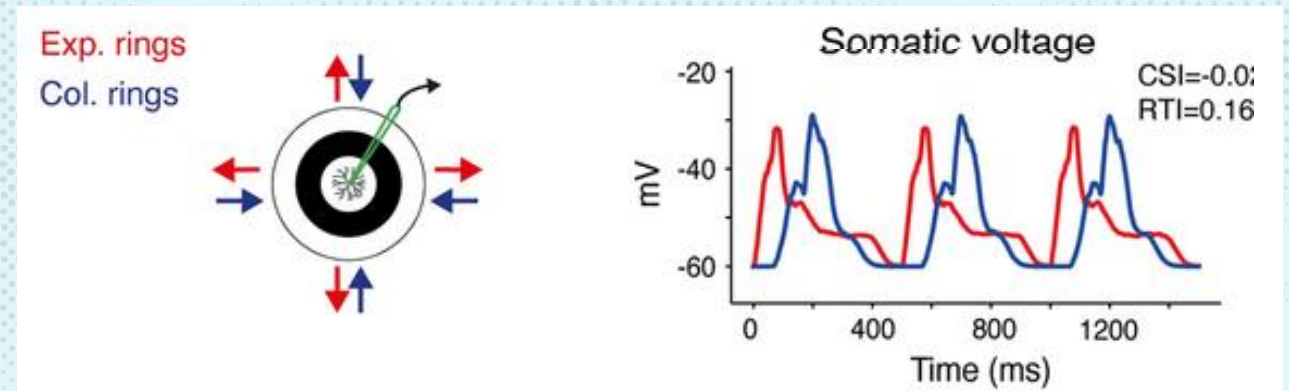
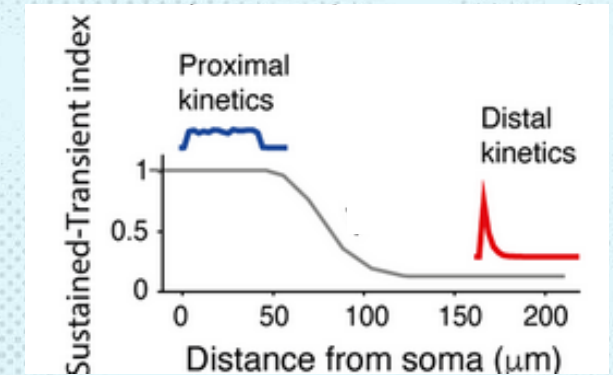
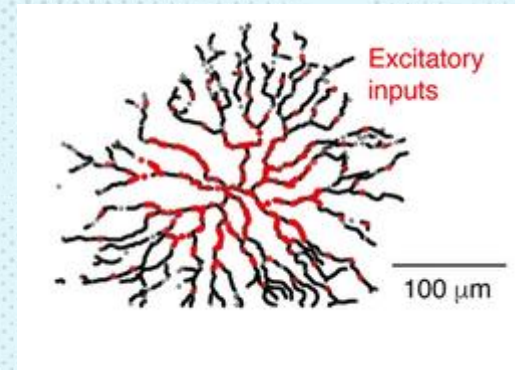
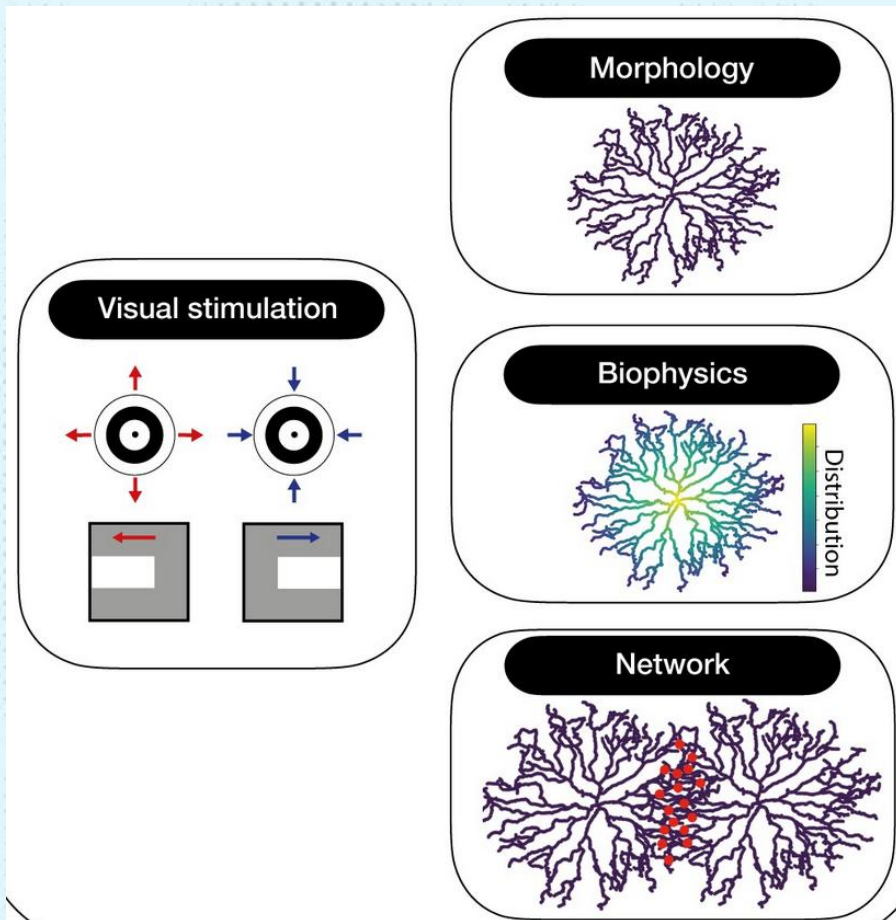
Ezra-Tsur et al. 2021

Example 2: Direction selectivity in the retina

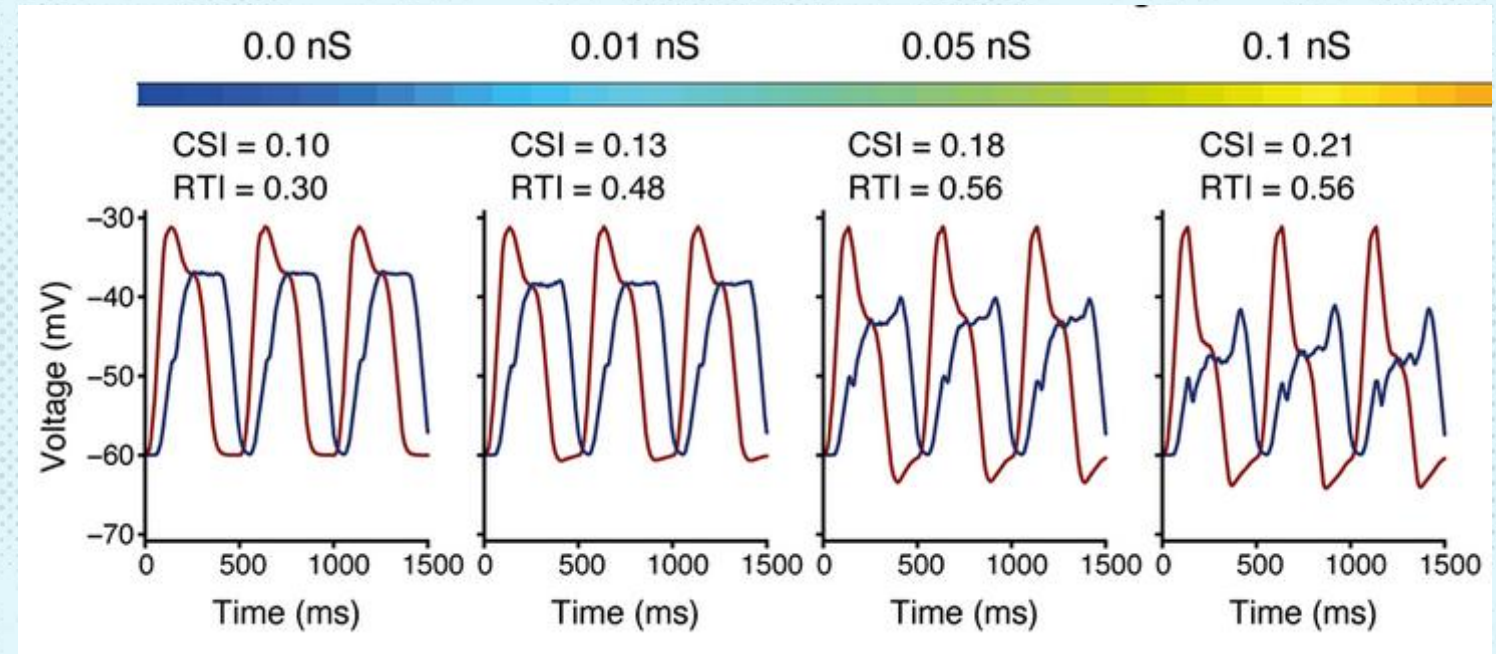
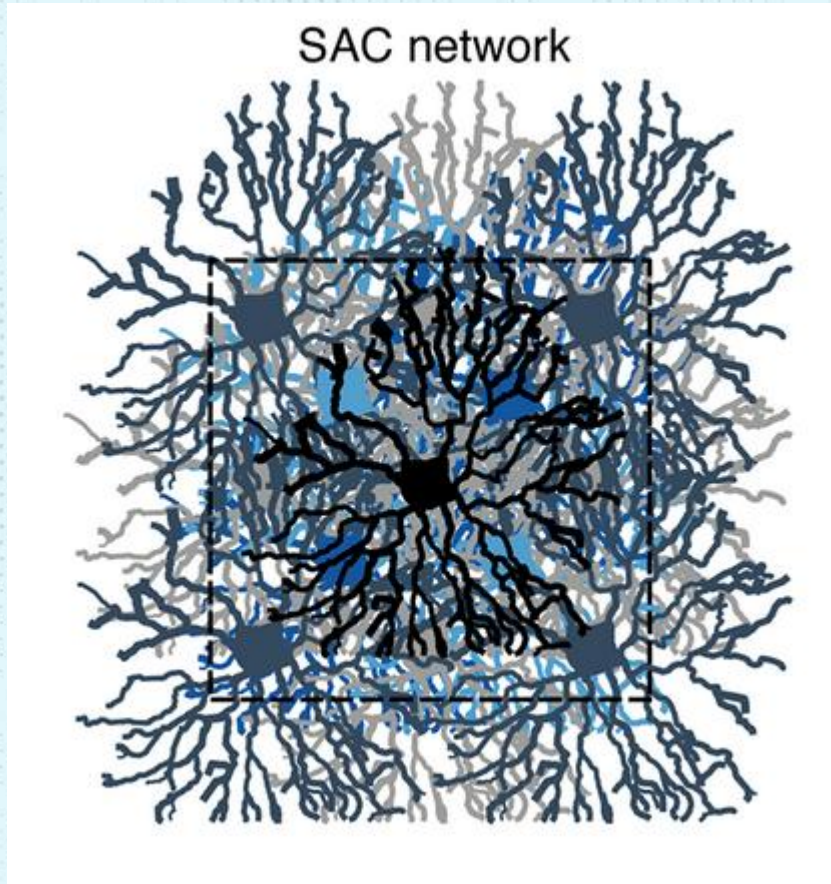
Goal: uncover mechanism underlying direction selectivity & role of starburst amacrine cells



Example 2: Direction selectivity in the retina



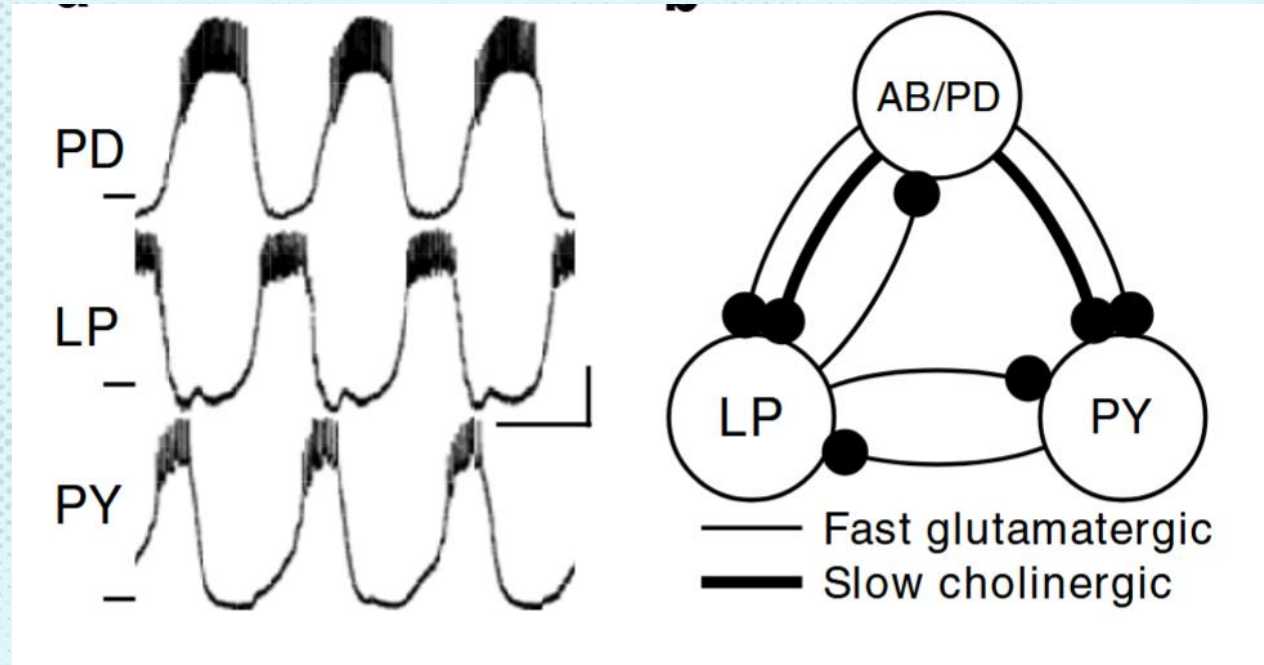
Example 2: Direction selectivity in the retina



Ezra-Tsur et al. 2021

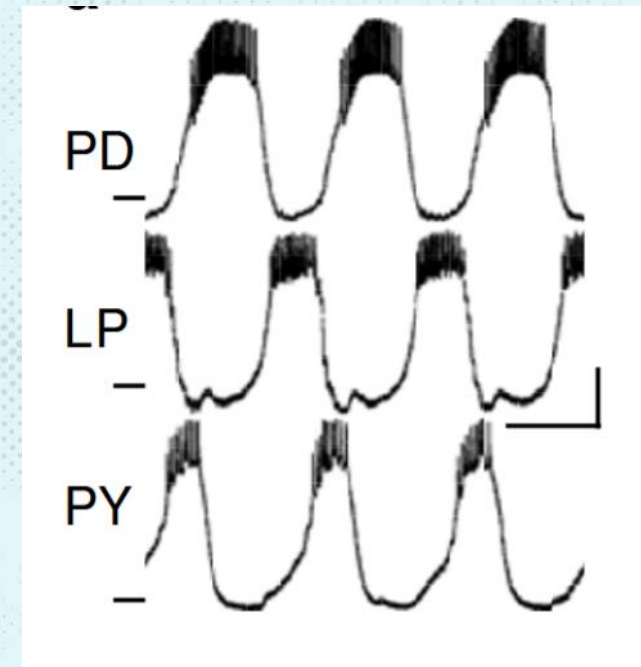
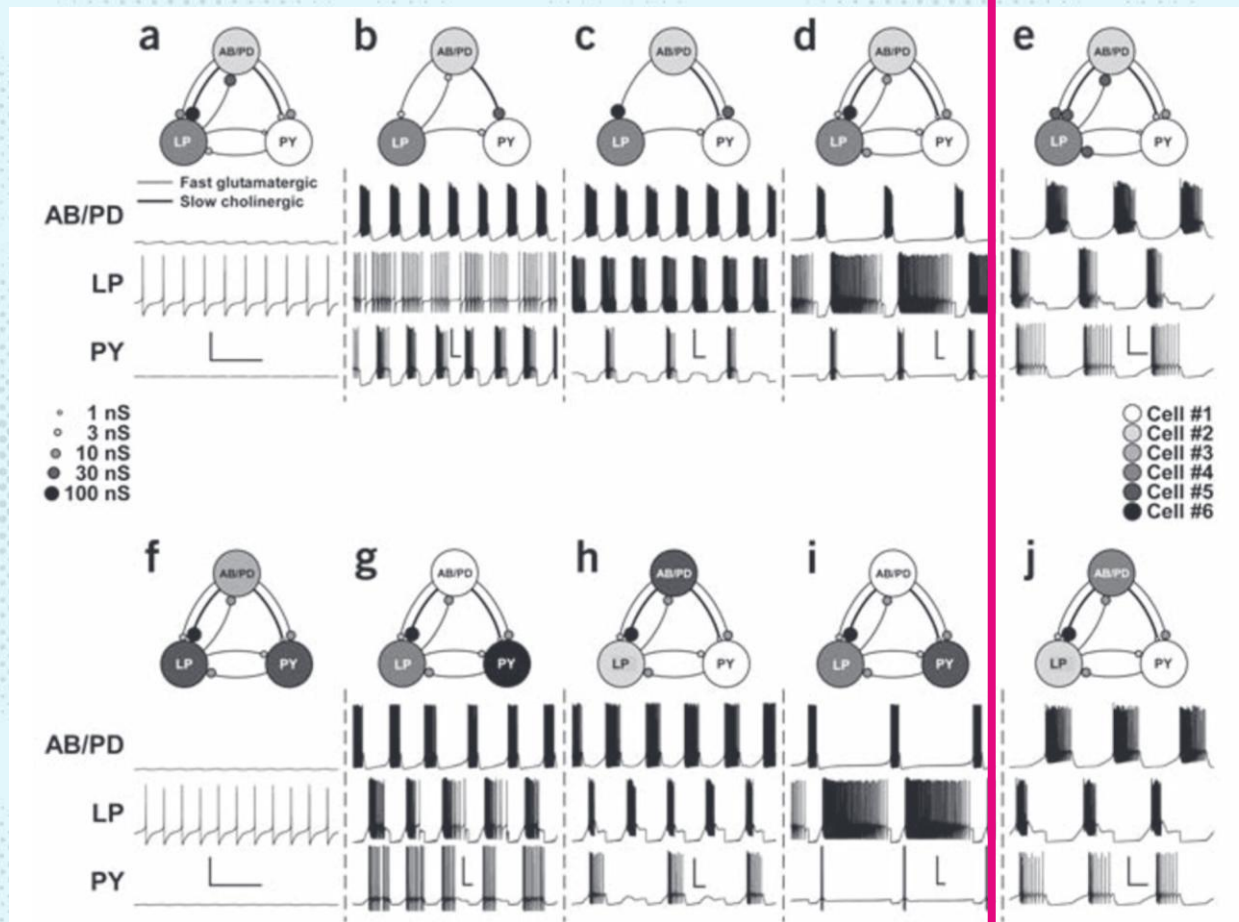
Example 3: STG circuit

- Stomatogastric ganglion of crabs
- 3 neurons: LP, PY and PD
- Rhythmic network
- 7 synapses



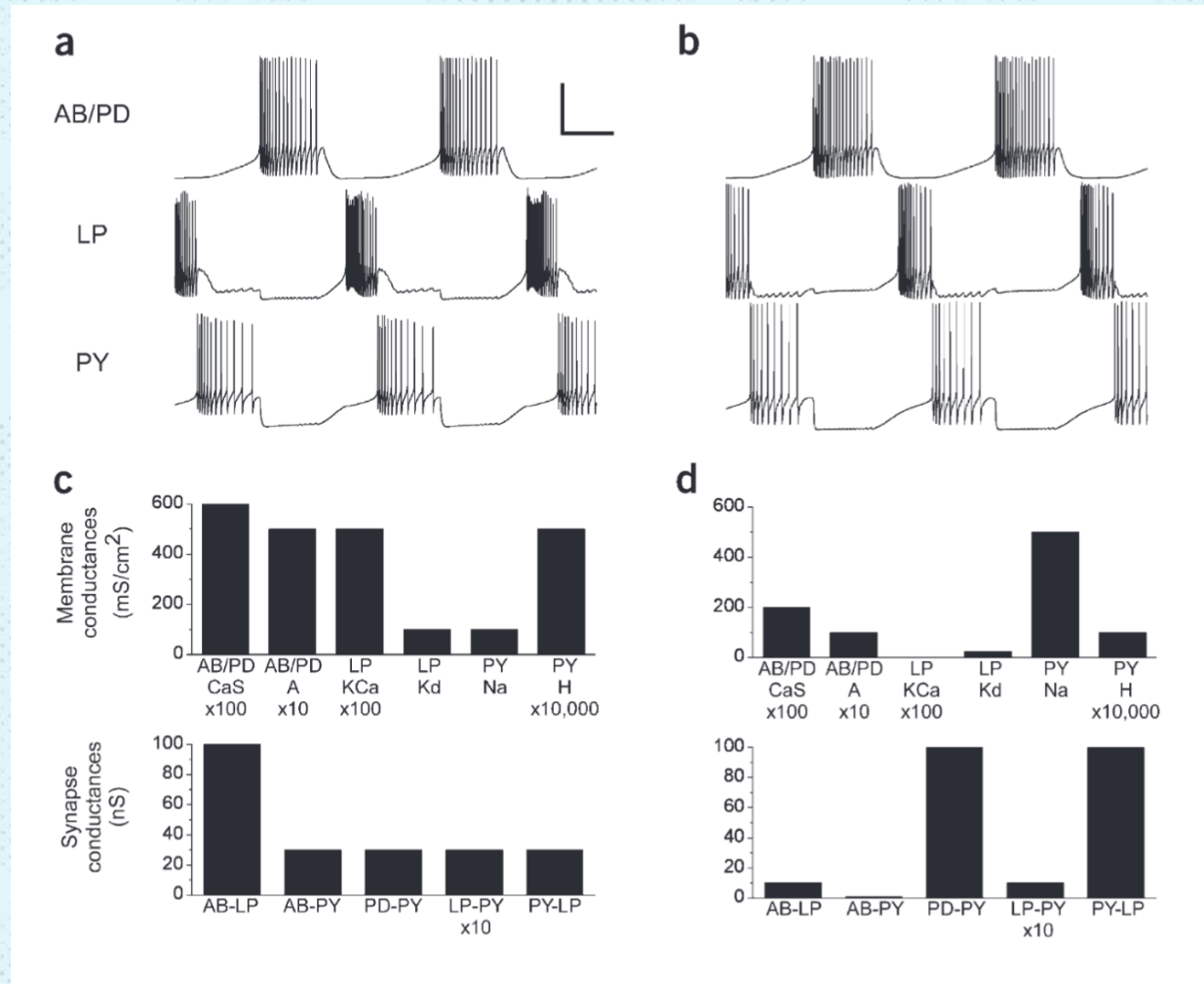
Prinz et al. 2004

Example 3: STG circuit



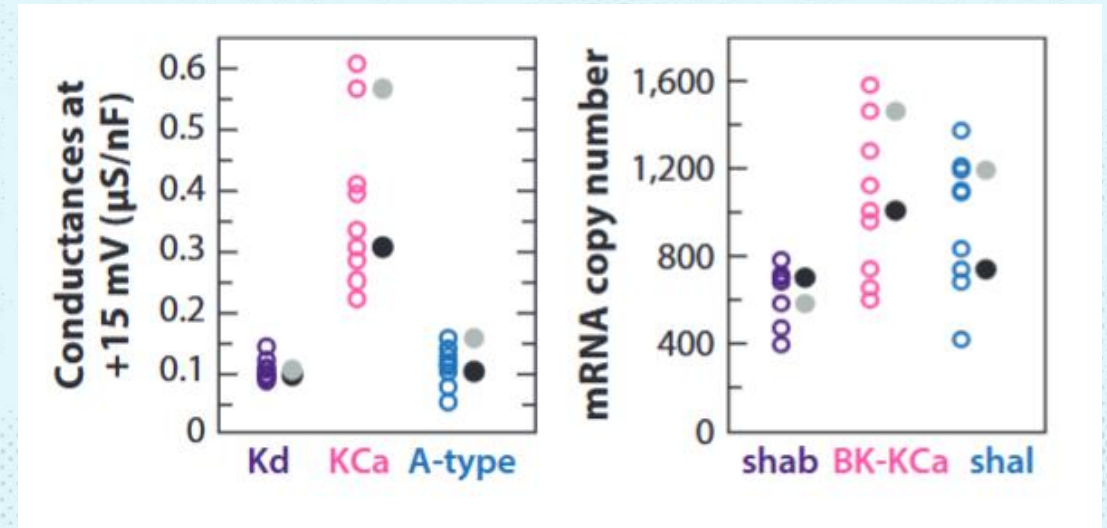
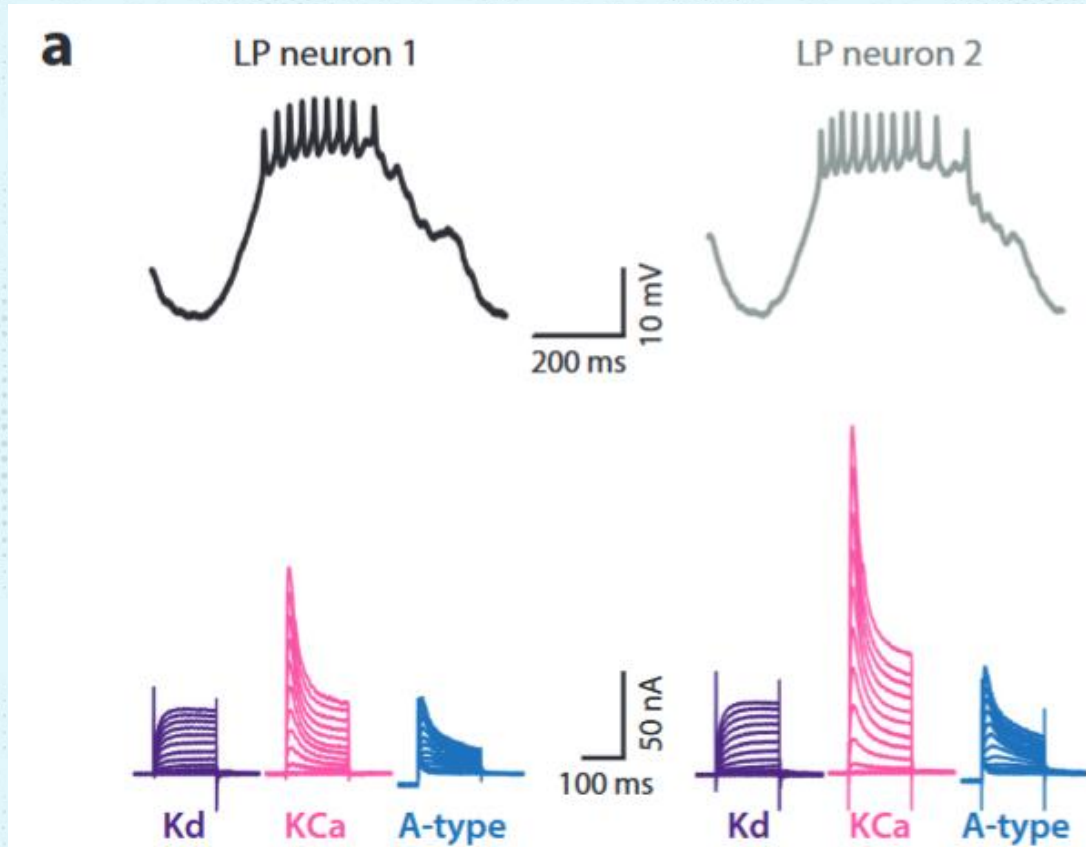
Prinz et al. 2004

Example 3: STG circuit



Prinz et al. 2004

Example 3: STG circuit



Goaillard & Marder 2021

Simulating cellular models with Forward Euler

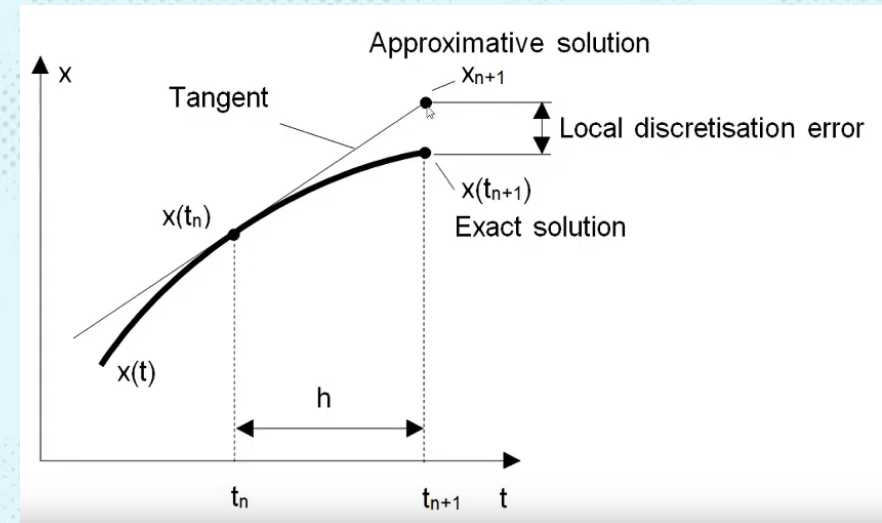
- Define initial time t_0 and time step Δt
→ solution grid: $t_n = n \cdot \Delta t$
- Compute discrete approximation to left hand side

$$c_m \frac{dV(t_n)}{dt} = c_m \frac{V(t_{n+1}) - V(t_n)}{\Delta t} = -i_m(t_n) + \frac{I_e(t_n)}{A}$$

$$V(t_{n+1}) = V(t_n) - \frac{i_m(t_n)\Delta t}{c_m} + \frac{I_e(t_n)\Delta t}{Ac_m}$$

- Approximation error is large for multicompartment models

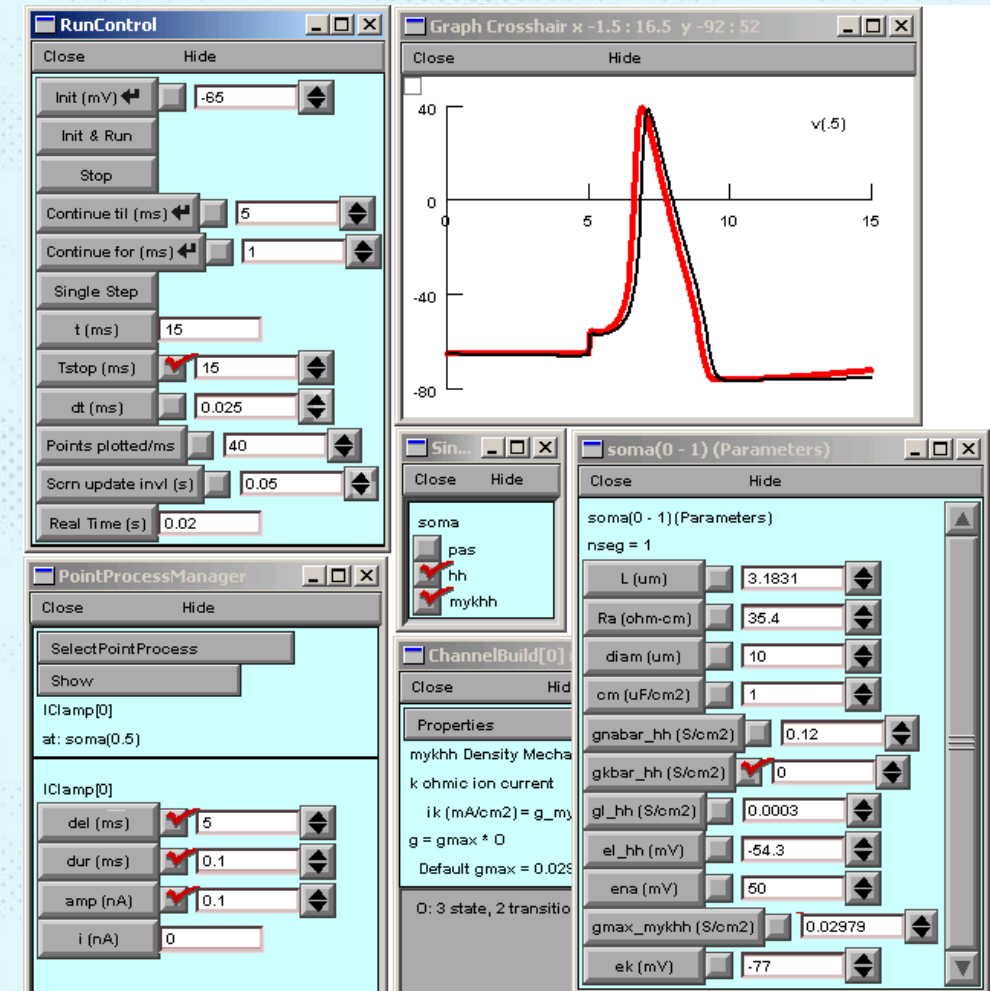
$$c_m \frac{dV}{dt} = -i_m + \frac{I_e}{A}.$$



<https://www.youtube.com/watch?v=rjeSqqHowTg>

Software solutions

- Specialized software solutions exist for simulation of multicompartment models
- Classic simulator: NEURON
- >2000 papers



Issues with classic simulators

Int J Biomed Comput, 24 (1989) 55—68
Elsevier Scientific Publishers Ireland Ltd.

55

A PROGRAM FOR SIMULATION OF NERVE EQUATIONS WITH BRANCHING GEOMETRIES

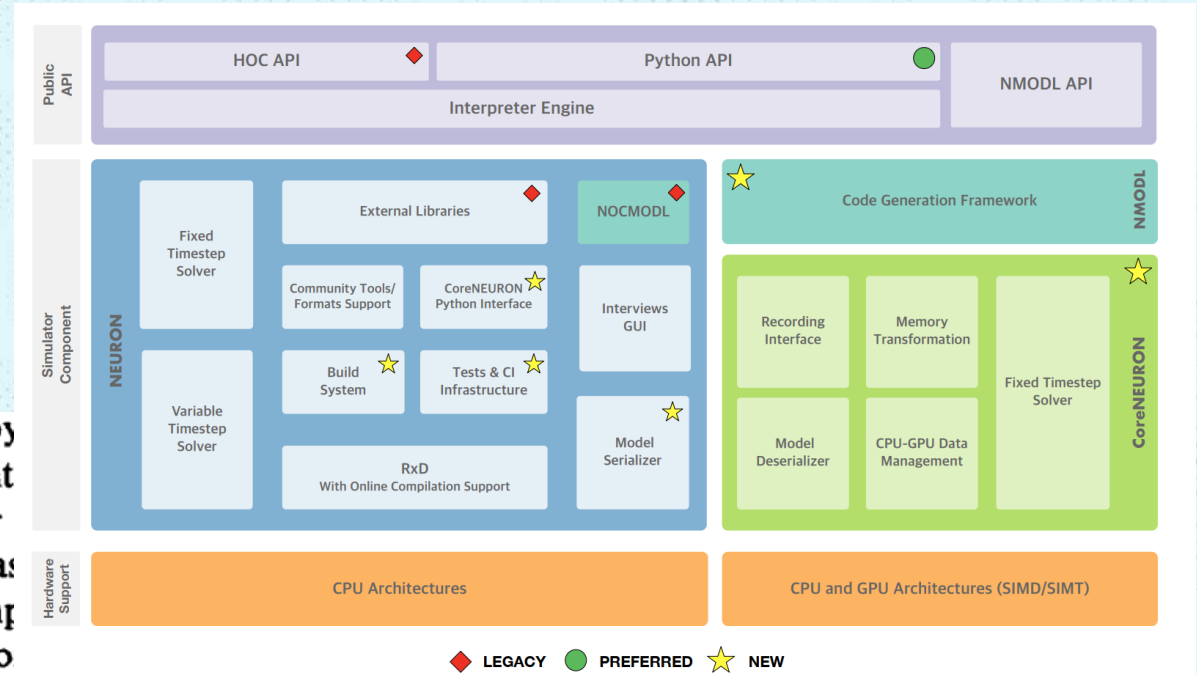
MICHAEL HINES

Department of Neurobiology, Duke University Medical Center, Durham, NC 27710 (U.S.A.)

(Received September 7th, 1988)

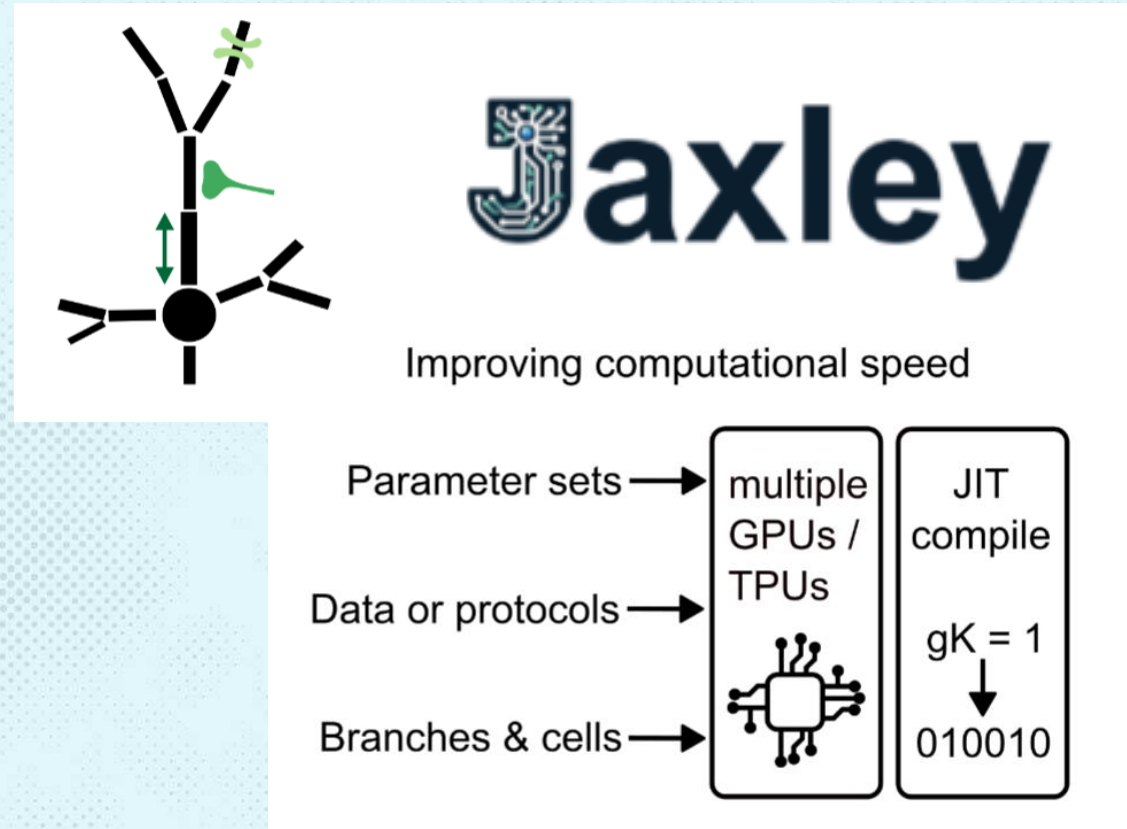
(Accepted December 6th, 1988)

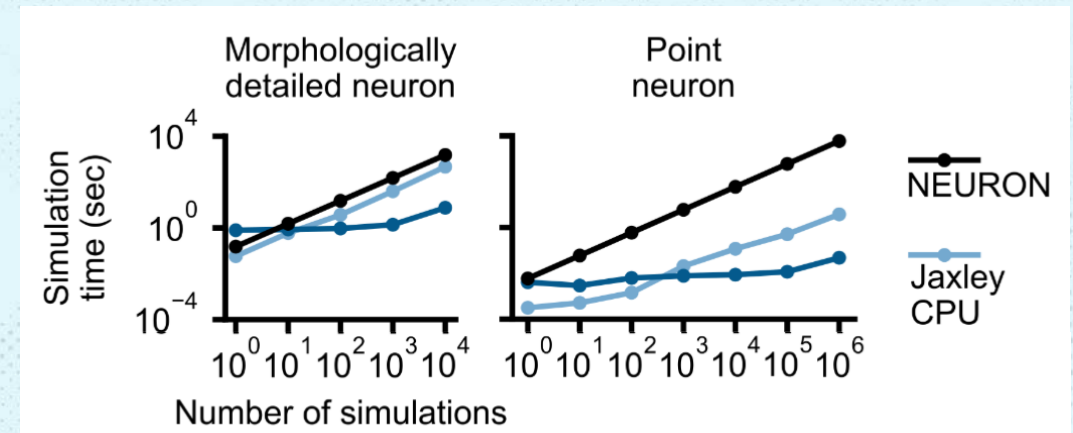
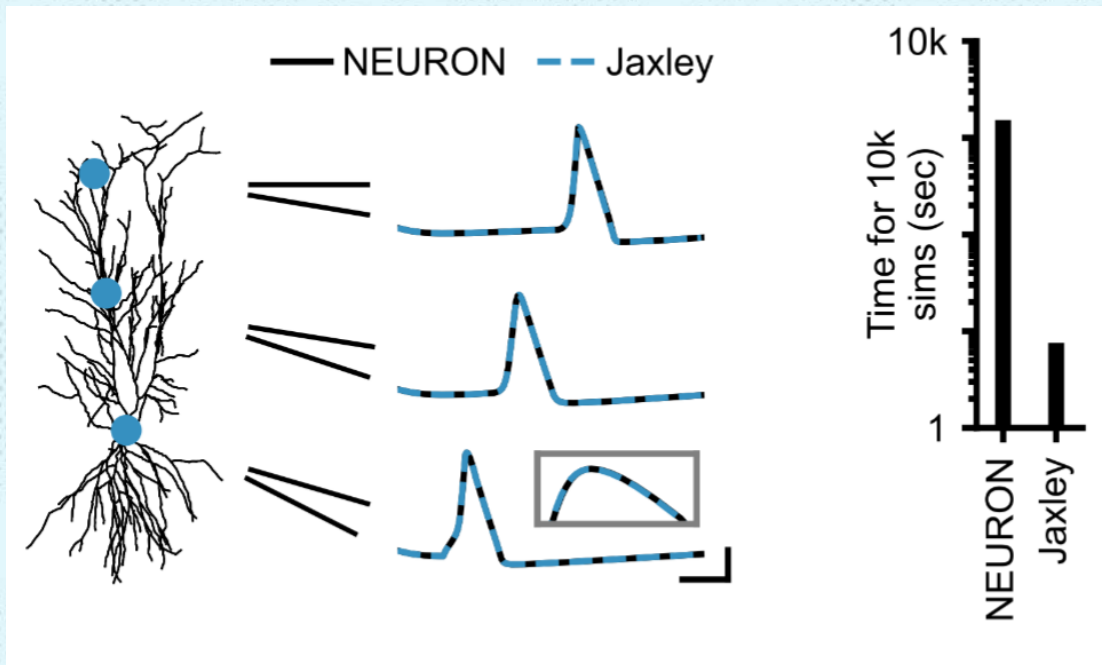
In the mid 1970s we began to carry out our simulations by equation subroutines written in *FORTTRAN* with a *FOCAL* interpreter (for Corporation minicomputers). We were astonished at the increase in this simulation methodology over the time consuming edit, compile — much less effort was required to create a casual or exploratory program to control the simulation. Unfortunately, since the interpreter portion of the cable program was written in Assembler (and therefore not portable to other machines) and since the program was limited to non-branching cables with Hodgkin-Huxley membrane kinetics, the program was unsuitable for general use outside our lab.



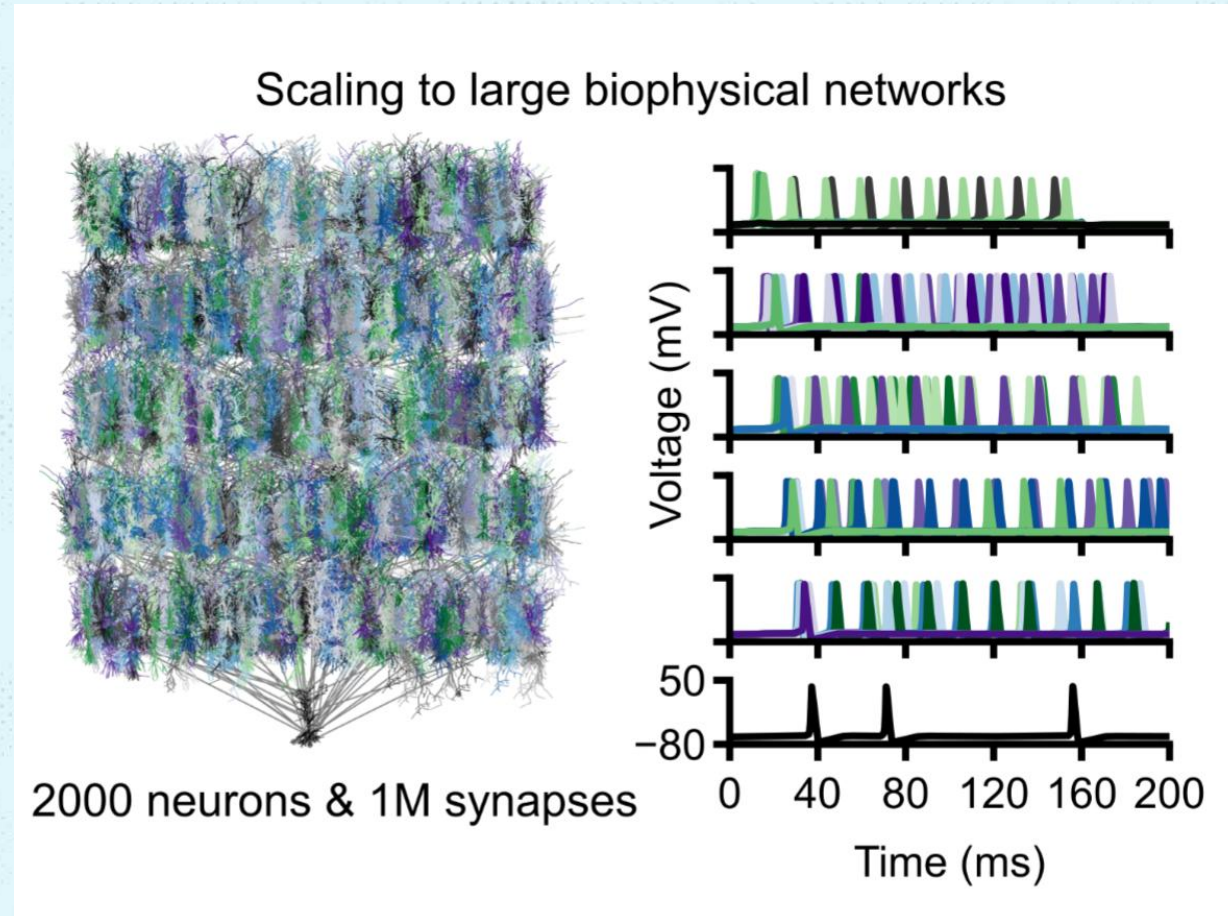
Jaxley – a modern simulator for biophysically detailed neurons

- Based on jax, a deep learning framework
- Native GPU support
- Speeds up simulation through native parallelization

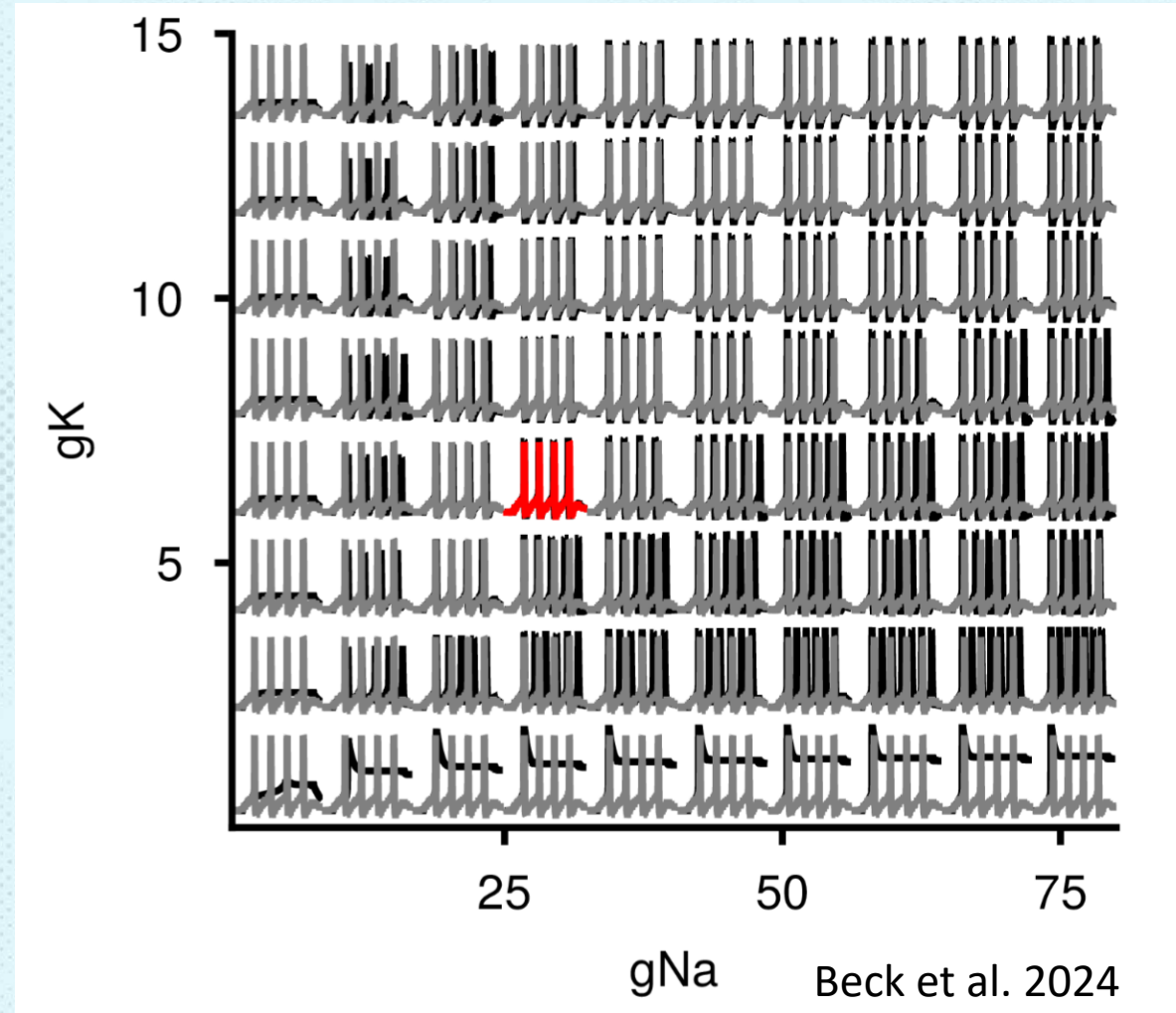
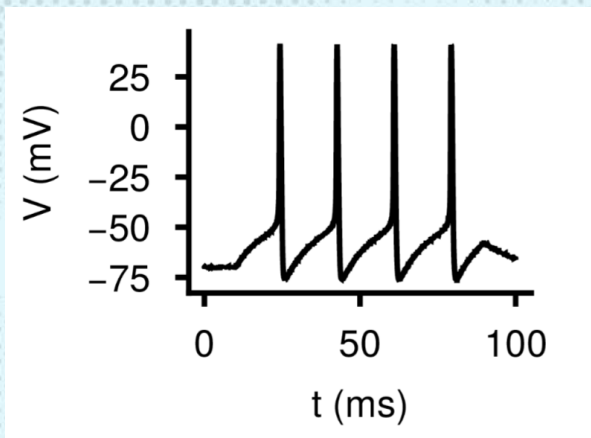
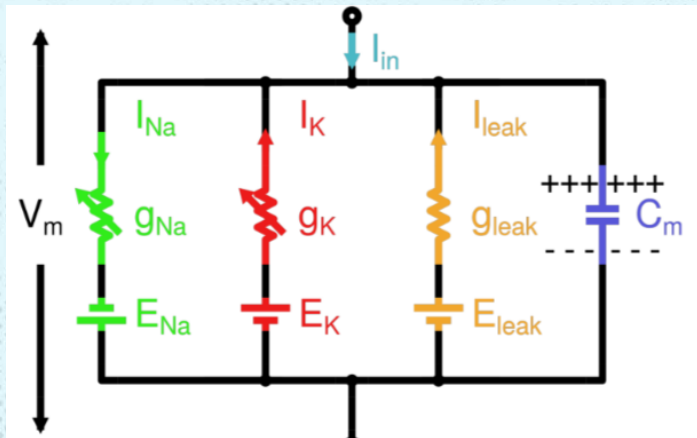




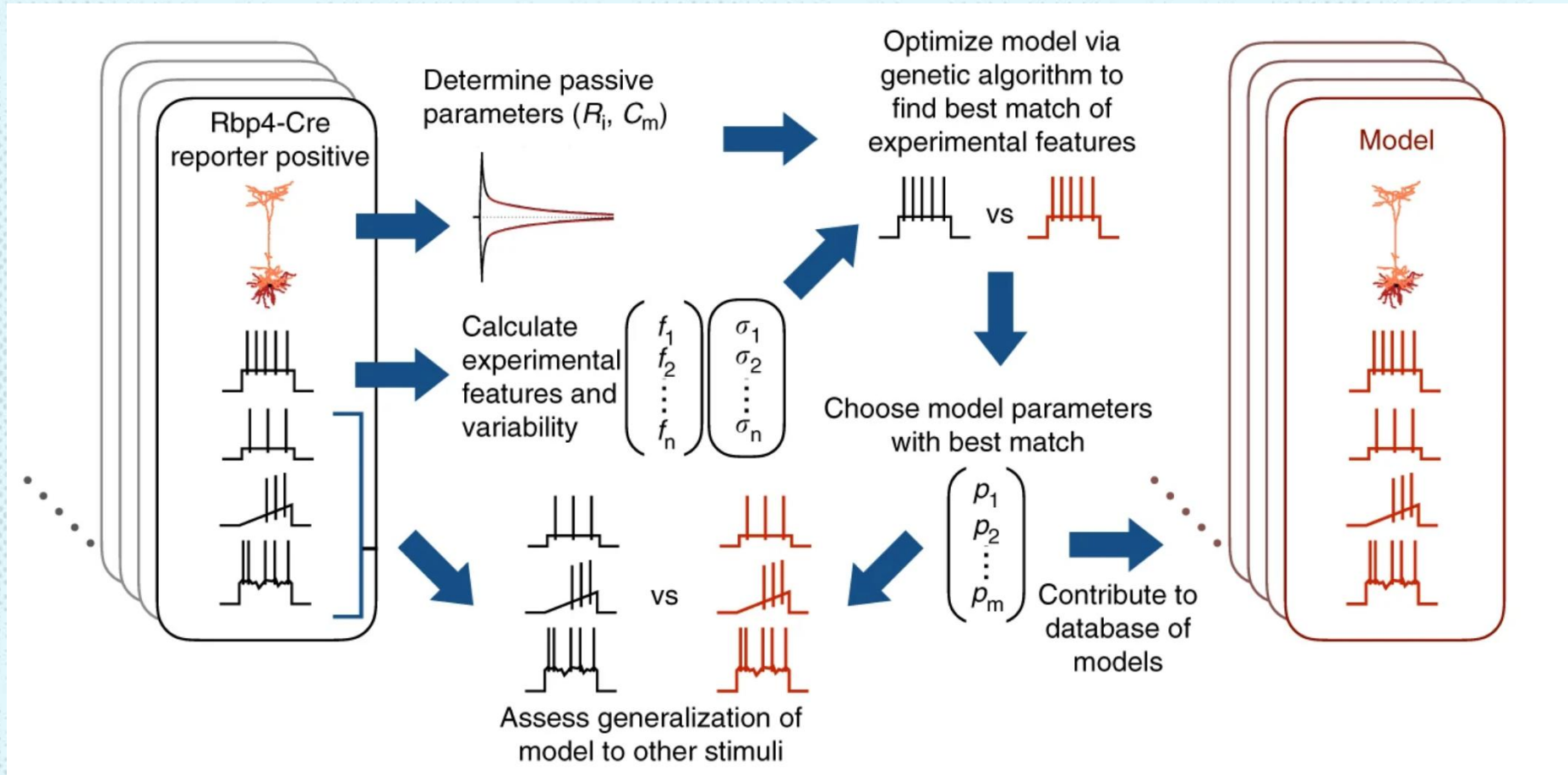
Simulation:
21 s



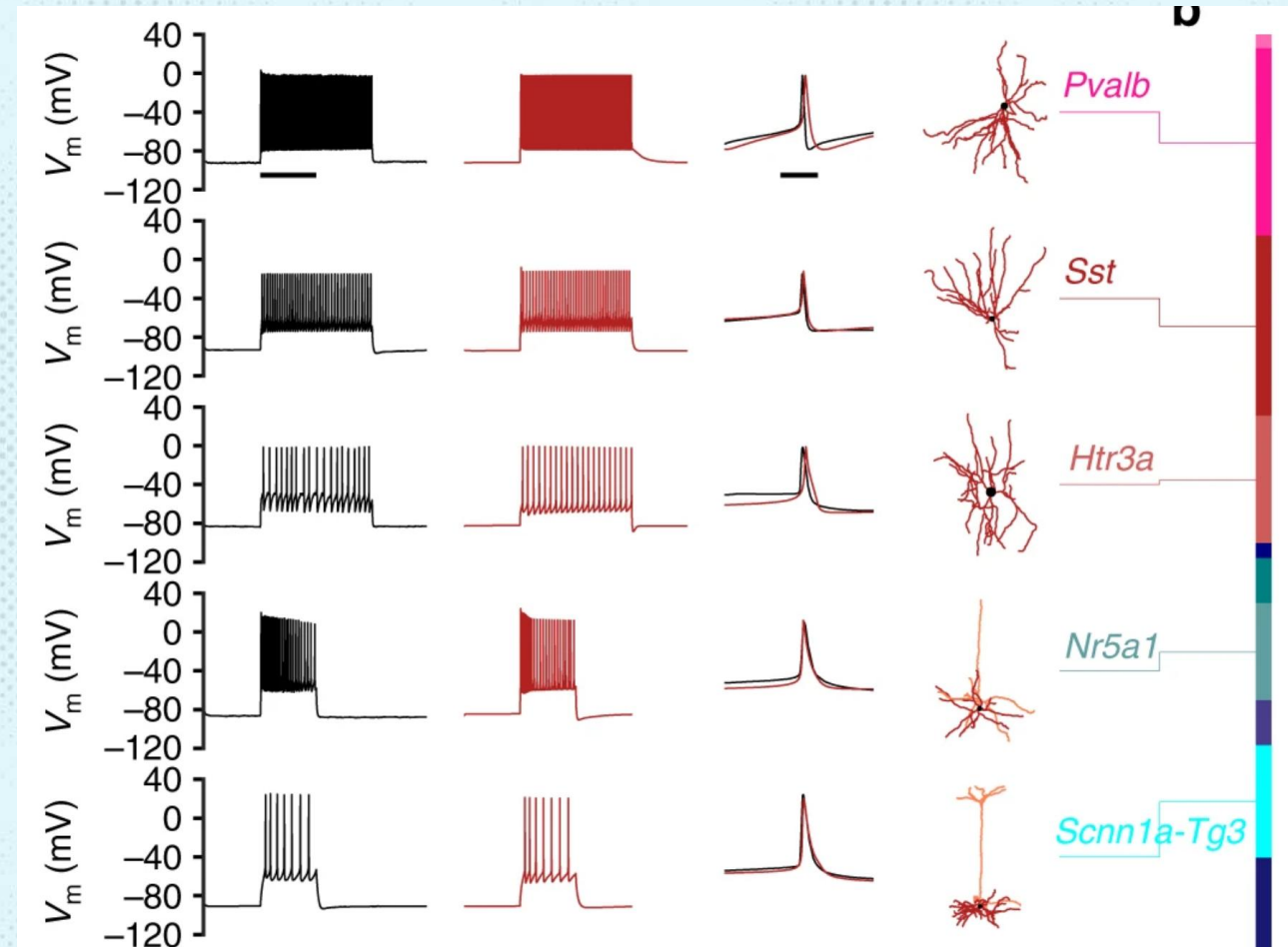
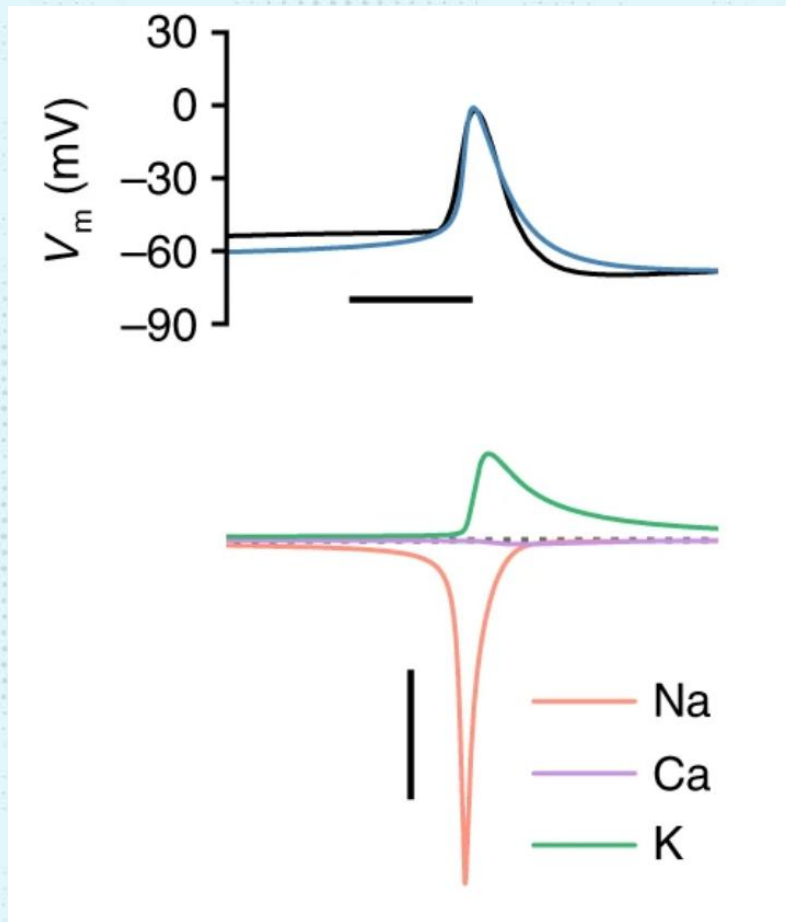
Adjusting parameters



Adjusting parameters with genetic algorithms

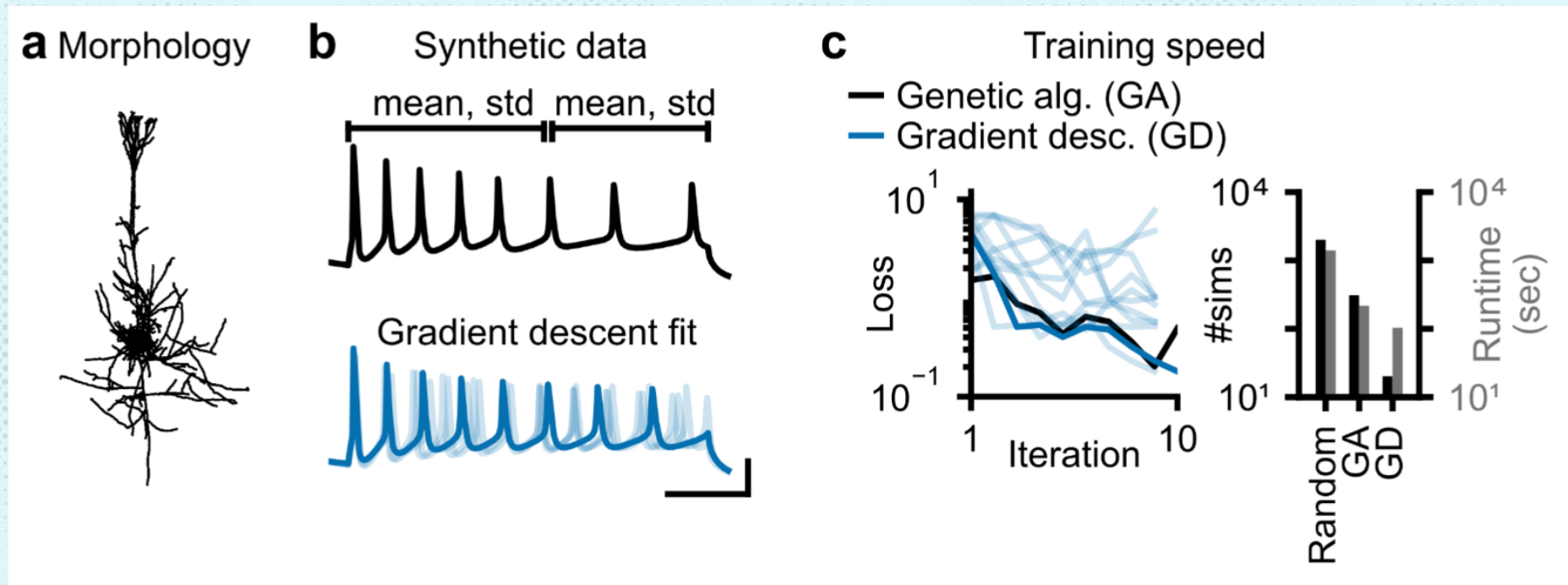


Adjusting parameters with genetic algorithms



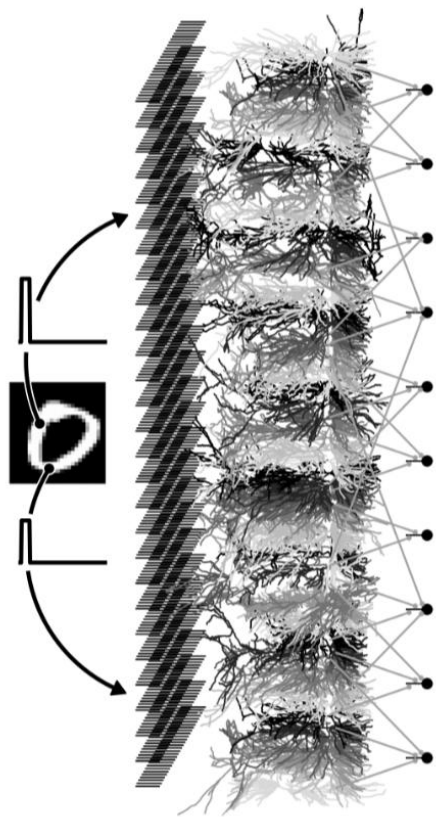
Adjusting parameters with gradient descent

- Idea: Use auto-diff functionality of jax to get gradients!

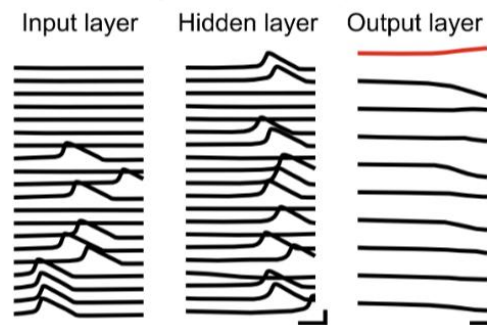


Allows to train biophysical models for ML tasks!

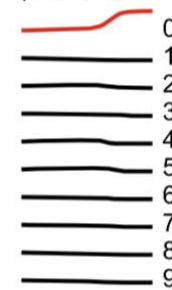
a Biophysical network to classify MNIST



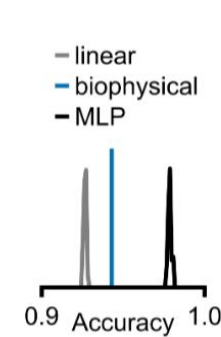
b Single neuron activity



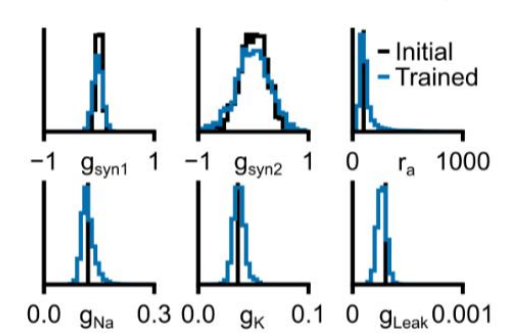
c Softmax probabilities



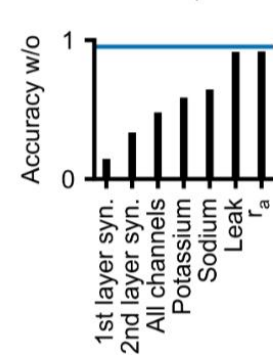
d Performance



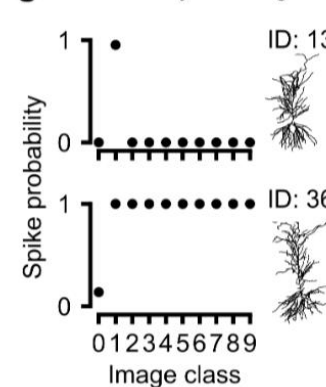
e Parameters before and after training



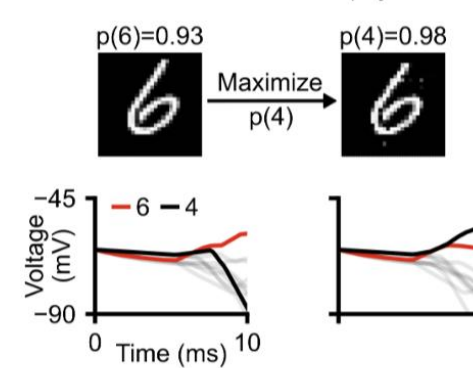
f Parameter importance



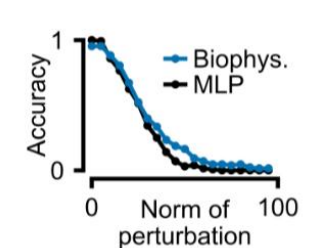
g Hidden layer tuning



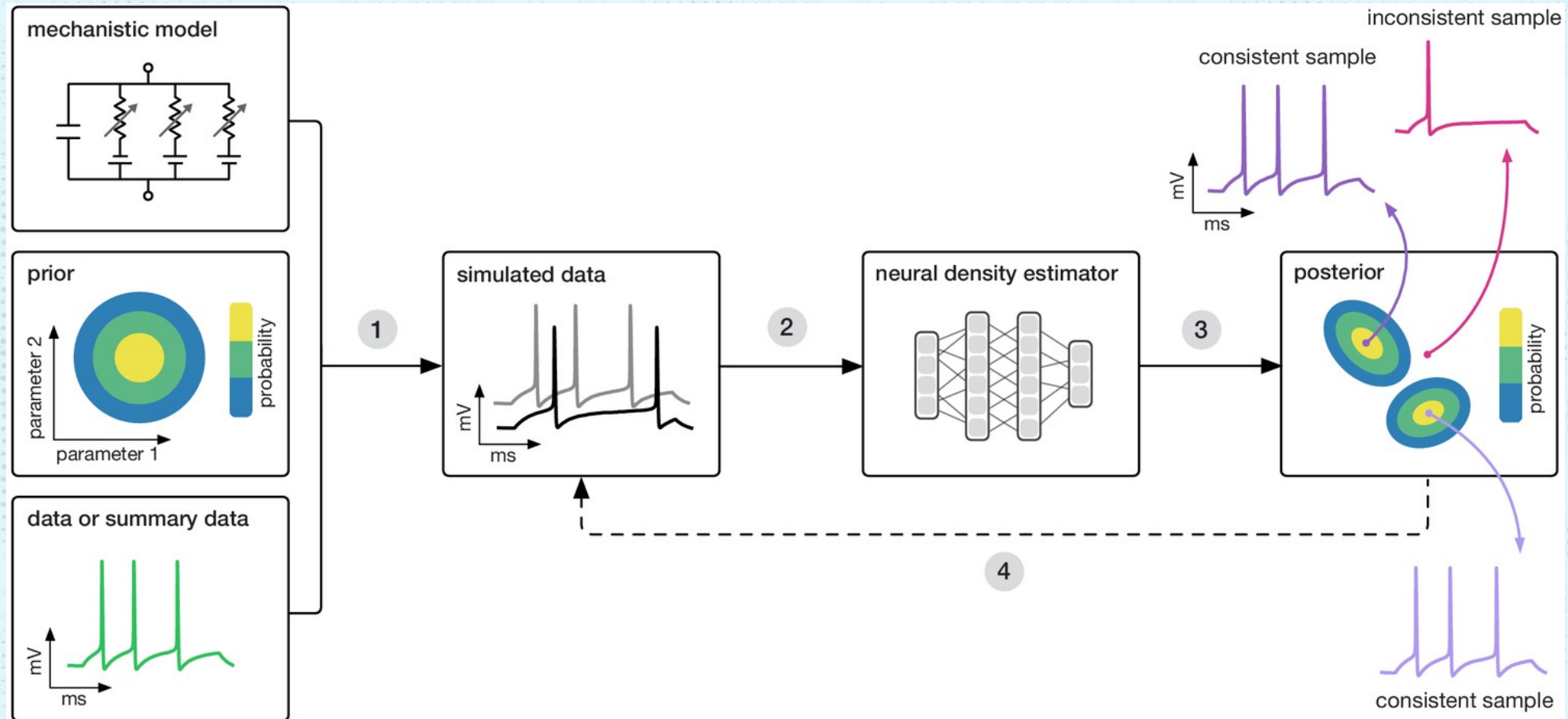
h Adversarial attack on biophysical net



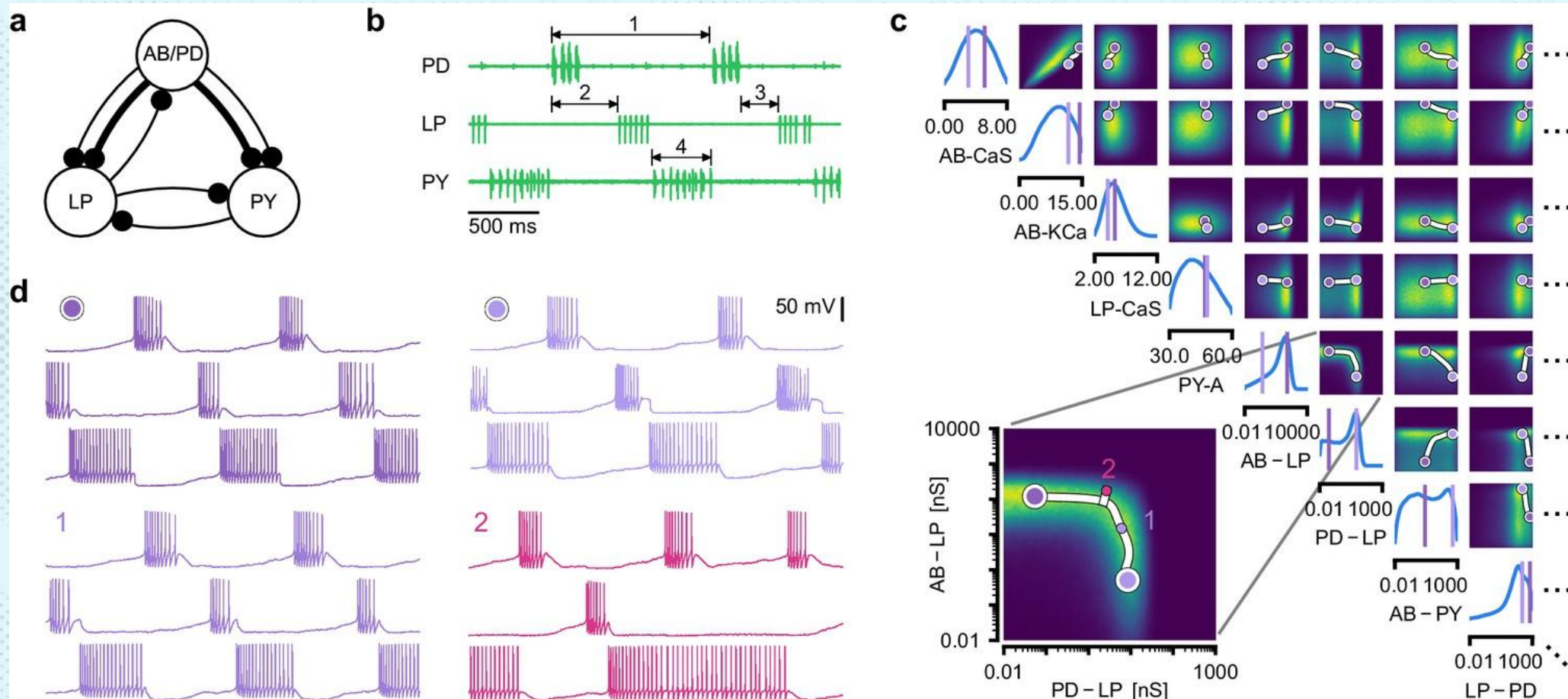
i Adversarial robustness



Adjusting parameters with simulation-based inference



SBI allows to identify multiple solutions



Recap

- Roles of models in science
- Passive neuron membranes
 - Equilibrium potential of important ions
 - RC circuit model – equation, response to step stimulus
- Active neuron membranes
 - Action potentials – time course, phases, involved ions
 - Hodgkin-Huxley model – equation, time course, ion channels
- Forward Euler integration
- Parameter estimation – algorithms