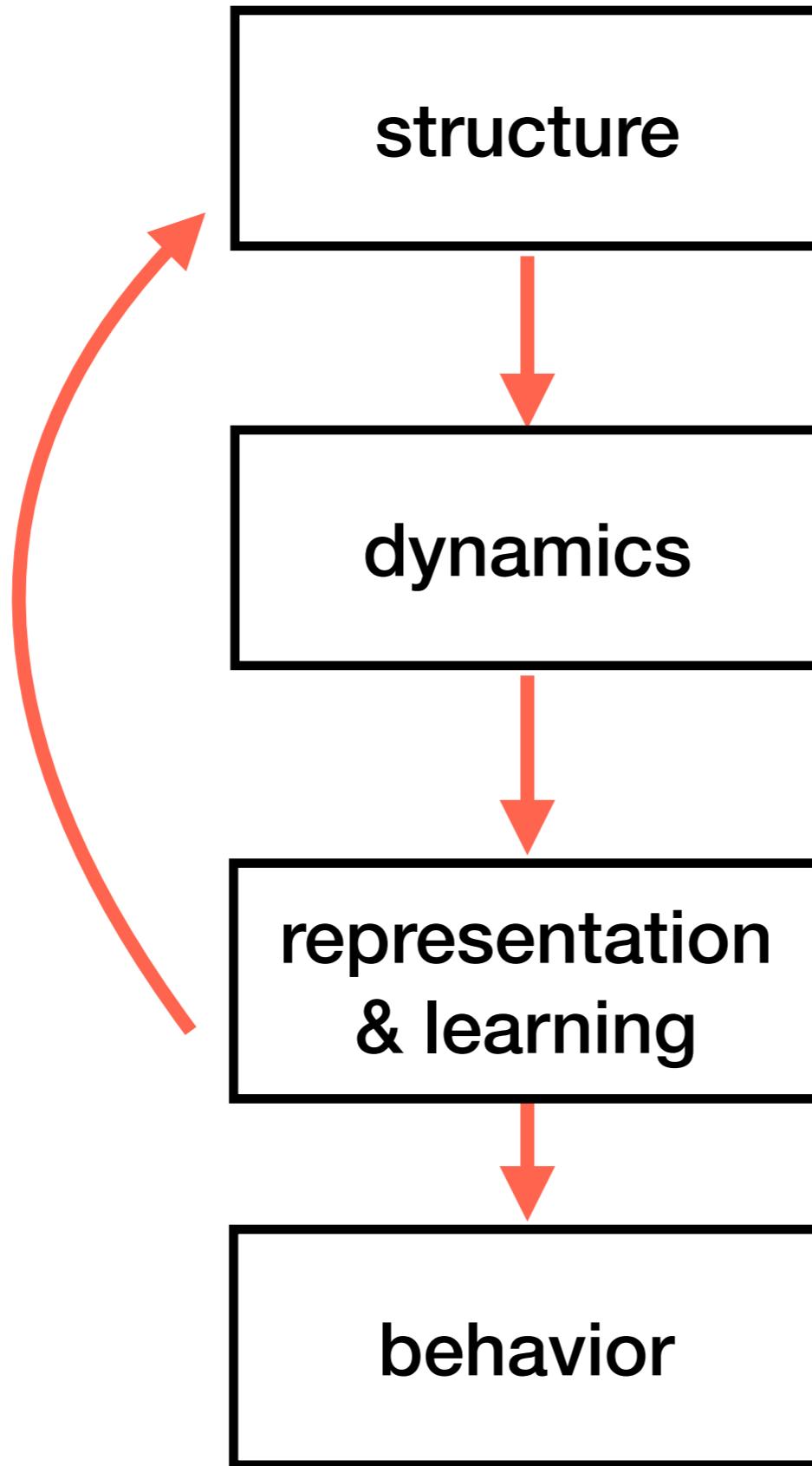


Single Neuron Computation



Course Curriculum

- **Introduction to Computational Neuroscience**
- **Single Neuron Computation**
- **Connect Neurons to Perform Computation**
- **Recurrent network and Attractor Dynamics**
- **E/I balance and Chaotic Dynamics**
- **Efficient coding principle**
- **Stimulus discrimination, Fisher information and Information-limited correlation**
- **Normative approach to neural computation and the emergence of sensory representation**
- **Dimensionality reduction and motor representation**
- **How does an animal move, an integrated approach, and insights from *C. elegans***
- **Reward and Reinforcement Learning**

<https://github.com/Wenlab/Computation-Neuro-Course/tree/Fall2024>

The screenshot shows a GitHub repository page with a dark theme. At the top, there are links for 'README' and 'MIT license'. Below the title, there is a brief description of the repository's purpose. Underneath, there are sections for 'When', 'Where', 'Teacher', 'Teaching Fellow', 'Recommended Textbooks', 'For a general reader', 'Course Performance Evaluation', and 'Prerequisite'. Each section contains a bulleted list of links.

README MIT license

Computational-Neuroscience-Course

This repository serves as an ongoing effort to create a systems and computational neuroscience course. I hope to develop a two-semester course for USTC biophysics students. The first semester will cover some basic materials, and the second semester will introduce more advanced topics for undergraduate and graduate students.

When: Thursday 9:45 am - 11:20 am, Fall 2024

Where: 东区第二教学楼2403

Teacher: 温泉 gwen@ustc.edu.cn

Teaching Fellow: 洗奇琪 xianqiqi@mail.ustc.edu.cn 李懿轩 hyperdunk2019@mail.ustc.edu.cn

Recommended Textbooks:

- [Theoretical Neuroscience: Computational and Mathematical Modeling the Neural System](#)
- [Principles of Neural Design](#)

For a general reader:

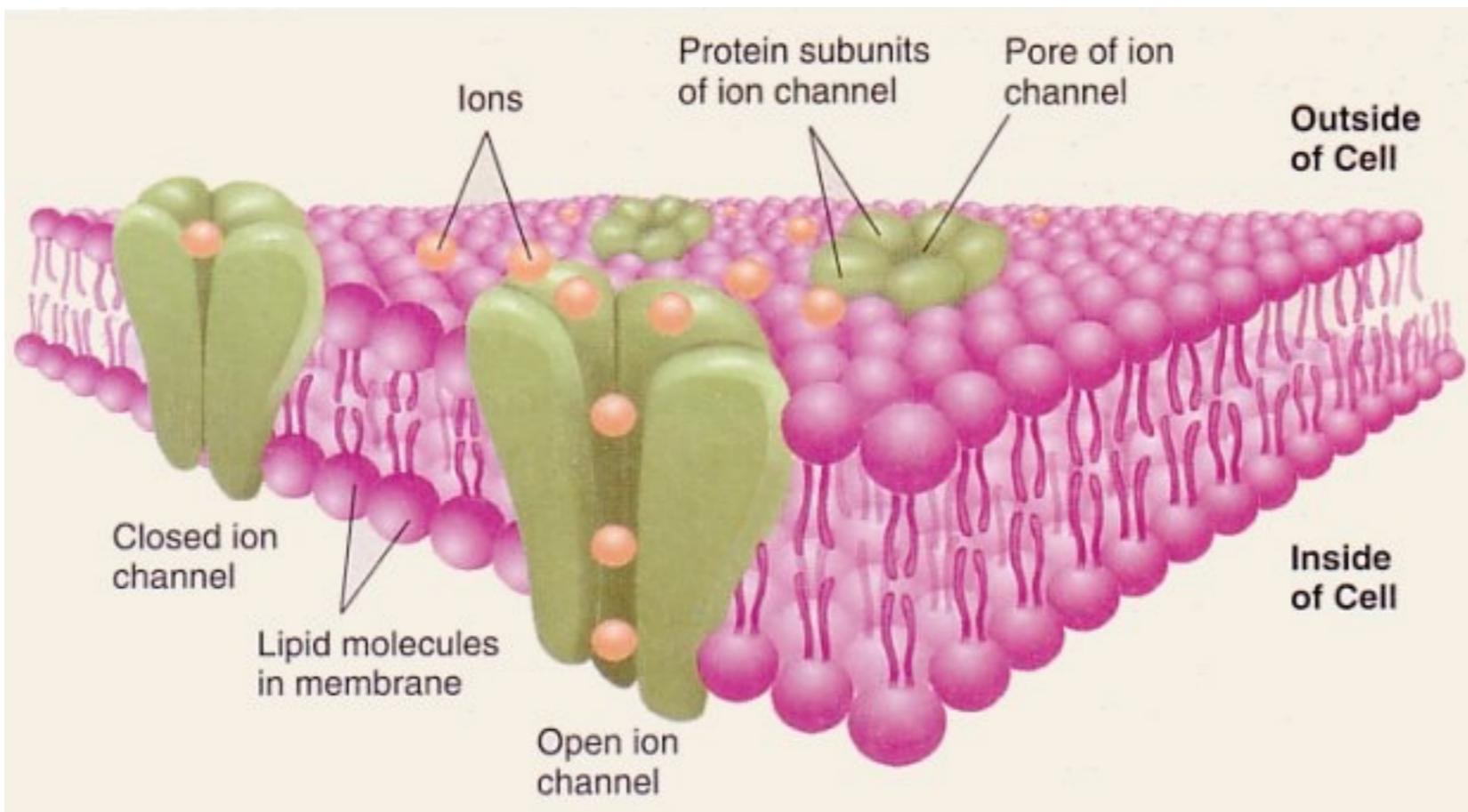
- [Model of the Mind](#)

Course Performance Evaluation:

- Homework: 70%
- Final: 30% (take-home exam)

Prerequisite:

- High school knowledge of biology and neuroscience

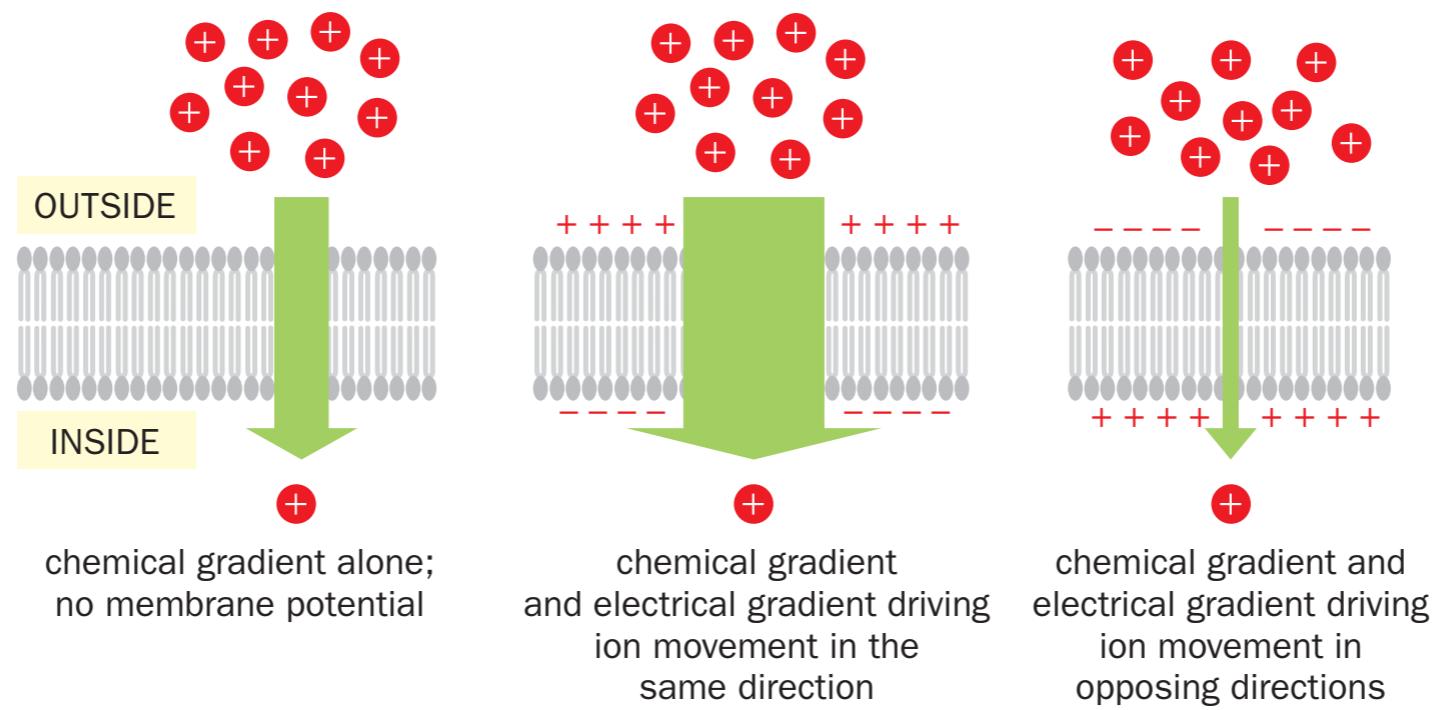


$$qV \sim k_B T$$

$$k_B = 8.6 \times 10^{-5} eV/K$$

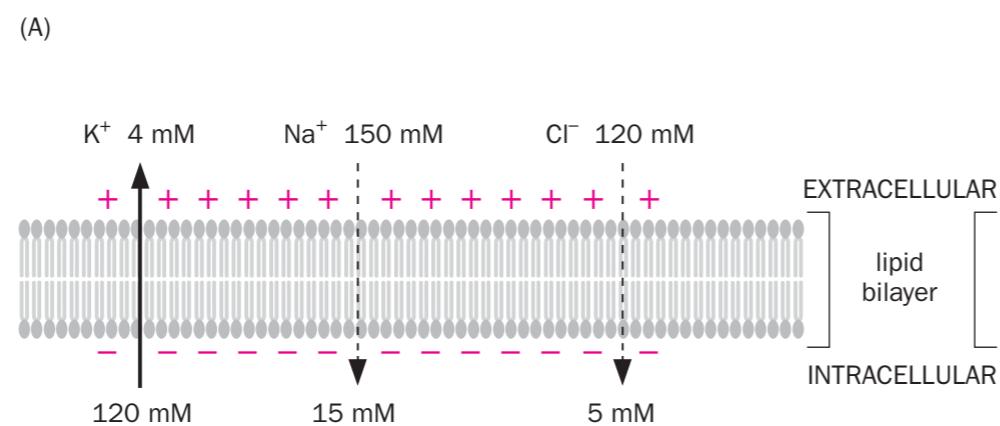
$$T = 300 K$$

What determines the resting potential of a neuron?

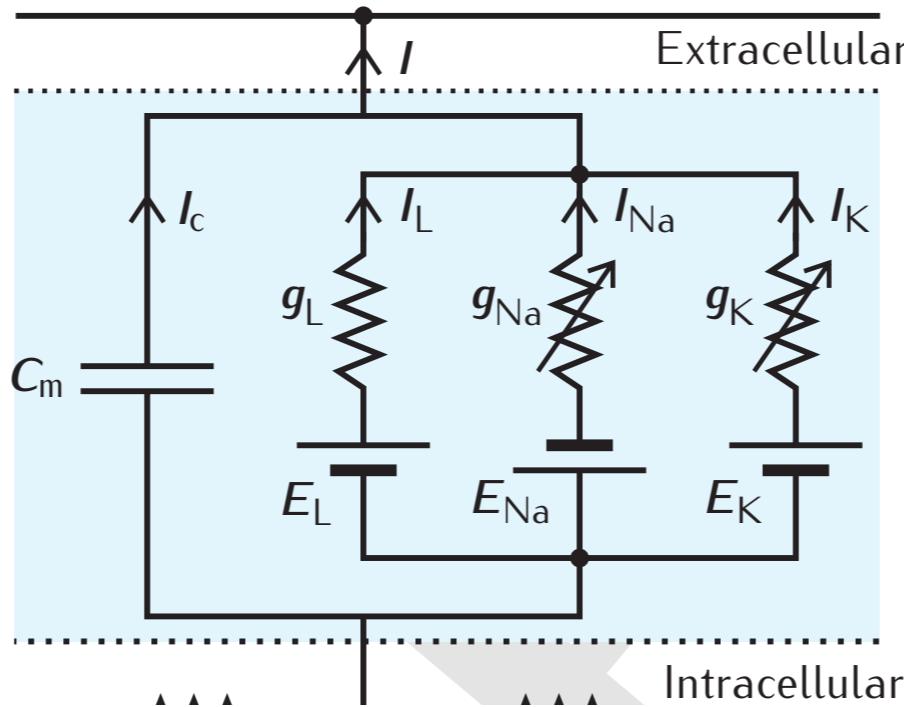


$$E = \frac{k_B T}{ze} \ln \frac{n_{out}}{n_{in}}$$

Ion pump



The Equivalent Electronic Circuit of a Neuron



$$C_m \frac{dV}{dt} = - \sum_i g_i(V)(V - E_i) - g_L(V - E_L) - I_e$$

$$C_m \approx 0.1 - 1 \text{ nF}$$

$$r_m = 1/g \approx 10 - 100 \text{ M}\Omega$$

$$\tau_m = C_m r_m \approx 10 - 100 \text{ ms}$$

Squid Giant Axon

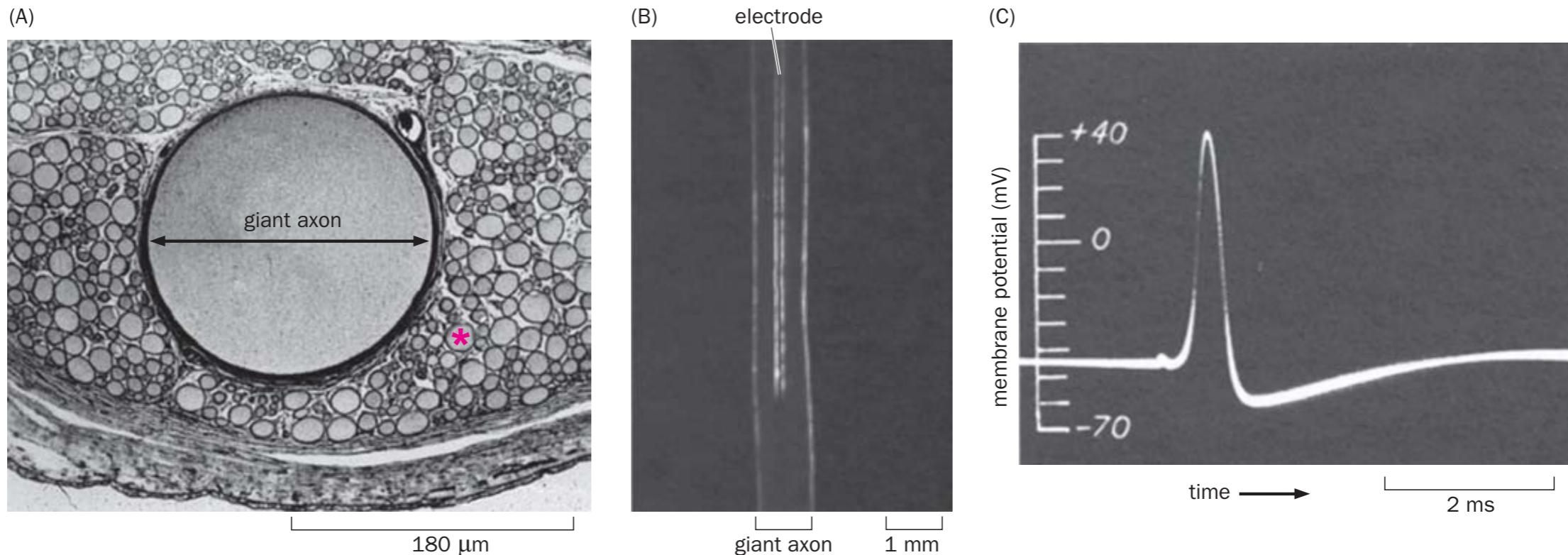
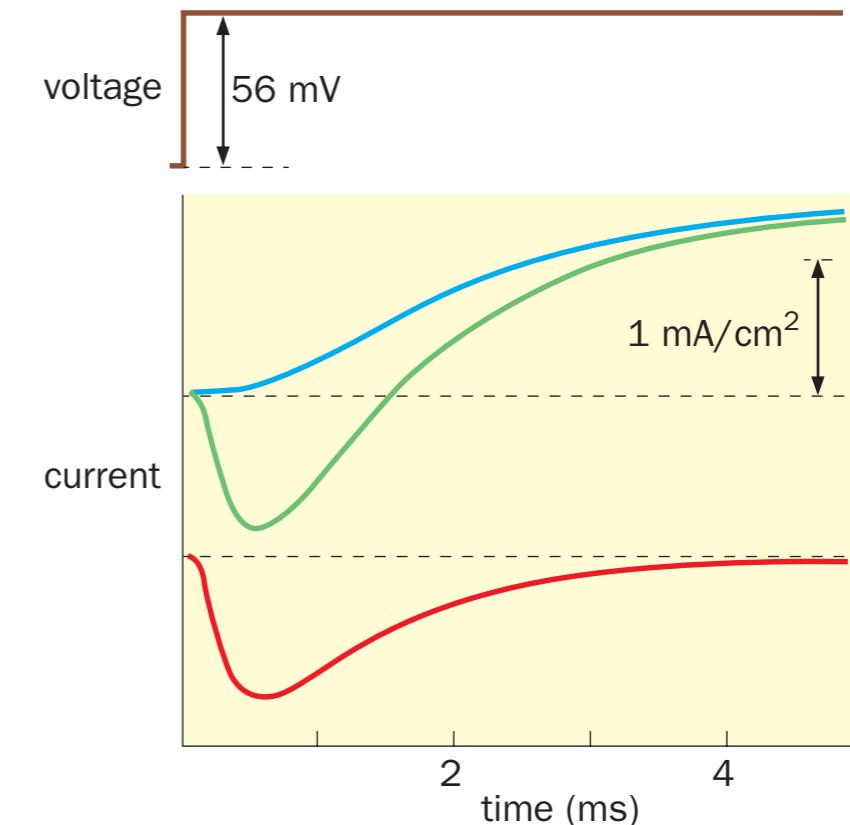
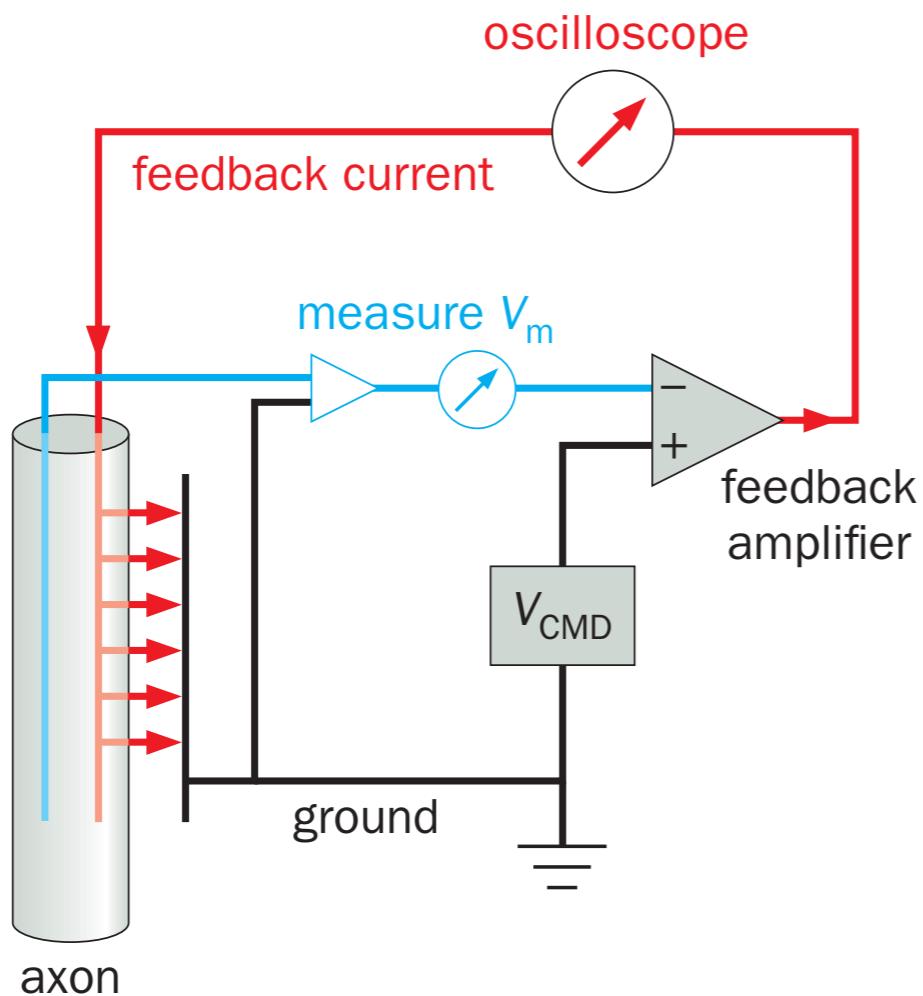
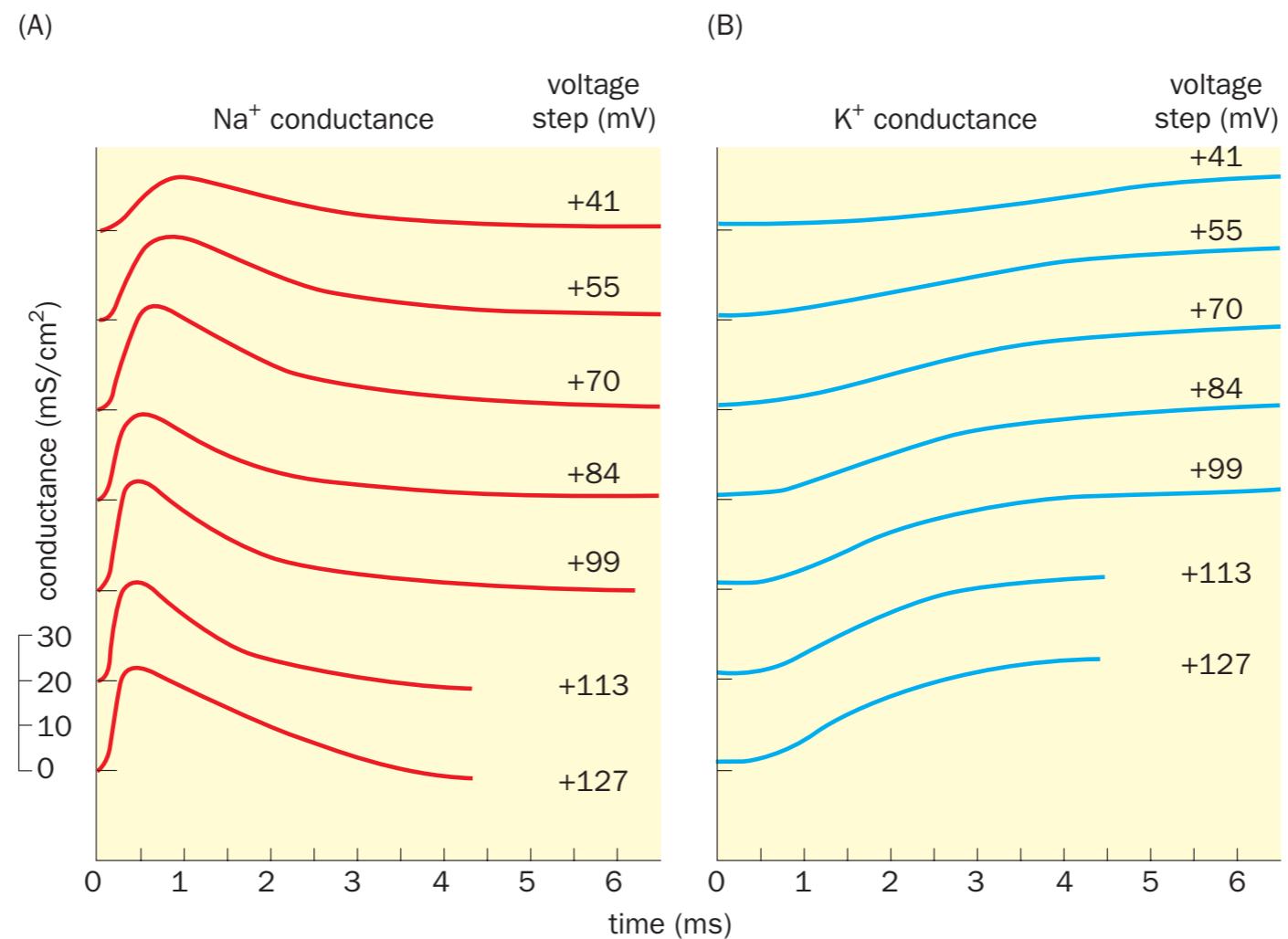


Figure 2–19 Studying action potentials in the squid giant axon. **(A)** Electron micrograph of a cross section of a squid giant axon showing its large diameter ($\sim 180 \mu\text{m}$ for this sample) as compared to neighboring axons (for example, the axon indicated by *). **(B)** Photograph of an electrode inserted inside a squid giant axon whose diameter is close to 1 mm. **(C)** An action potential recorded from the squid giant axon. (A, courtesy of Kay Cooper and Roger Hanlon; B, from Hodgkin AL & Keyes RD [1956] *J Physiol* 131:592–616; C, from Hodgkin AL & Huxley AF [1939] *Nature* 144:710–711. With permission from Macmillan Publishers Ltd.)

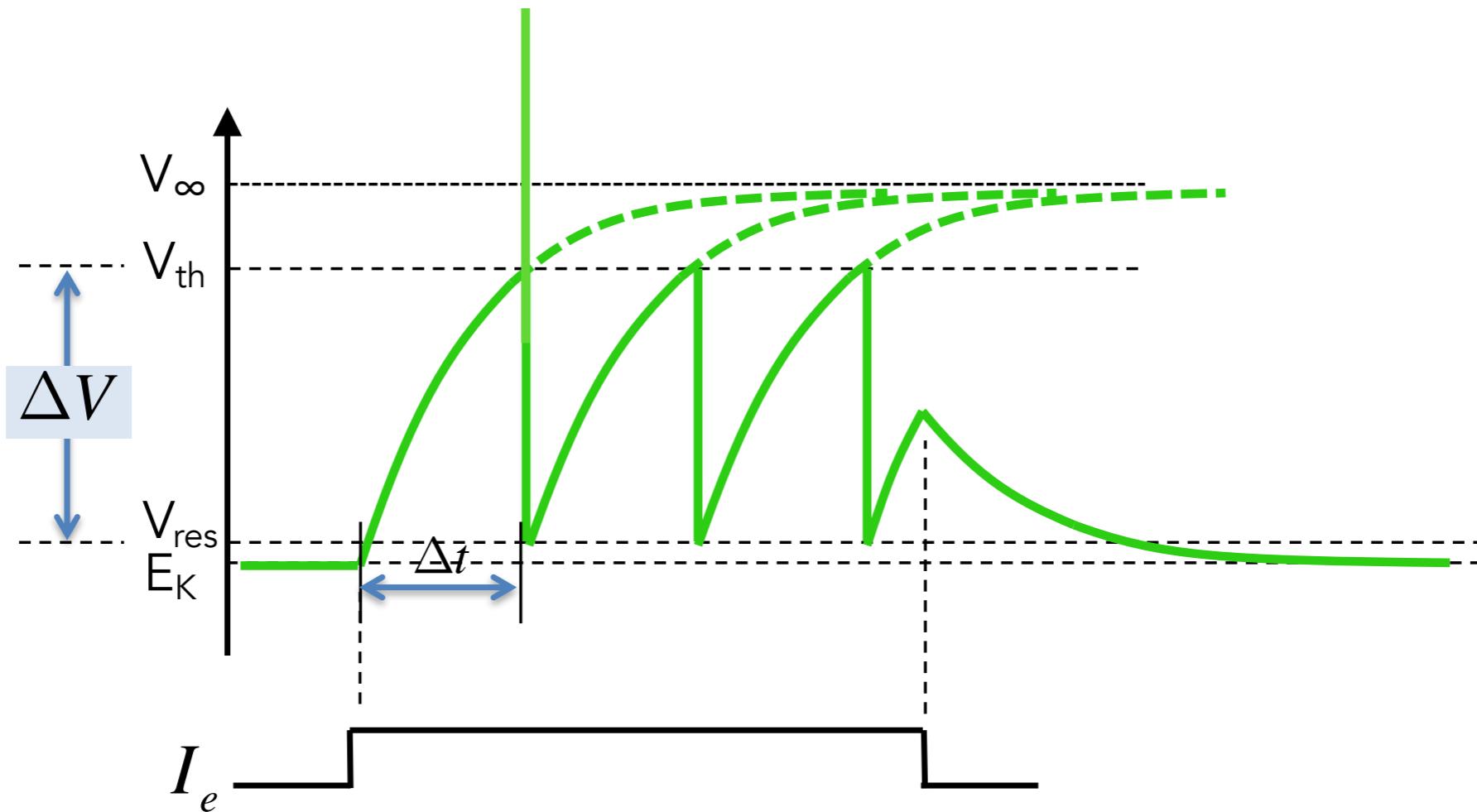
Voltage Clamp Recording



Voltage-gated Conductance



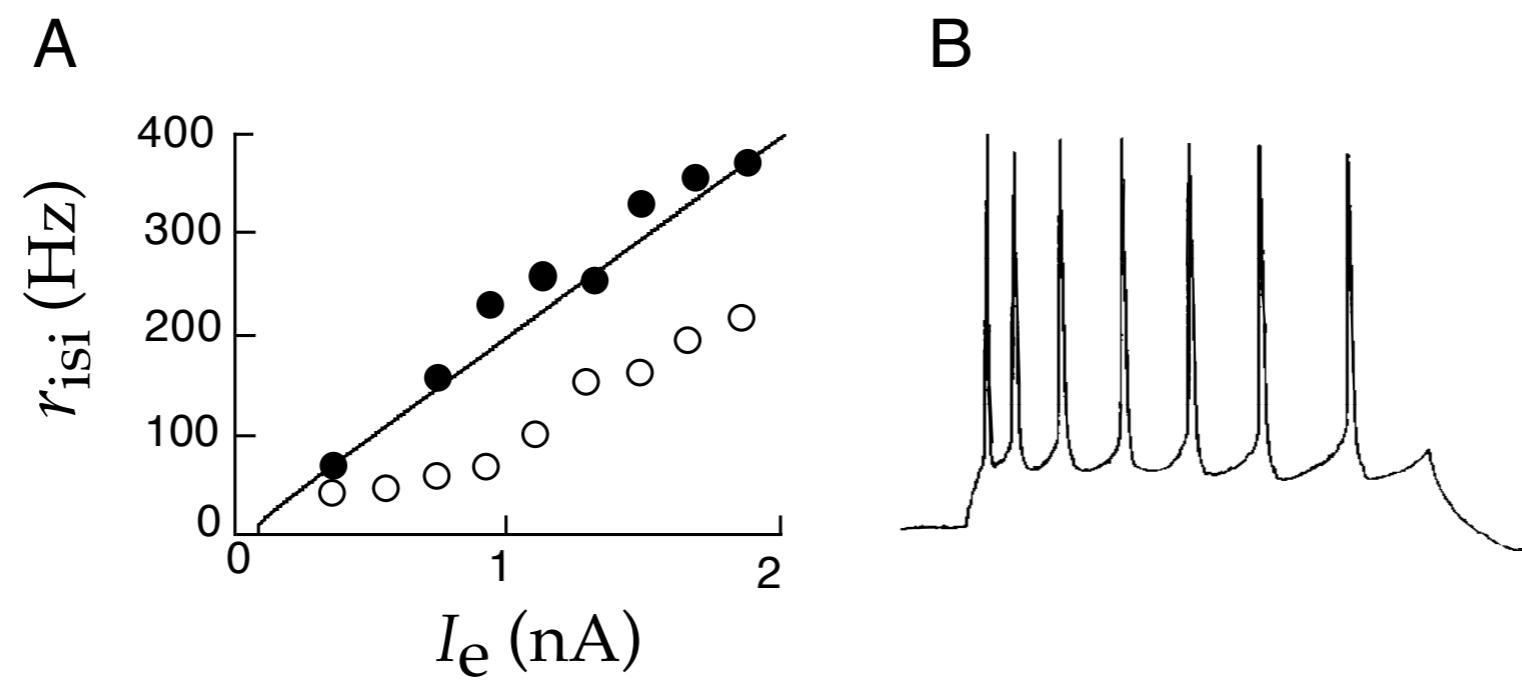
Integrate-and-Fire model



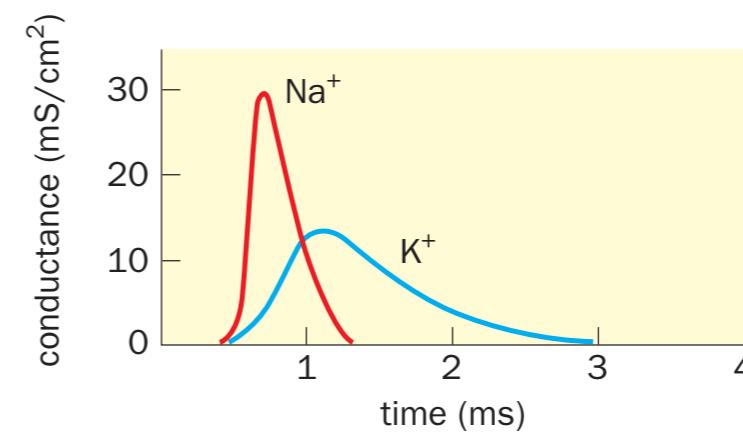
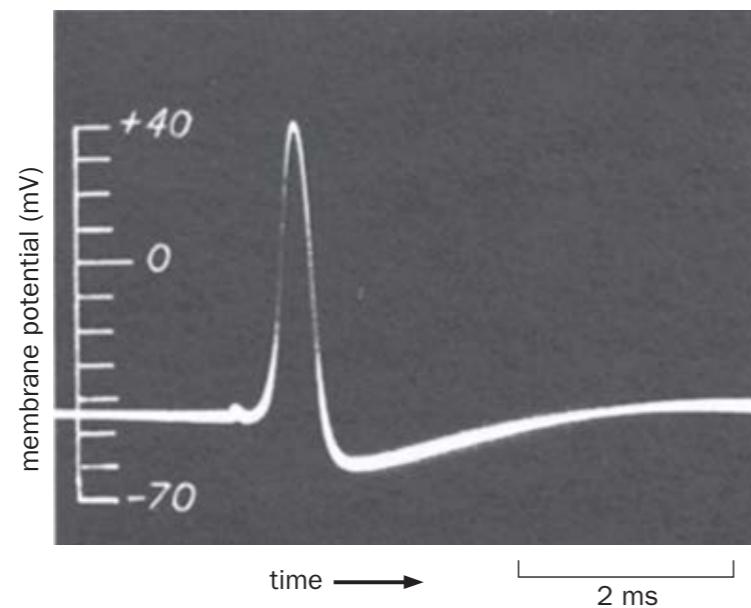
$$C \frac{dV}{dt} = -g(V - E_K) + I_e$$

$$V(t_{spike}^-) = V_{th}$$

$$V(t_{spike}^+) = V_{res}$$

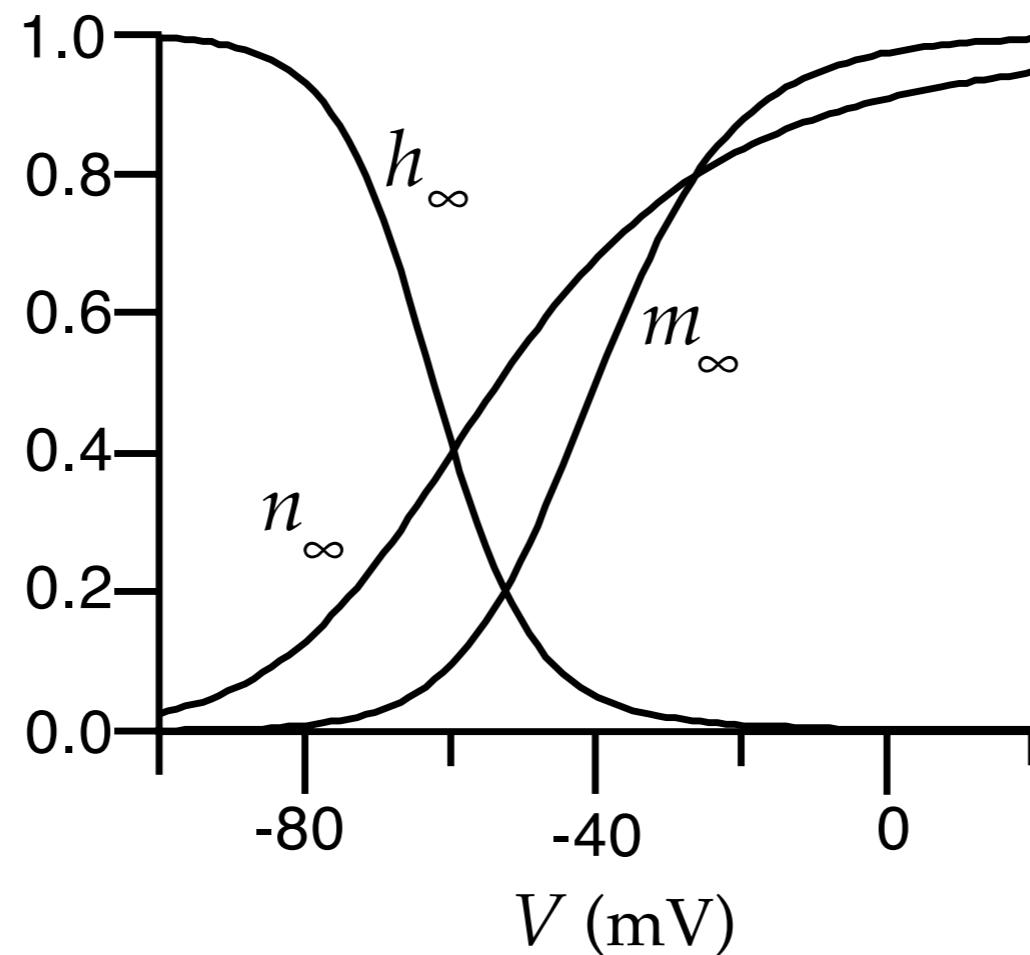


(C)



$$C \frac{dV}{dt} = -g_K n^4 (V - E_K) - g_{Na} m^3 h (V - E_{Na}) - g_L (V - E_L) - I_e$$

Transient Na^+ channel conductance

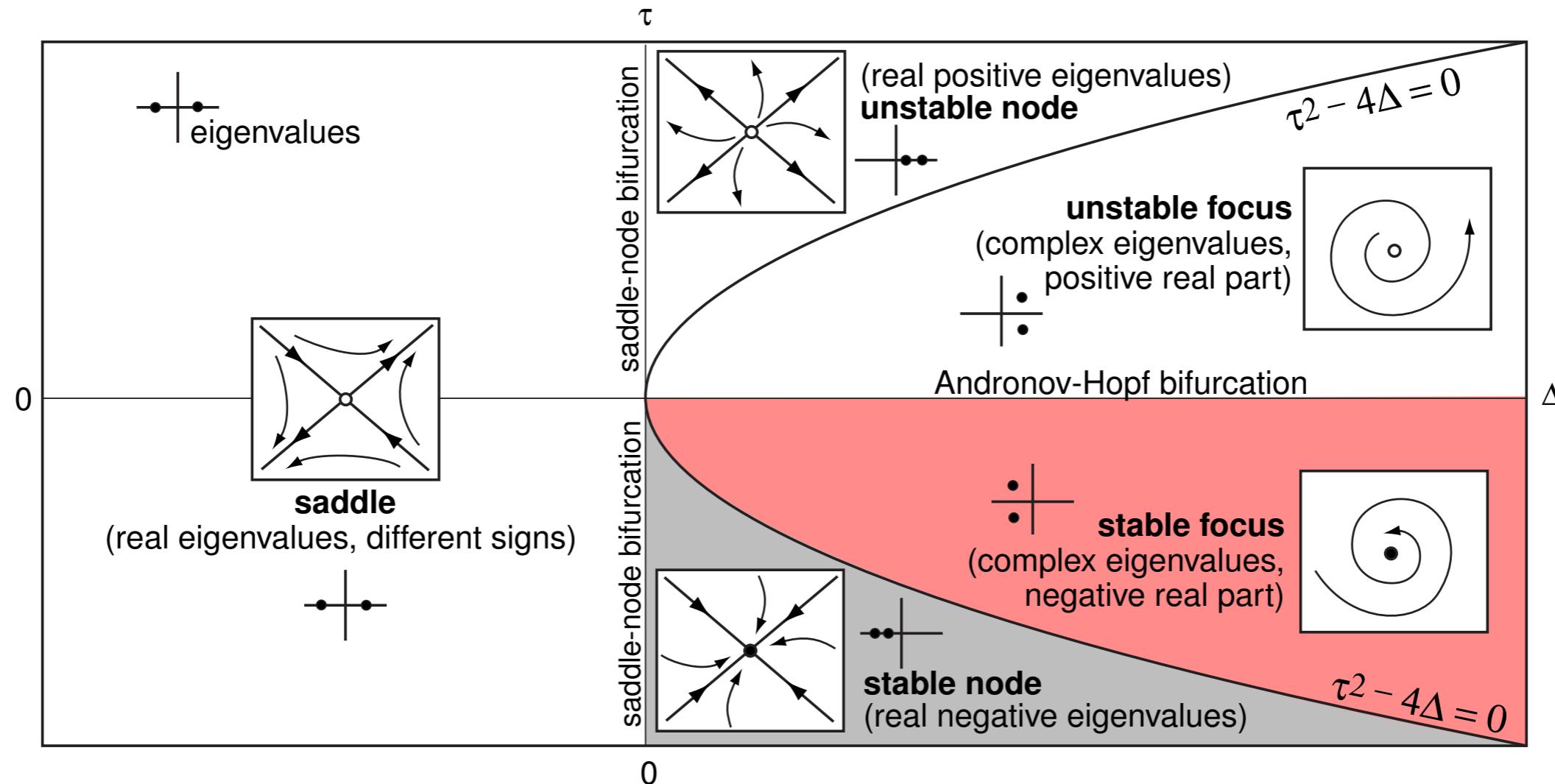


Fix Point one dimension

$$\frac{dV}{dt} = F(V)$$

$$F(V^*) = 0$$

Fix Point in two dimensions

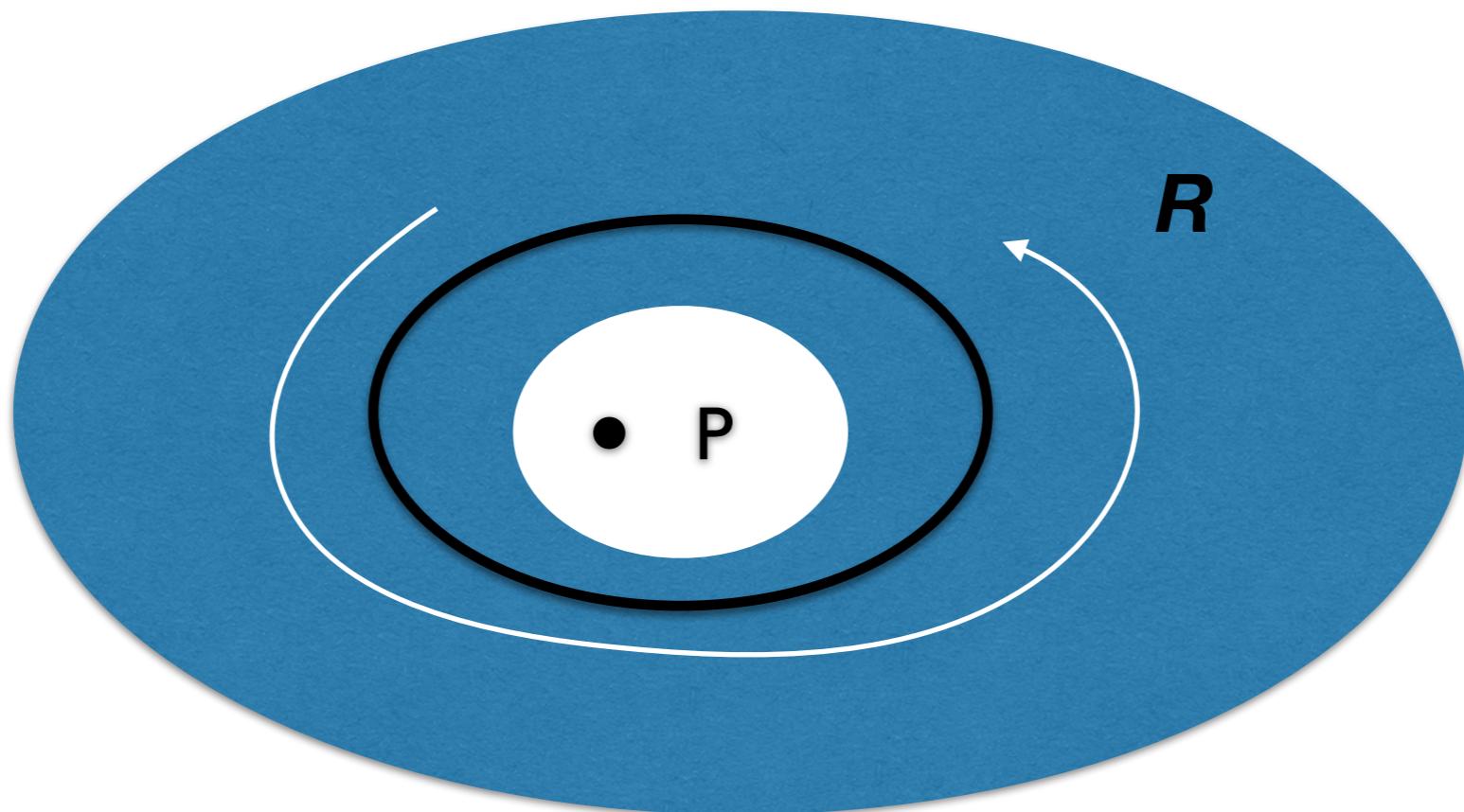


Poincare-Bendixson Theorem (2D)

- R is a closed, bounded subset of the plane;
- $\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x})$ is a continuously differentiable vector field on an open set containing R ;
- R does not contain any fixed points; and
- There exists a trajectory C that is “confined” in R , in the sense that it starts in R and stays in R for all future time

Then either C is a closed orbit, or it spirals toward a closed orbit at $t \rightarrow \infty$. In either case, R contains a closed orbit.

Poincare-Bendixson Theorem (2D)



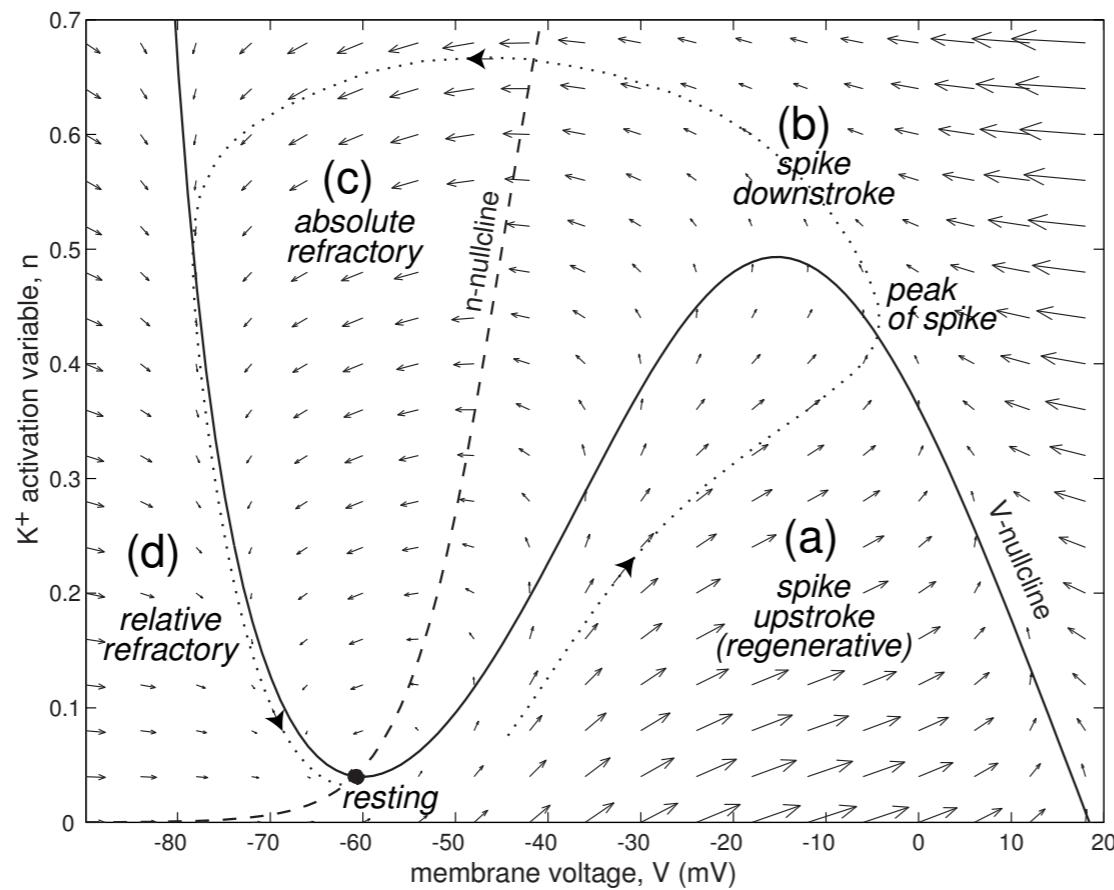


Figure 4.4: Nullclines of the $I_{Na,p} + I_K$ -model (4.1, 4.2) with low-threshold K^+ current in Fig.4.1b. (The vector field is slightly distorted for the sake of clarity of illustration).

Dynamic systems in Neuroscience

Eugene Izhikevich 2007

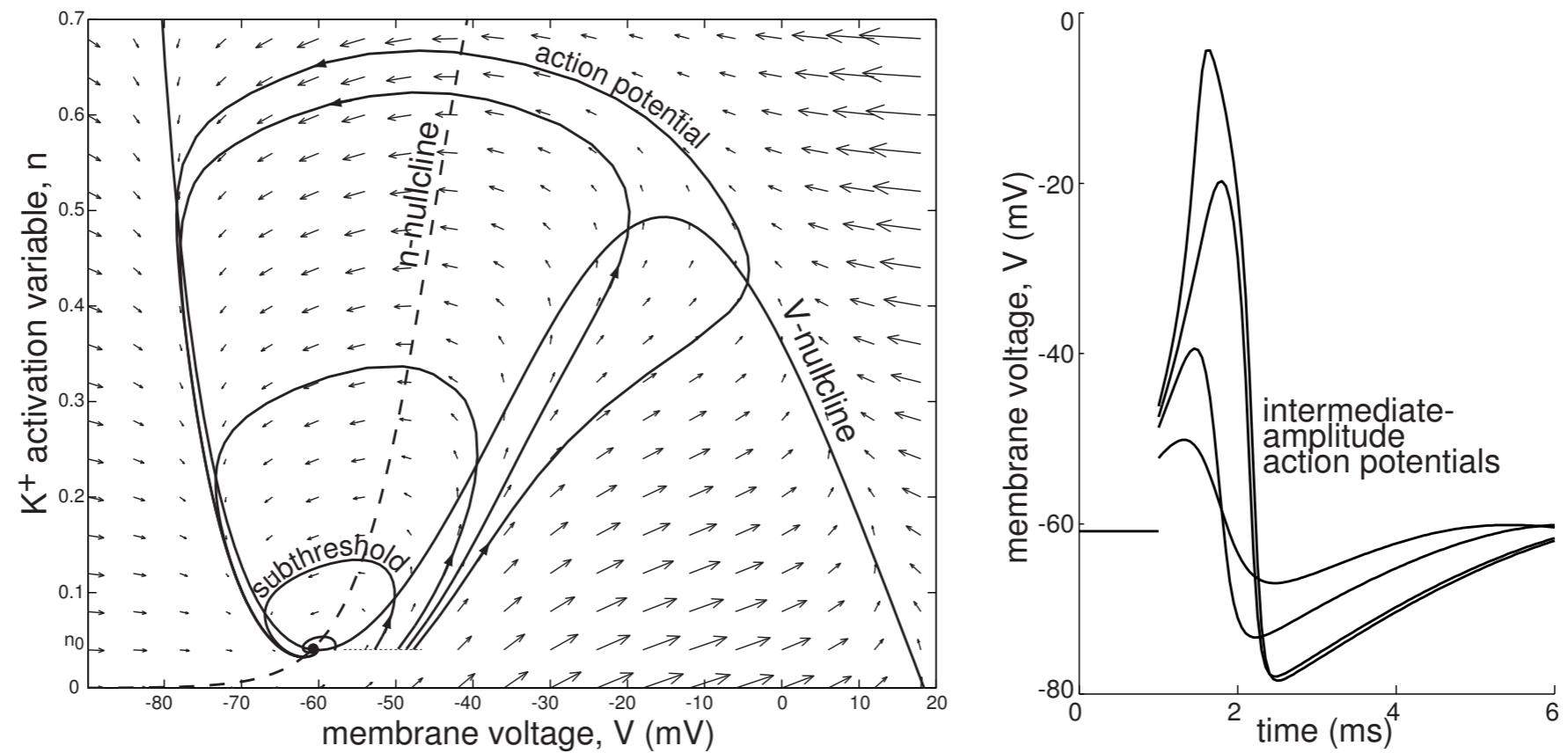
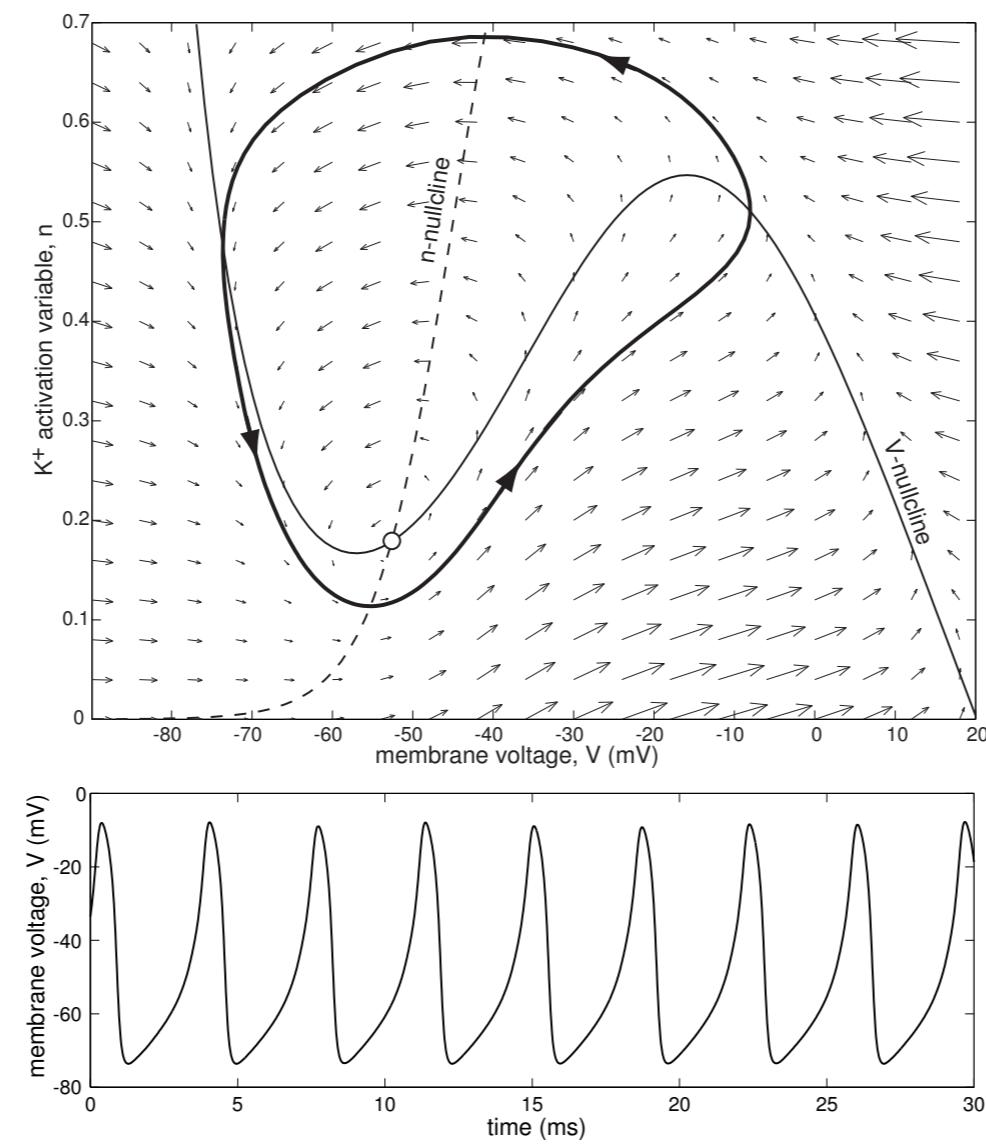


Figure 4.7: Failure to generate all-or-none action potentials in the $I_{Na,p} + I_K$ -model.

Dynamic systems in Neuroscience

Eugene Izhikevich 2007

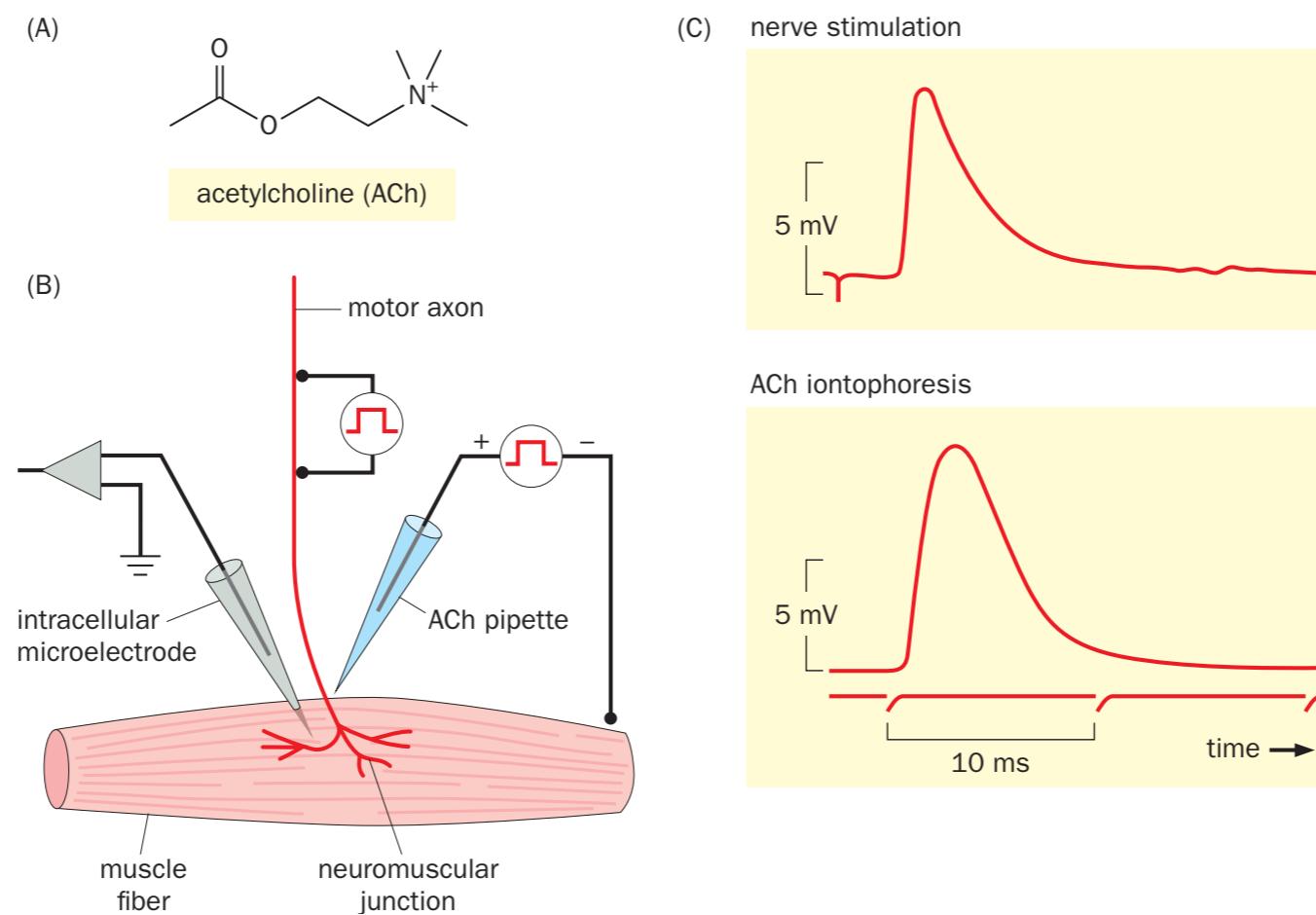


Dynamic systems in Neuroscience

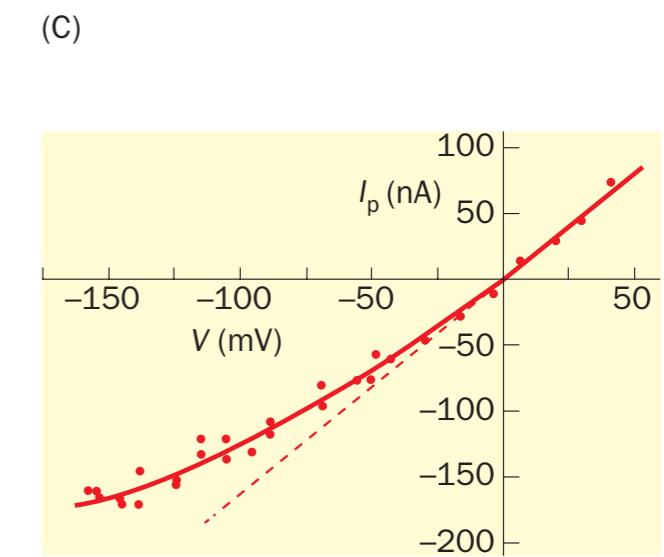
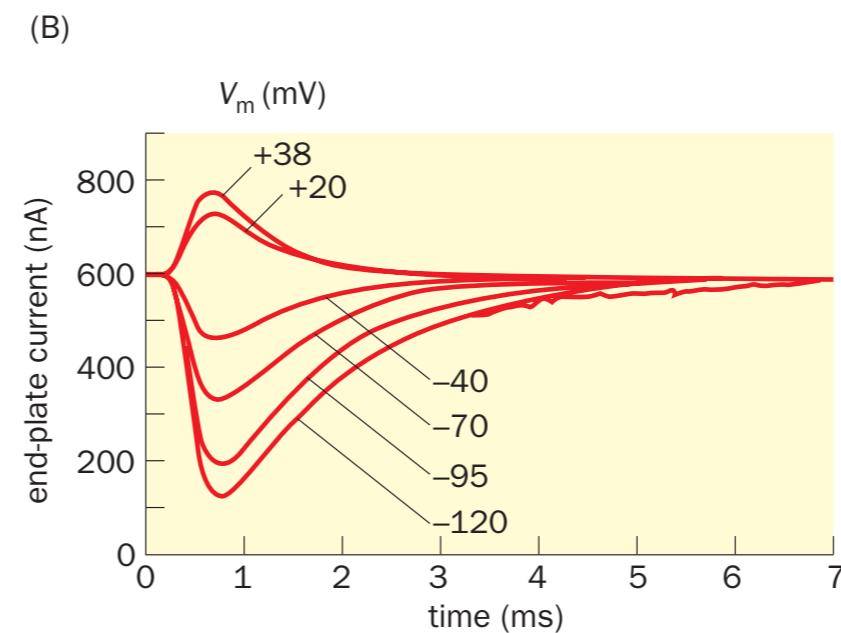
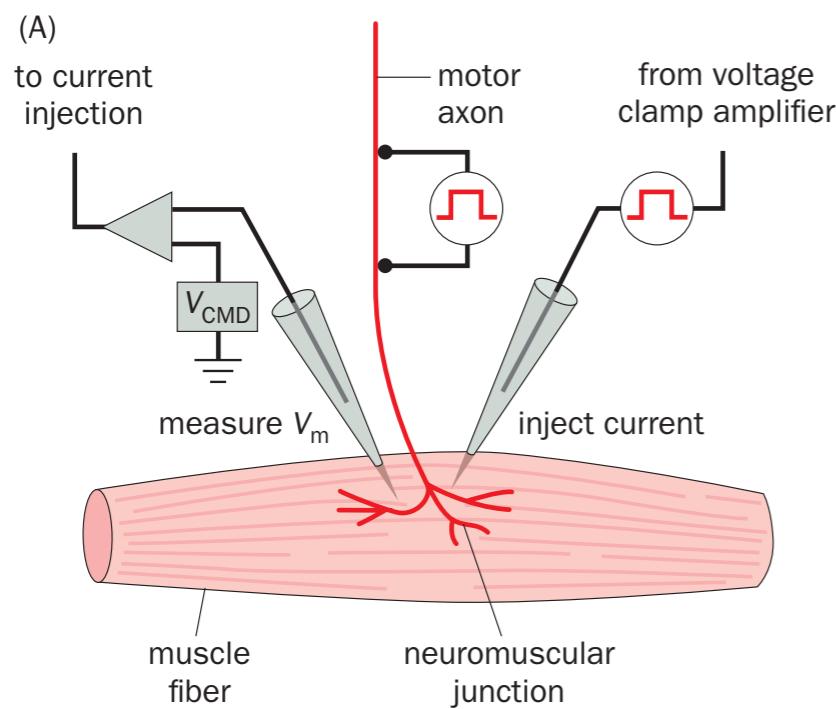
Eugene Izhikevich 2007

Synaptic Transmission

Neurotransmitter release evokes membrane potential change in the postsynaptic neuron



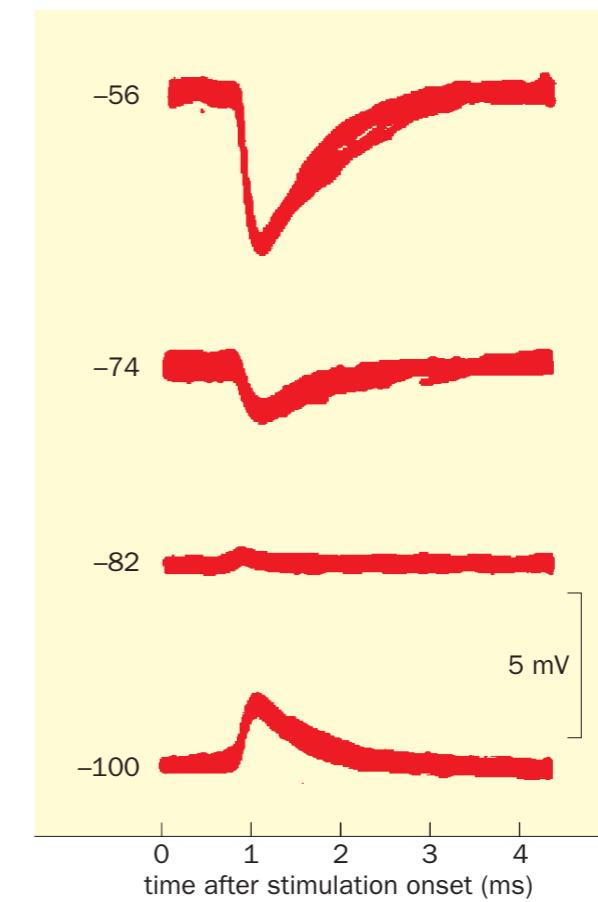
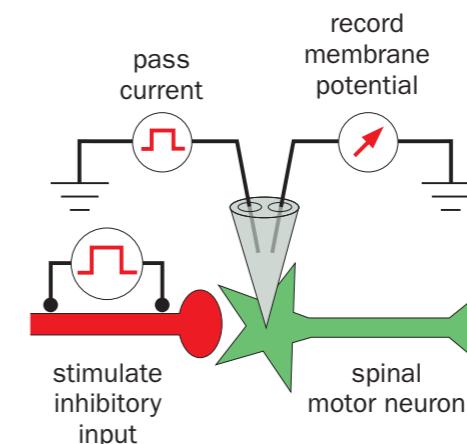
Reversal potential of a synapse



Excitatory and Inhibitory Synapses

Table 3–2: Commonly used neurotransmitters

Neurotransmitter	Major uses in the vertebrate nervous system ¹
Acetylcholine	motor neurons that excite muscle; ANS ² neurons; CNS excitatory and modulatory neurons
Glutamate	most CNS excitatory neurons; most sensory neurons
GABA	most CNS inhibitory neurons
Glycine	some CNS inhibitory neurons (mostly in the brainstem and spinal cord)
Serotonin (5-HT)	CNS modulatory neurons
Dopamine	CNS modulatory neurons
Norepinephrine	CNS modulatory neurons; ANS ² neurons
Histamine	CNS modulatory neurons
Neuropeptides	usually co-released from excitatory, inhibitory, or modulatory neurons; neurosecretory cells

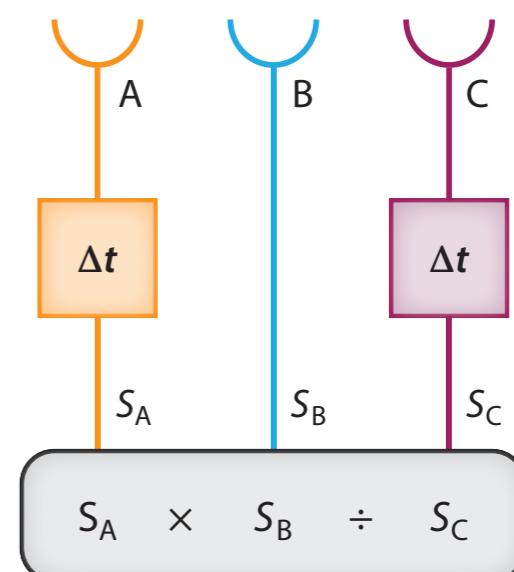
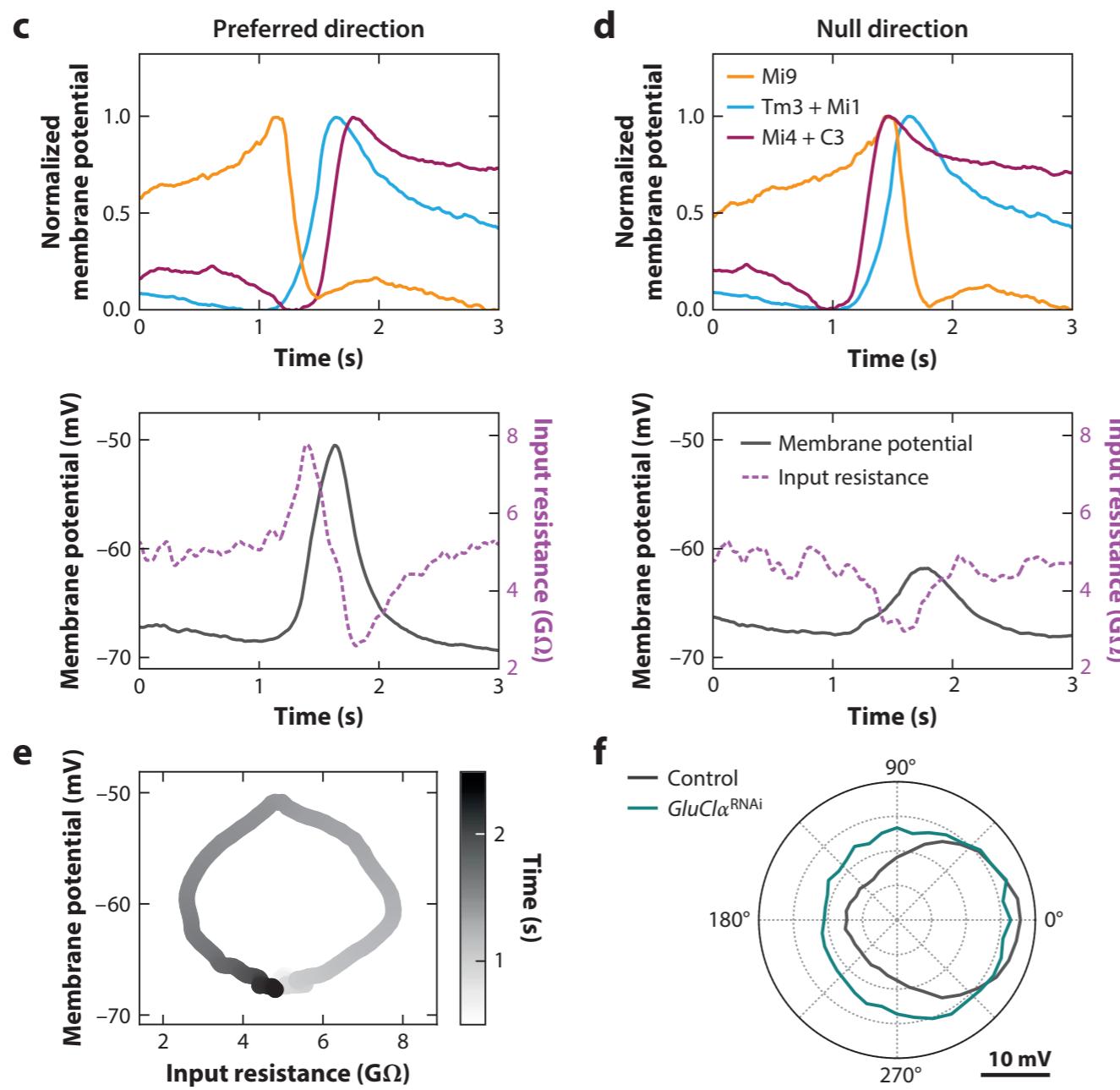
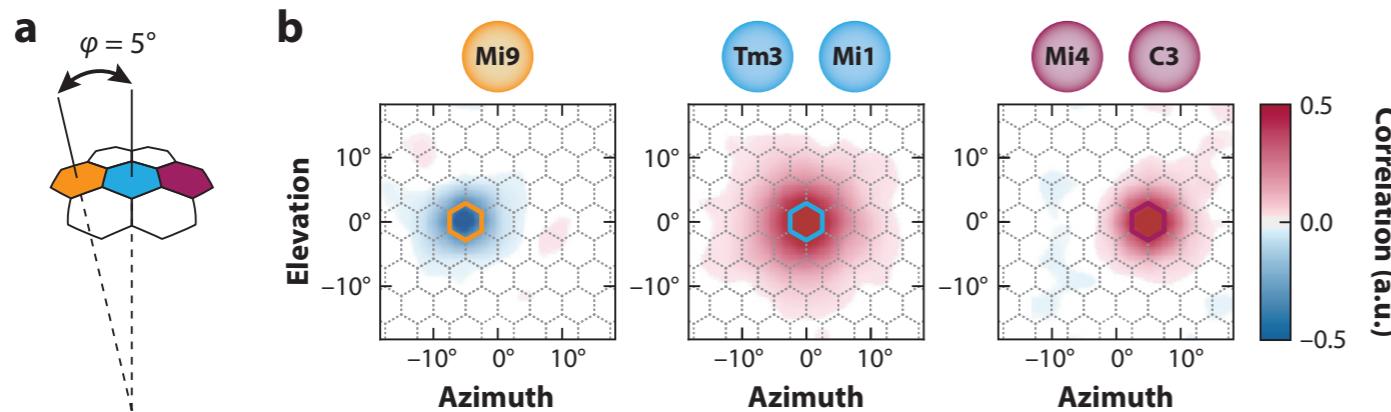


Shunting inhibition and divisive operation

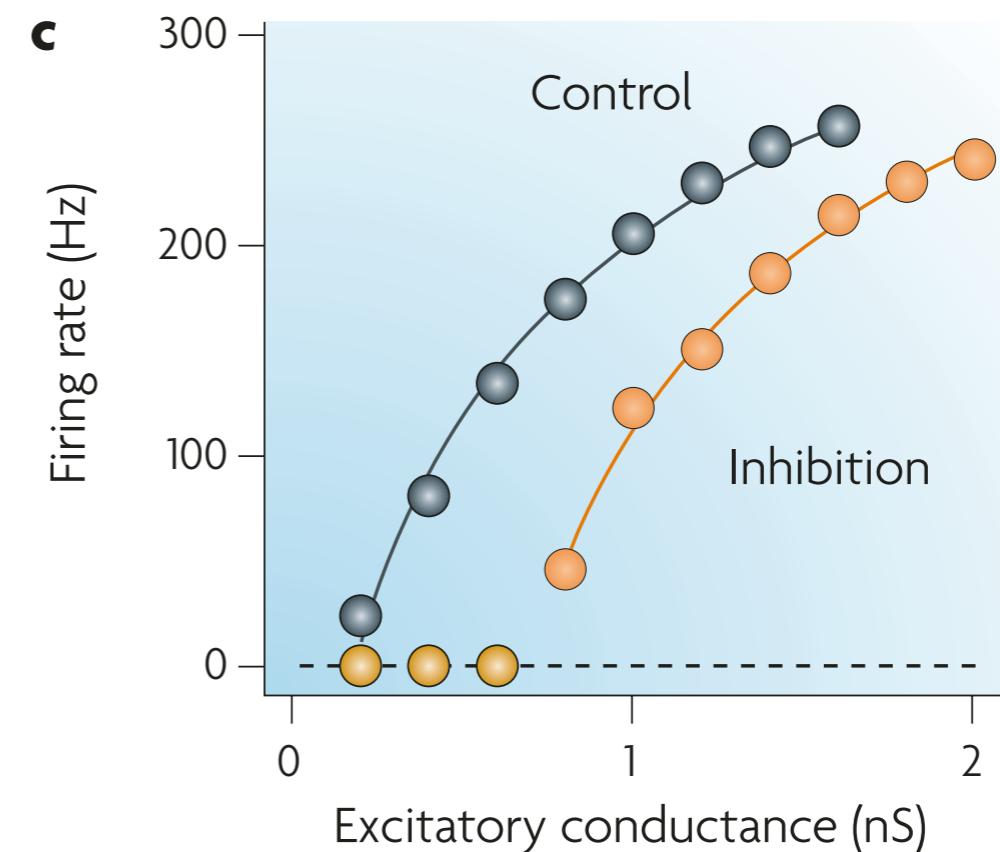
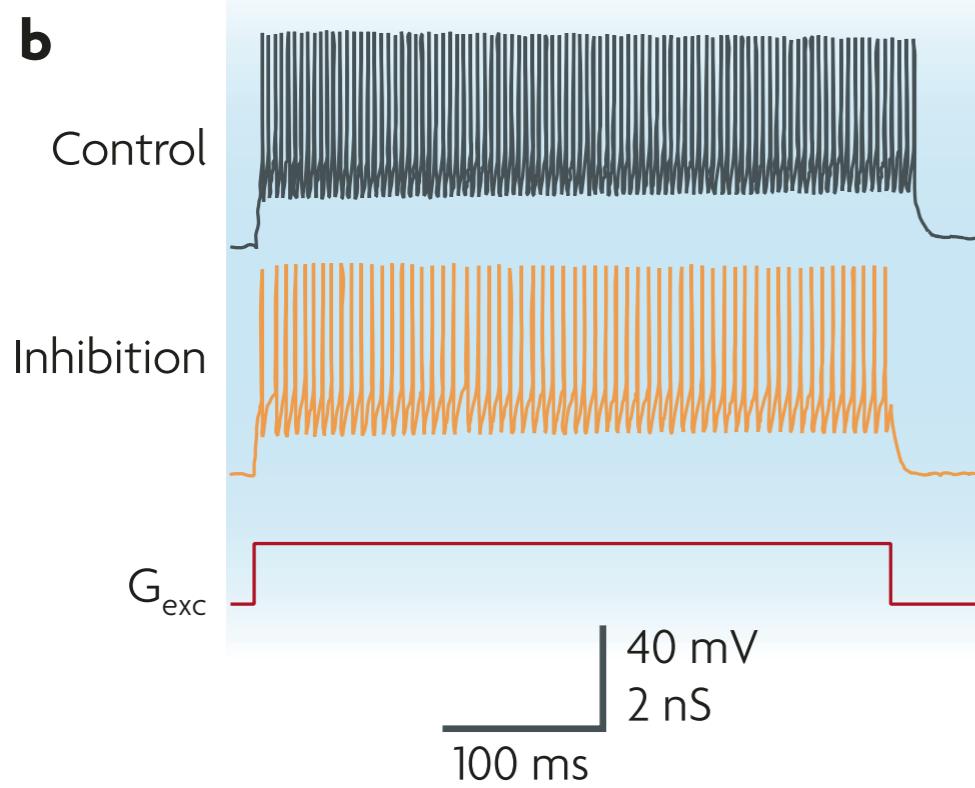
$$C_m \frac{dV}{dt} = -g_L(V - E_L) - g_{inh}(V - E_L) - g_{exc}(V - E_0)$$

$$V = \frac{(g_L + g_{inh})E_L + g_{exc}E_0}{g_L + g_{inh} + g_{exc}}$$

$$g_{inh} \gg g_L + g_{exc} \quad V \approx E_L + \frac{g_{exc}}{g_{inh}} E_0$$



However, shunting inhibition becomes subtractive for firing rate



Why? (Your homework)

But Shunting inhibition can become divisive
in the presence of large synaptic input noise

