

# Biological Modeling of Neural Networks



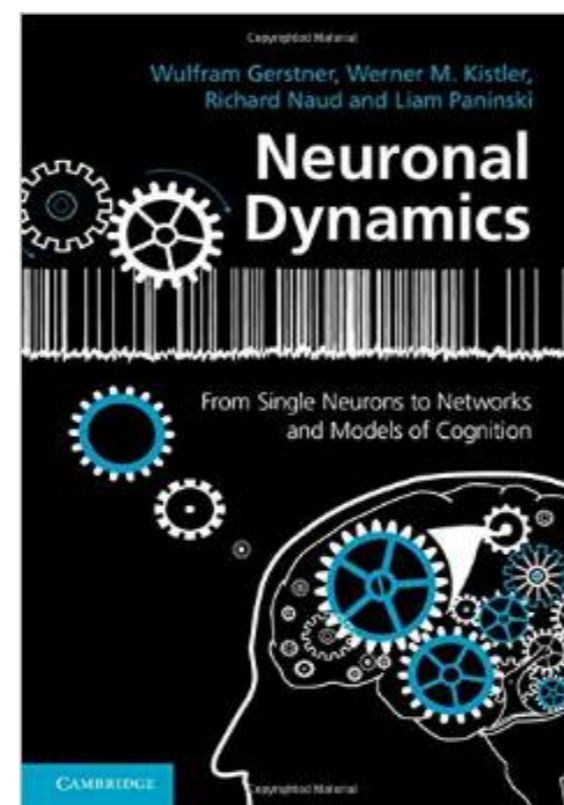
## Week 2 – Biophysical modeling: The Hodgkin-Huxley model

Wulfram Gerstner

EPFL, Lausanne, Switzerland

*Reading for week 2:*  
**NEURONAL DYNAMICS**  
- Ch. 2 (without 2.3.2 - 2.3.5)

Cambridge Univ. Press



## 2.1 Biophysics of neurons

- Overview

## 2.2 Reversal potential

- Nernst equation

## 2.3 Hodgkin-Huxley Model

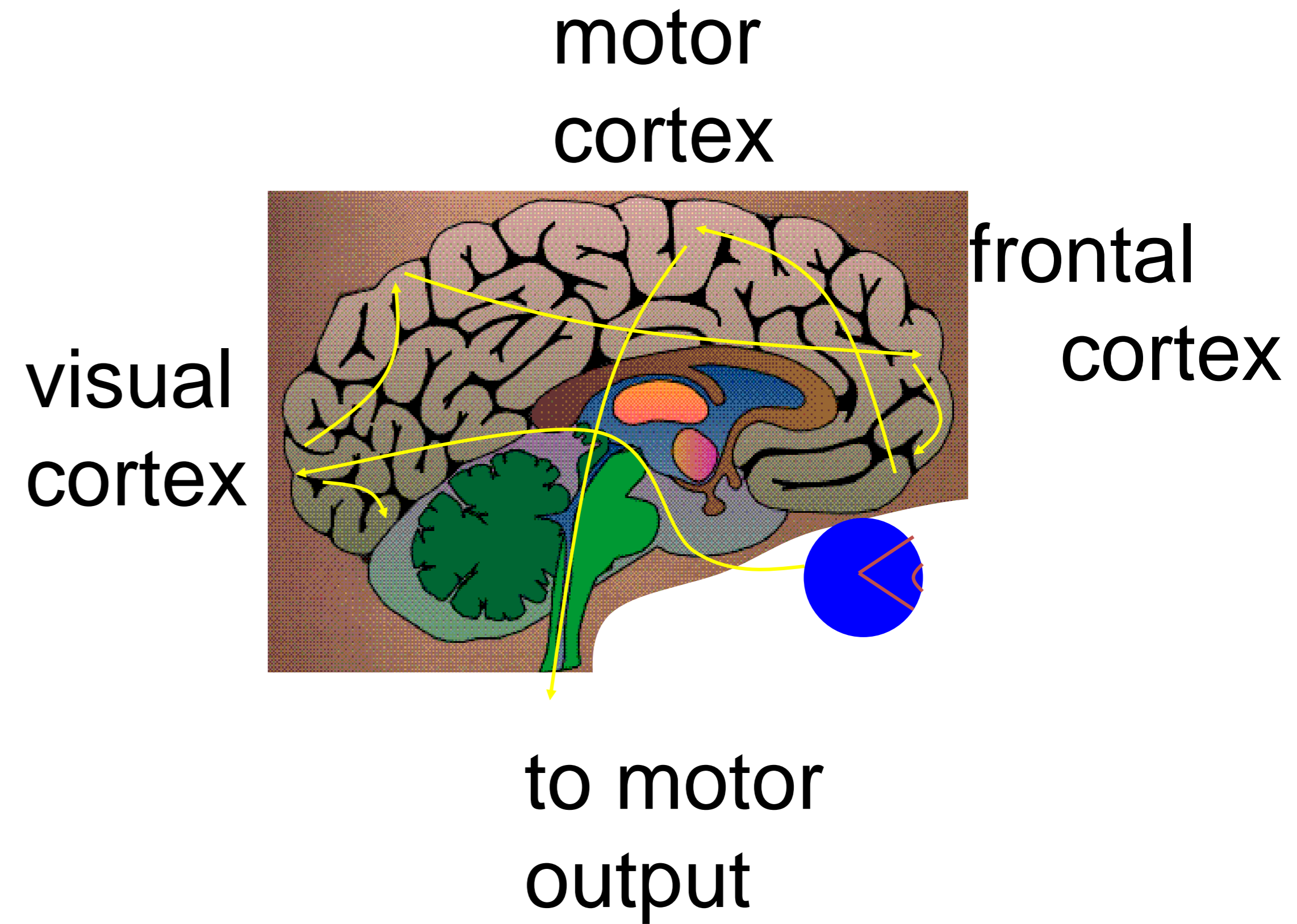
## 2.4 Threshold in the Hodgkin-Huxley Model

- where is the firing threshold?

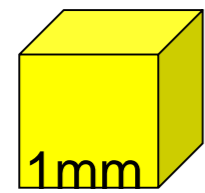
## 2.5. Detailed biophysical models

- the zoo of ion channels

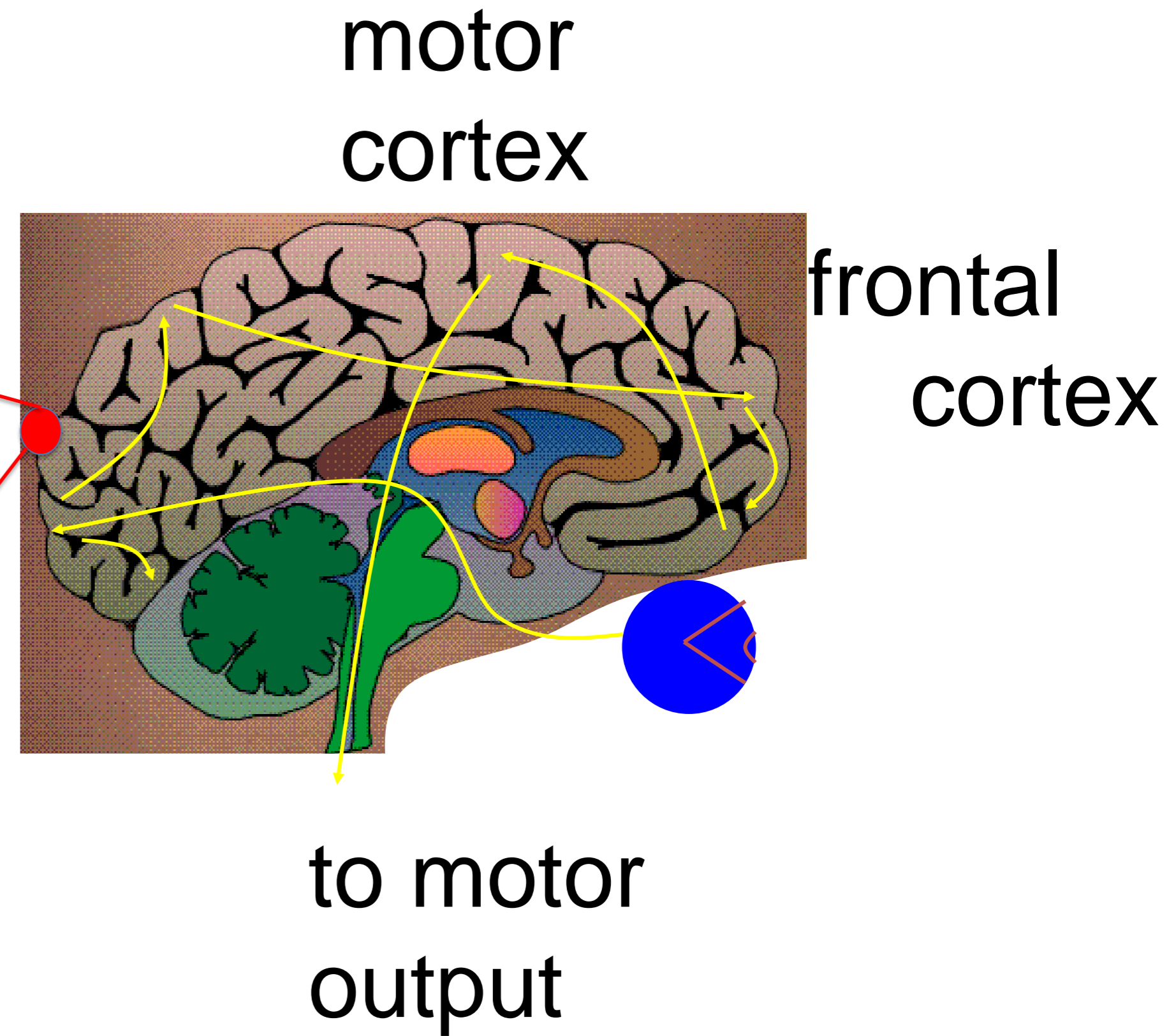
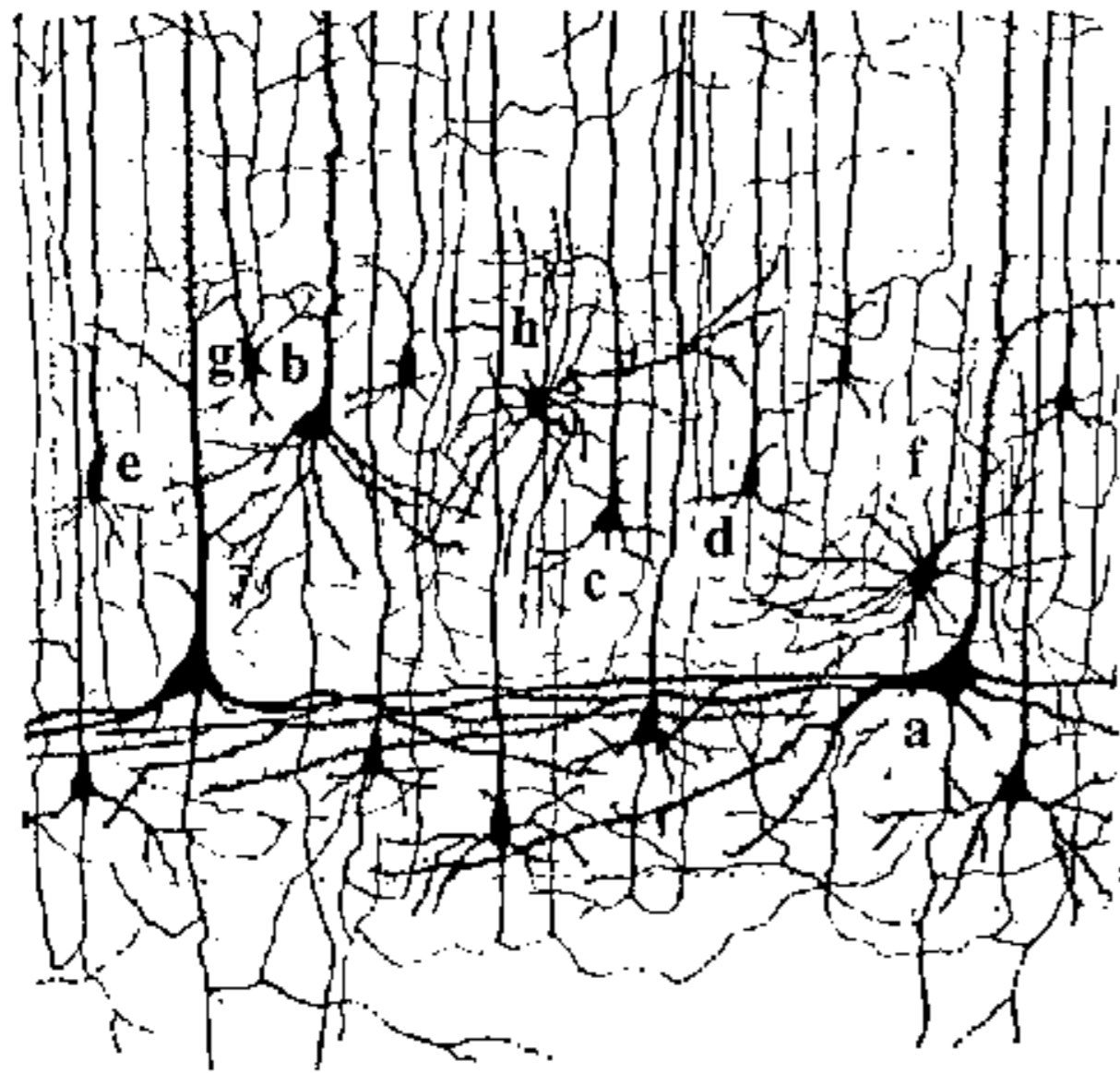
# Review of week 1: Neurons and synapses



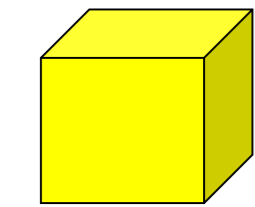
# Review of week 1: Neurons and synapses



10 000 neurons  
3 km of wire

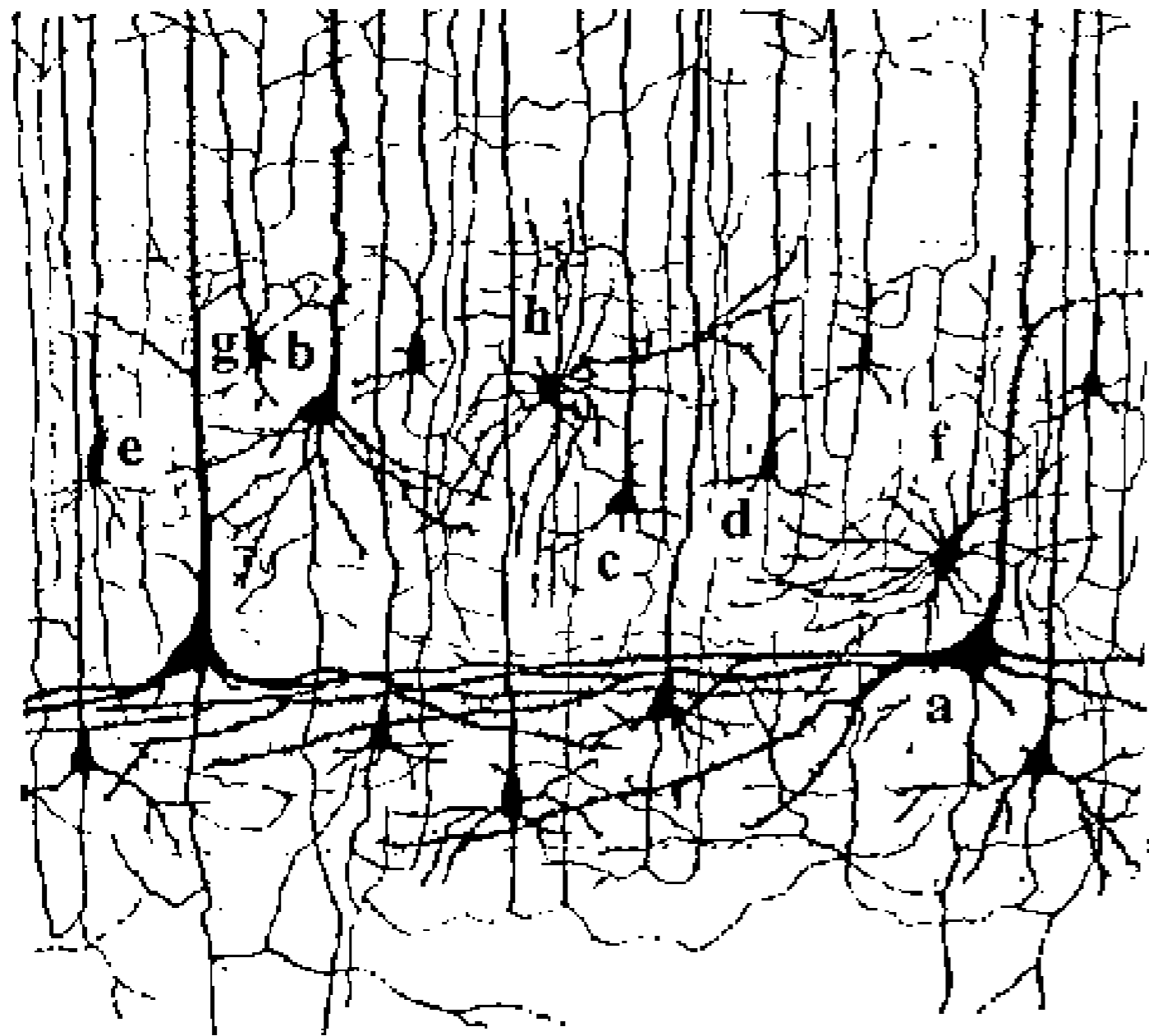


# Review of week 1: Neurons and synapses



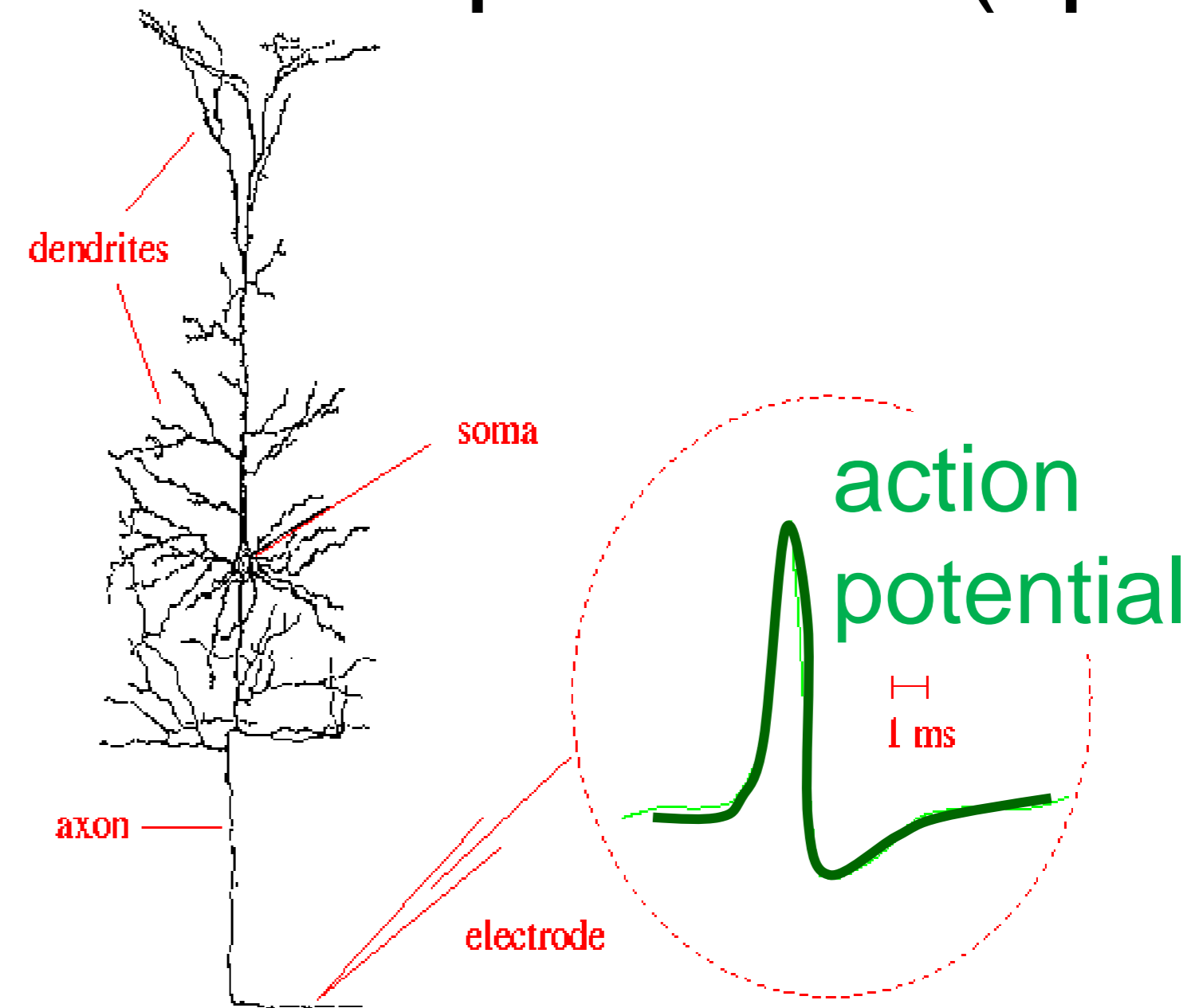
1mm

10 000 neurons  
3 km of wire



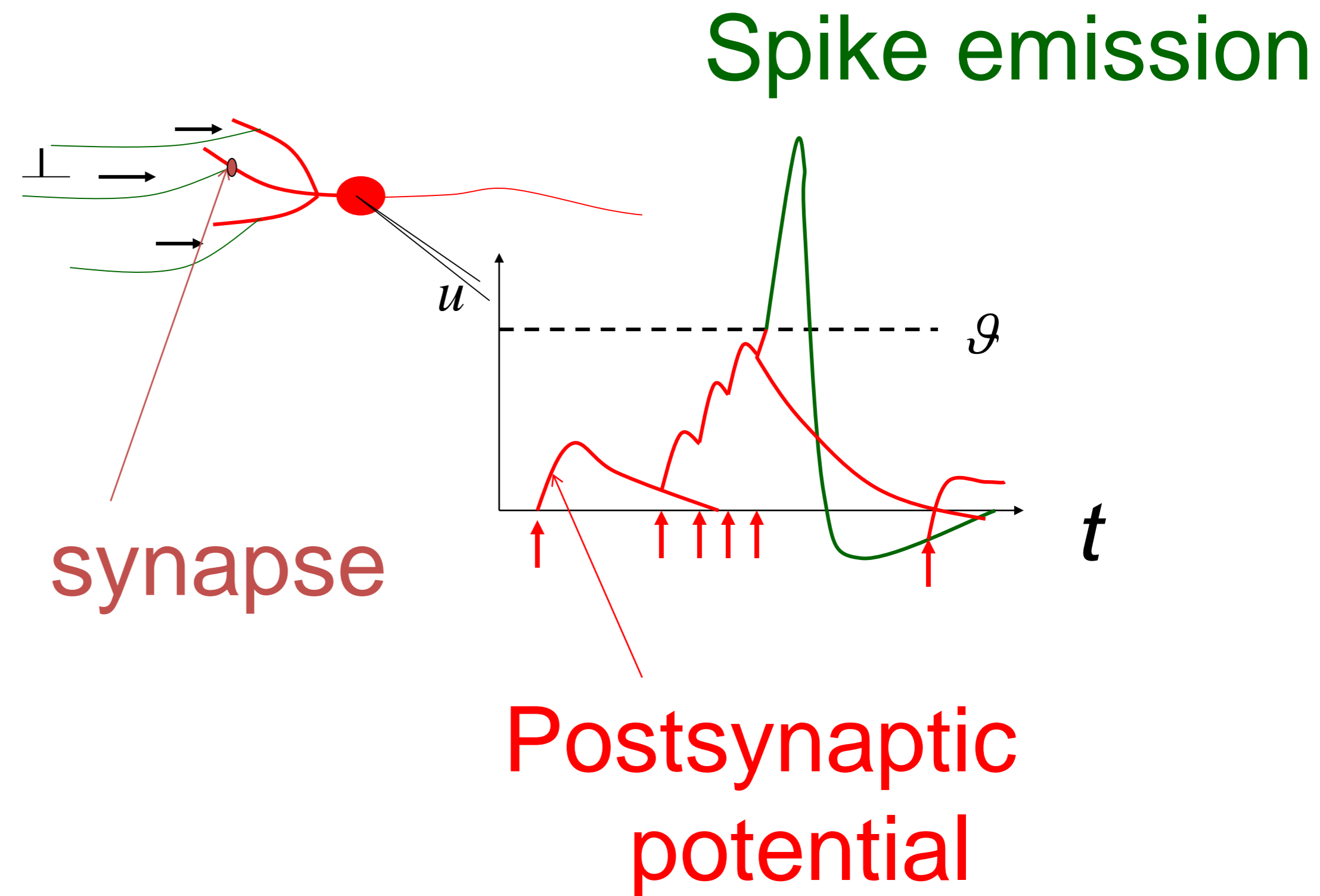
*Ramon y Cajal*

Signal:  
action potential (spike)



How is a spike generated?

# Review of week 1: Integrate-and-Fire models



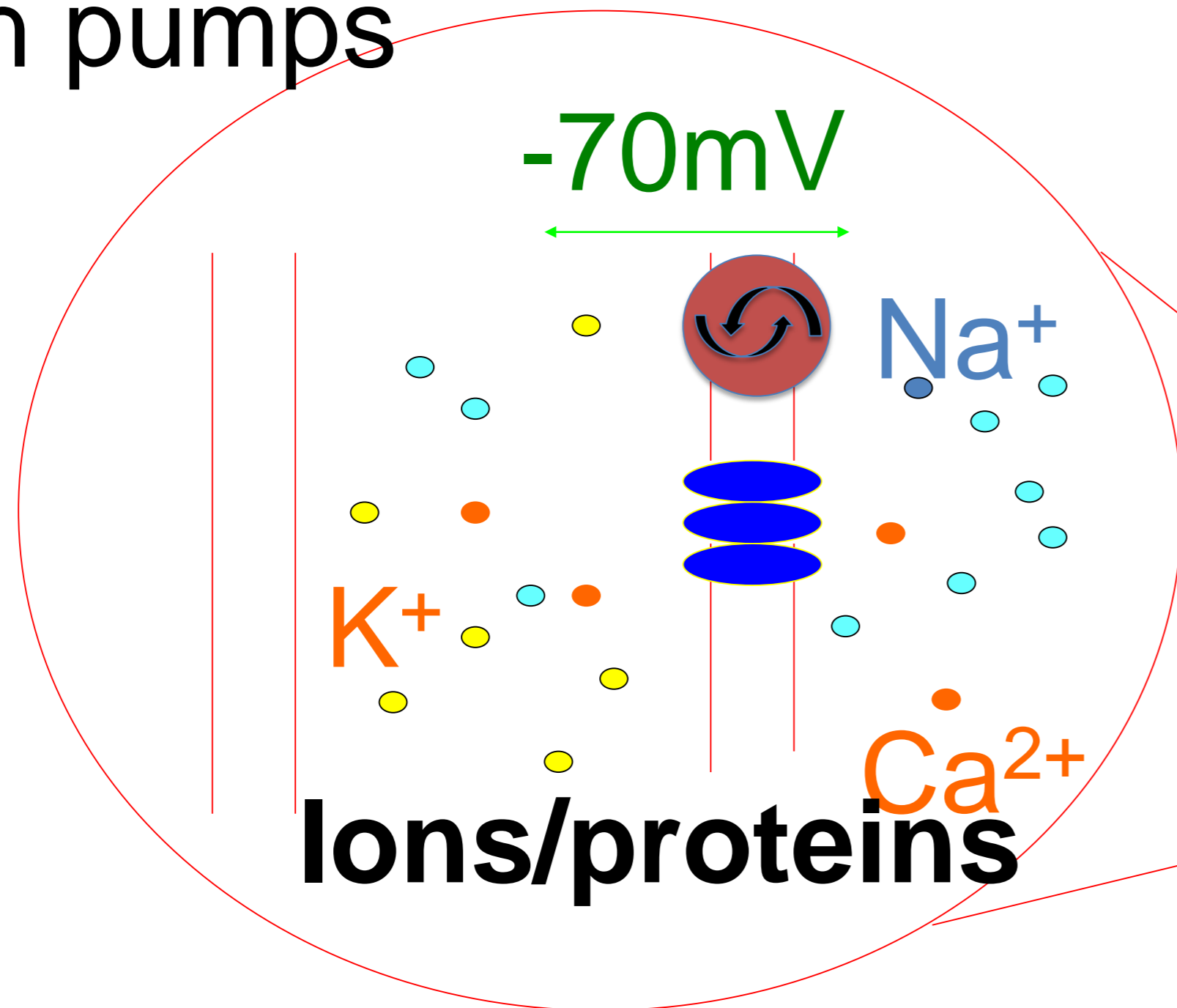
- spikes are events
- triggered at threshold
- spike/reset/refractoriness

# Neuronal Dynamics – week 2: Biophysics of neurons

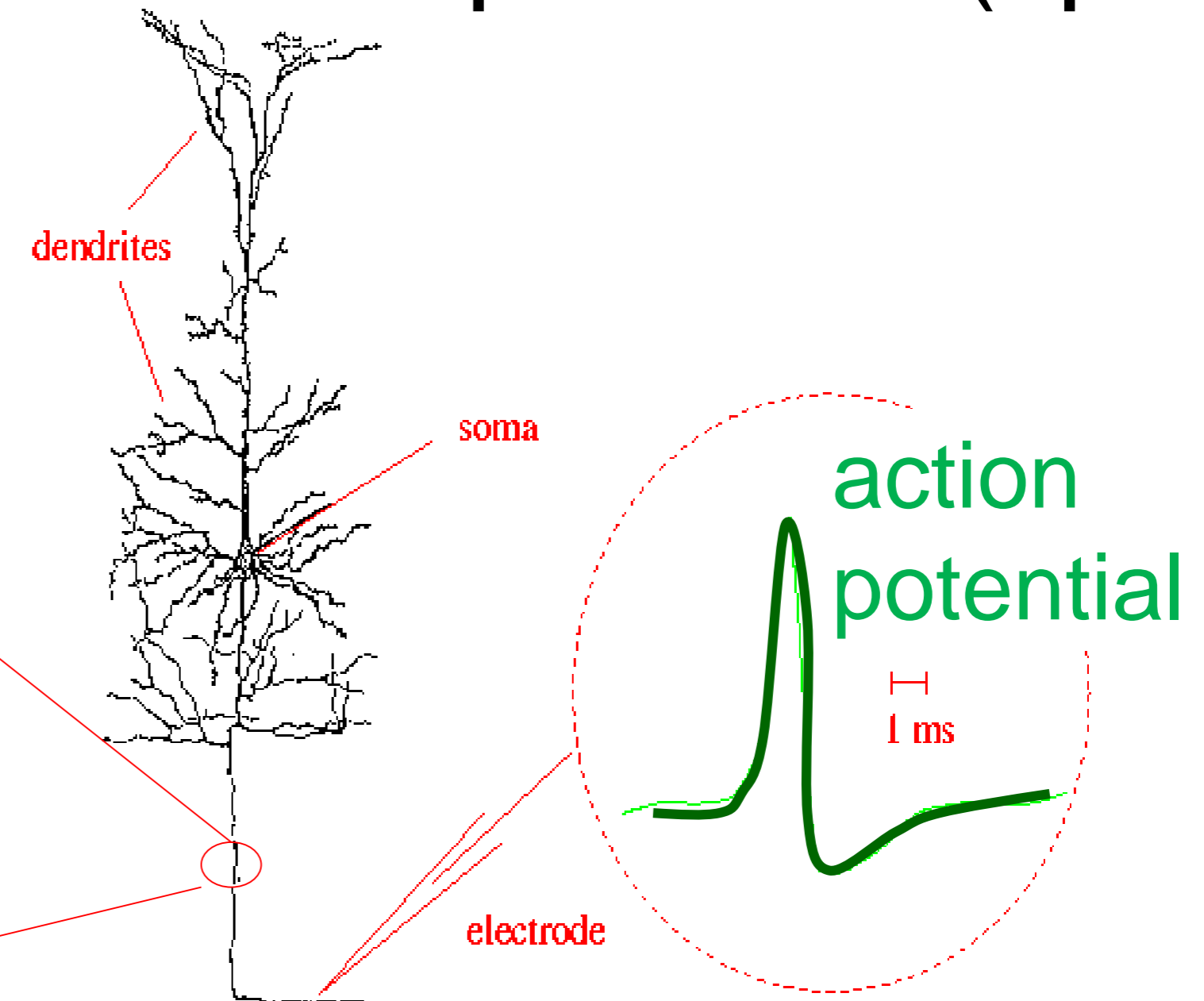
## Cell surrounded by membrane

Membrane contains

- ion channels
- ion pumps



Signal:  
action potential (spike)

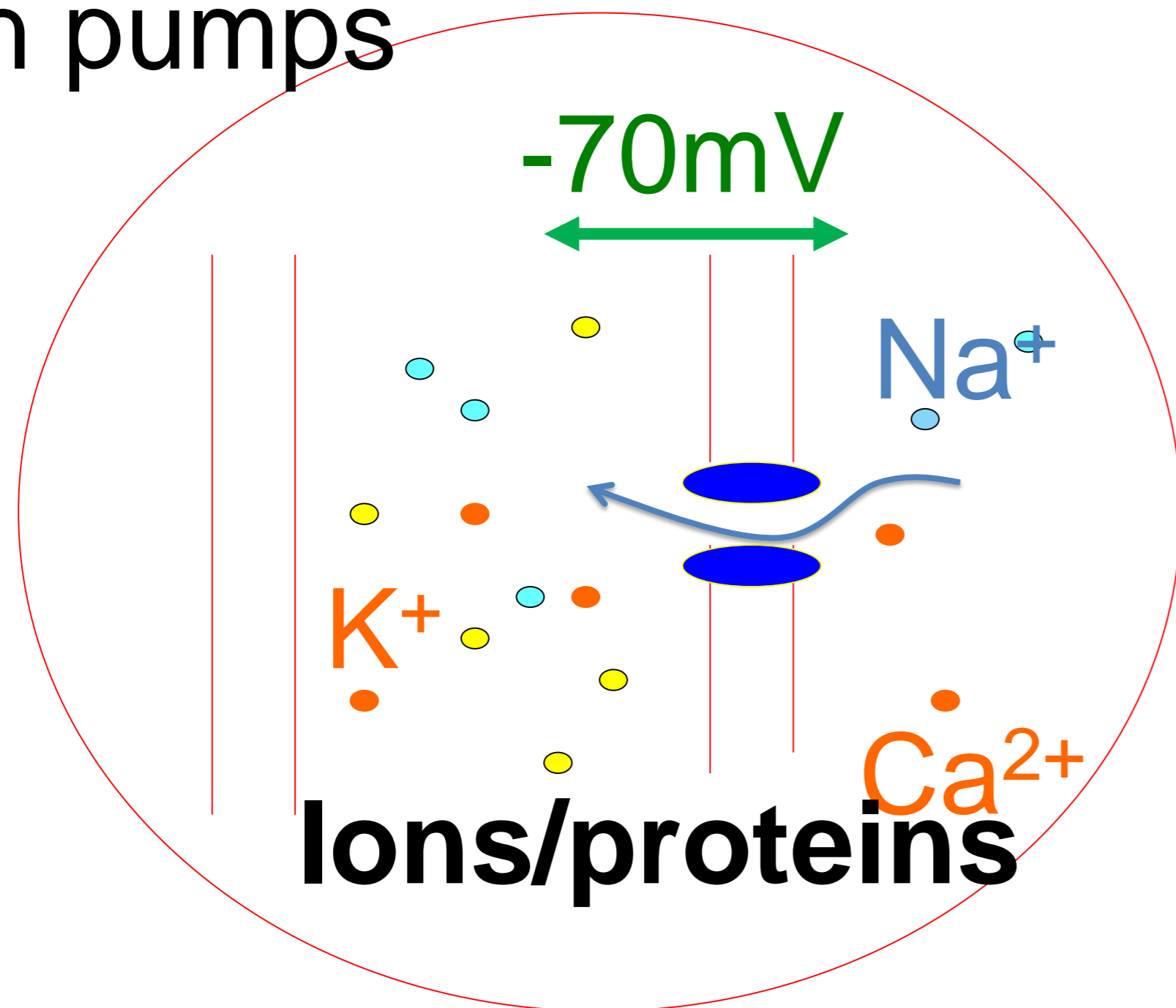


# Neuronal Dynamics – week 2: **Biophysics of neurons**

## Cell surrounded by membrane

Membrane contains

- ion channels
- ion pumps



Resting potential  $-70mV$

→ how does it arise?

Ions flow through channel

→ in which direction?

Neuron emits action potentials

→ why?



# Neuronal Dynamics – 2.1. Biophysics of neurons

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Resting potential  $-70\text{mV}$

→ how does it arise?

Ions flow through channel

→ in which direction?

Neuron emits action potentials

→ why?

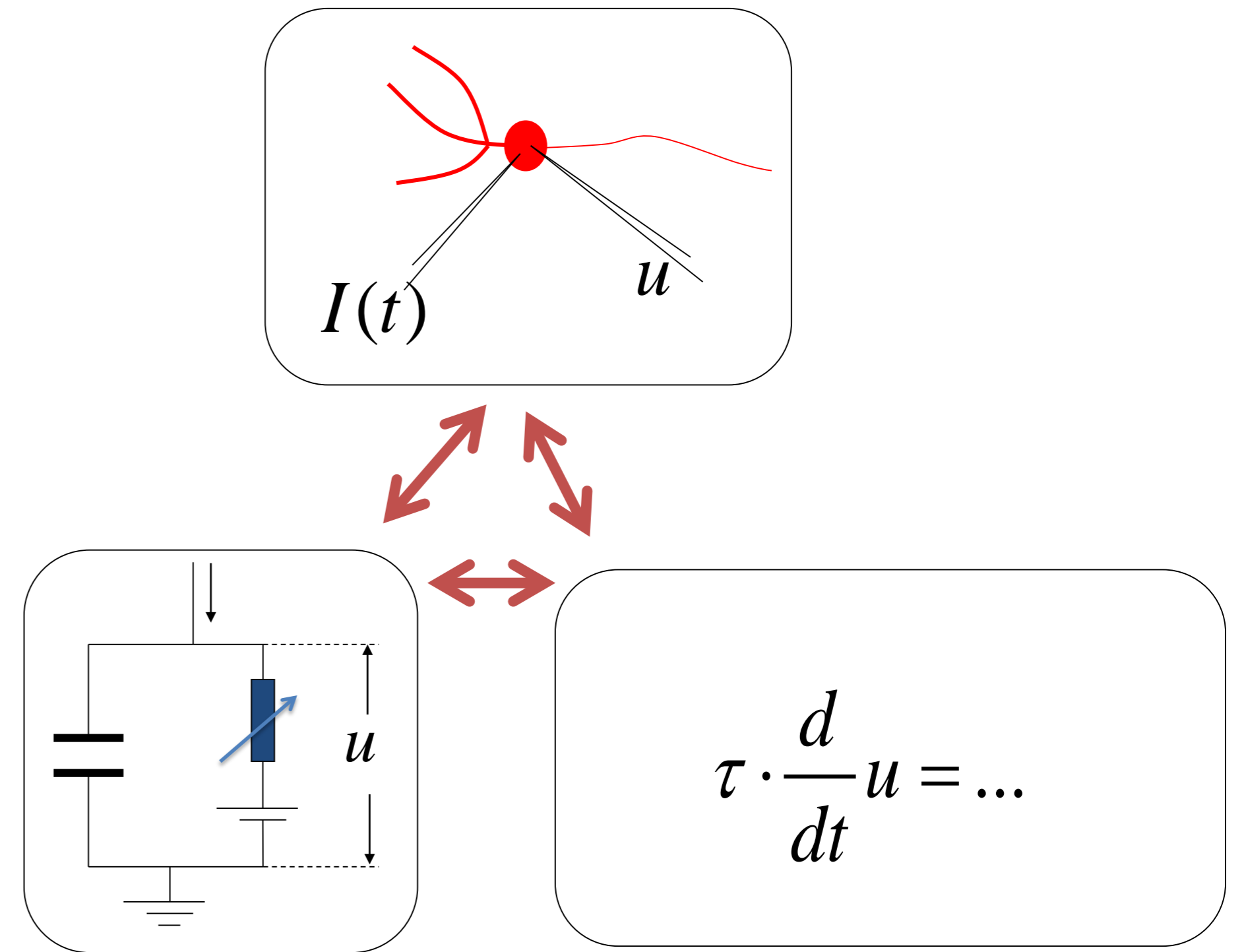
→ Hodgkin-Huxley model

*Hodgkin&Huxley (1952)*

*Nobel Prize 1963*



# Neuronal Dynamics – 2.1. Biophysics of neurons



→ Hodgkin-Huxley model

*Hodgkin&Huxley (1952)*

*Nobel Prize 1963*

## Week 2 – Quiz

In a natural situation, the electrical potential inside a neuron is

- the same as outside
- is different by 50-100 microvolt
- is different by 50-100 millivolt

Neurons and cells

- Neurons are special cells because they are surrounded by a membrane
- Neurons are just like other cells surrounded by a membrane
- Neurons are not cells

Ion channels are

- located in the cell membrane
- special proteins
- can switch from open to closed

If a channel is open, ions can

- flow from the surround into the cell
- flow from inside the cell into the surrounding liquid

*Multiple answers possible!*

# Week 2 – part 2: Reversal potential and Nernst equation



## Biological Modeling of Neural Networks

### Week 2 – Biophysical modeling: The Hodgkin-Huxley model

Wulfram Gerstner

EPFL, Lausanne, Switzerland

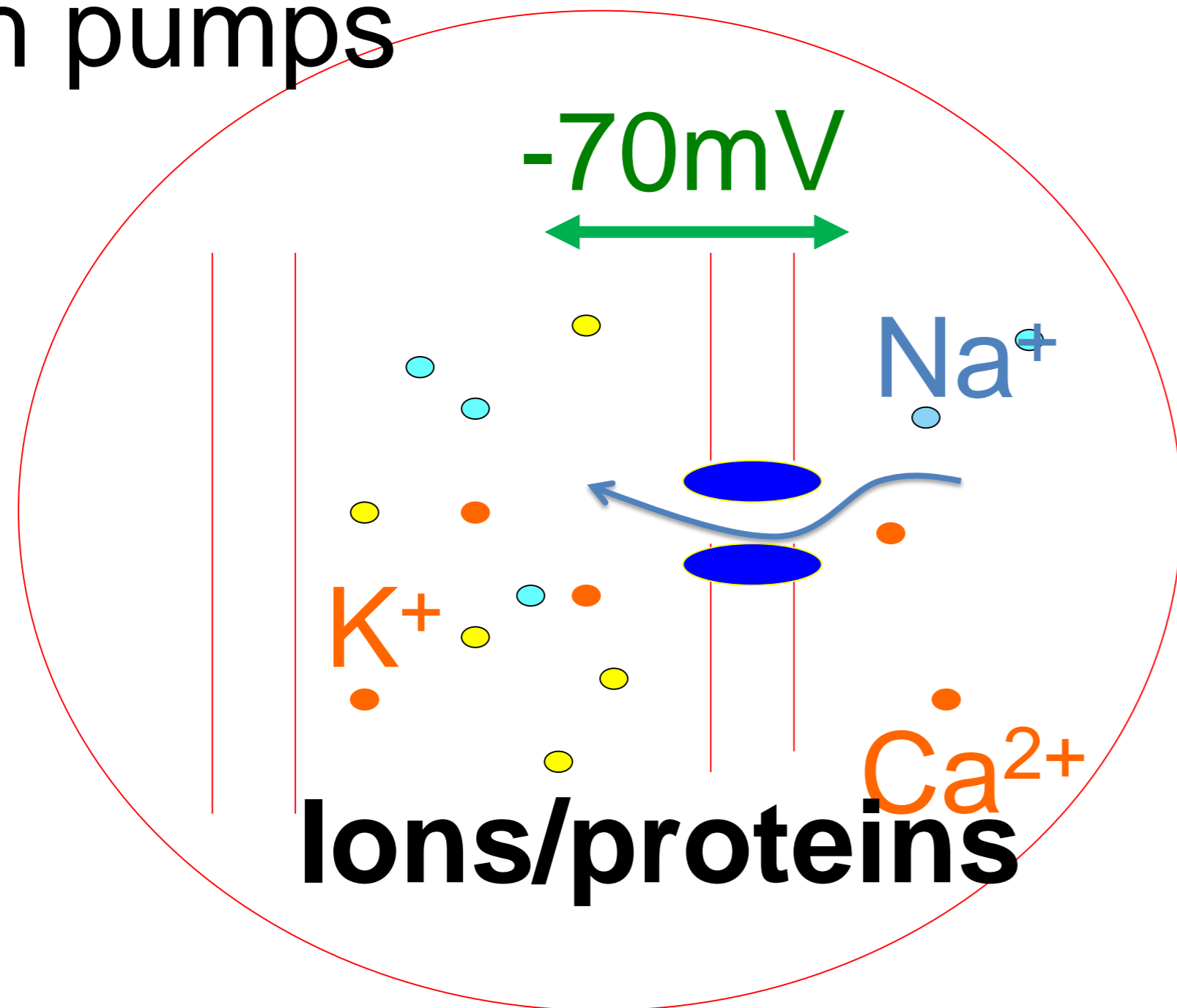
- ✓ 2.1 Biophysics of neurons
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  - where is the firing threshold?
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# Neuronal Dynamics – 2.2. Resting potential

## Cell surrounded by membrane

Membrane contains

- ion channels
- ion pumps



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Neuron emits action potentials

→ why?

# Neuronal Dynamics – 2.2. Resting potential

Resting potential  $-70\text{mV}$   
→ how does it arise?

Ions flow through channel  
→ in which direction?

Neuron emits action potentials  
→ why?

→ Hodgkin-Huxley model

*Hodgkin&Huxley (1952)*

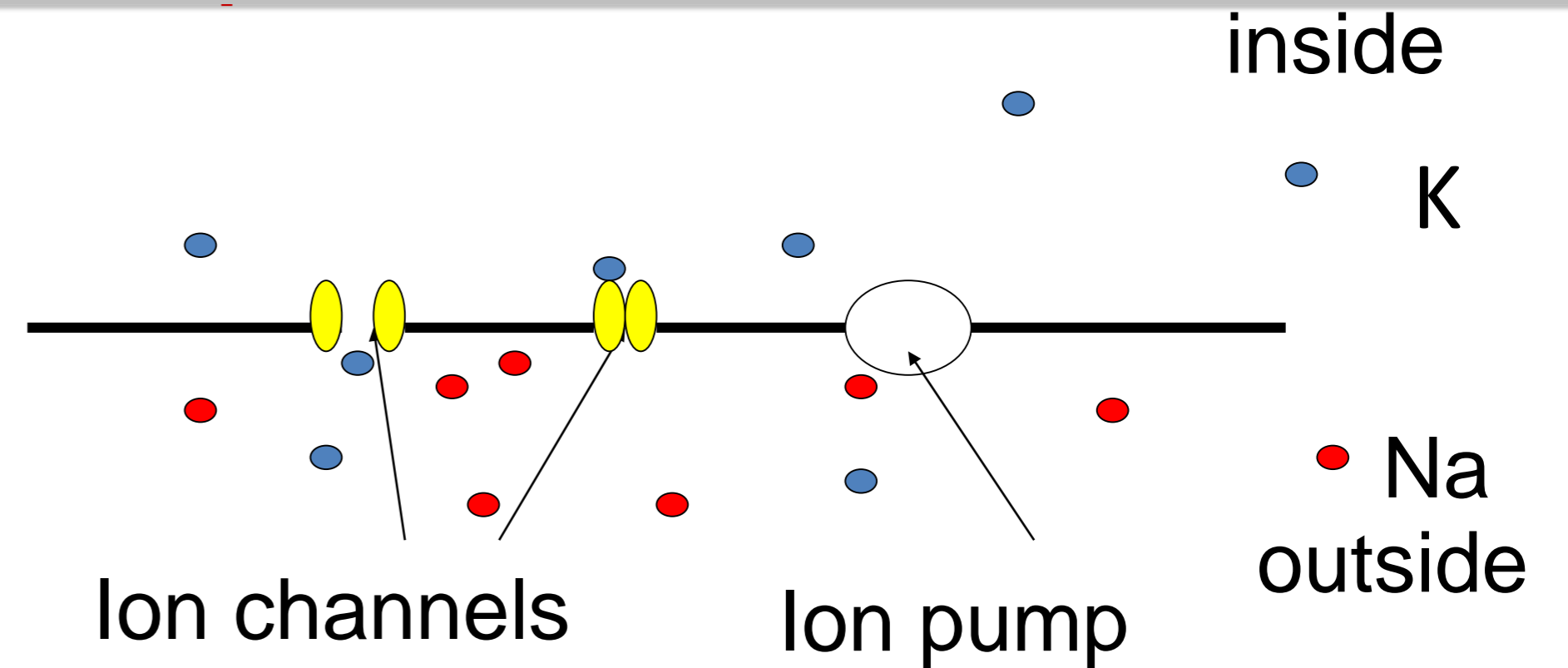
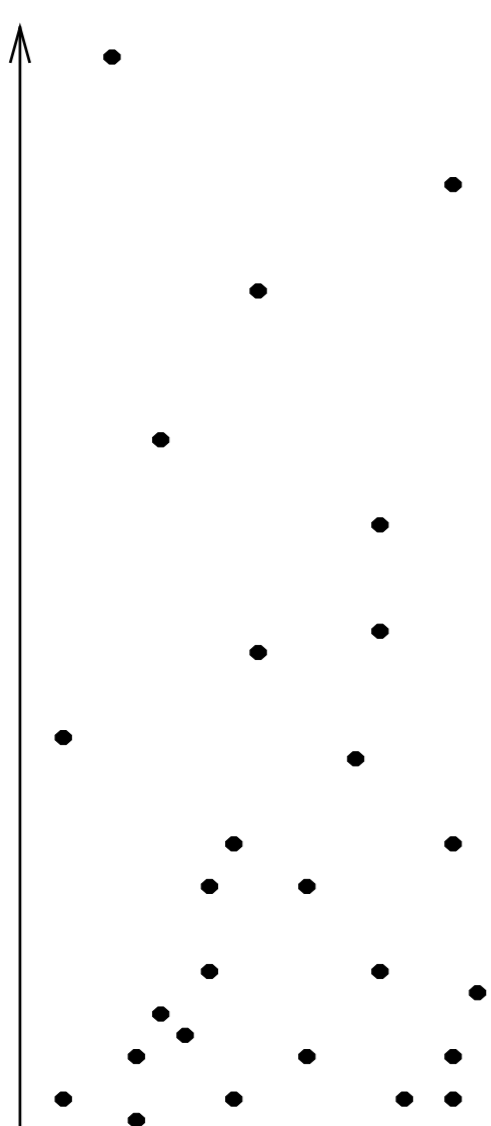
*Nobel Prize 1963*

# Neuronal Dynamics – 2.2. Reversal potential

density

$$n \propto e^{-\frac{E}{kT}}$$

E

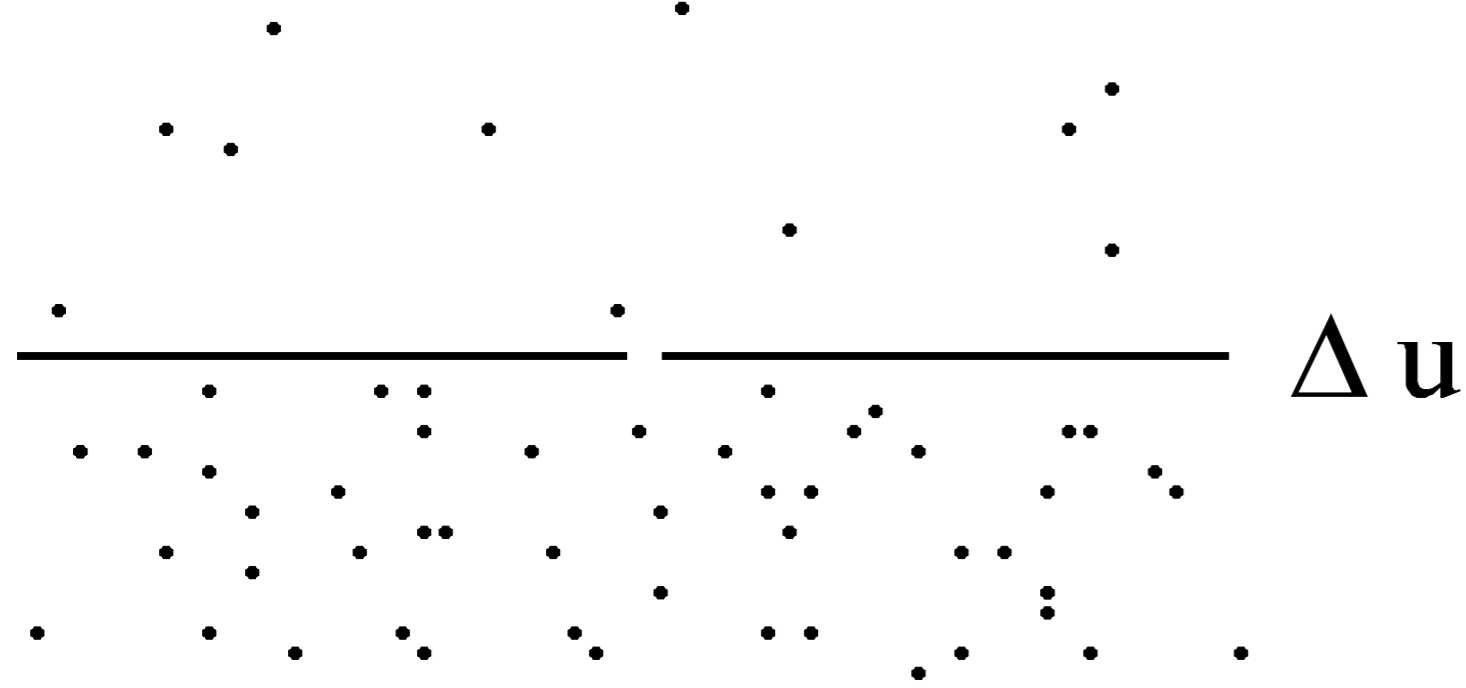


**Ion pump  $\Leftrightarrow$  Concentration difference**

*Mathetical derivation*

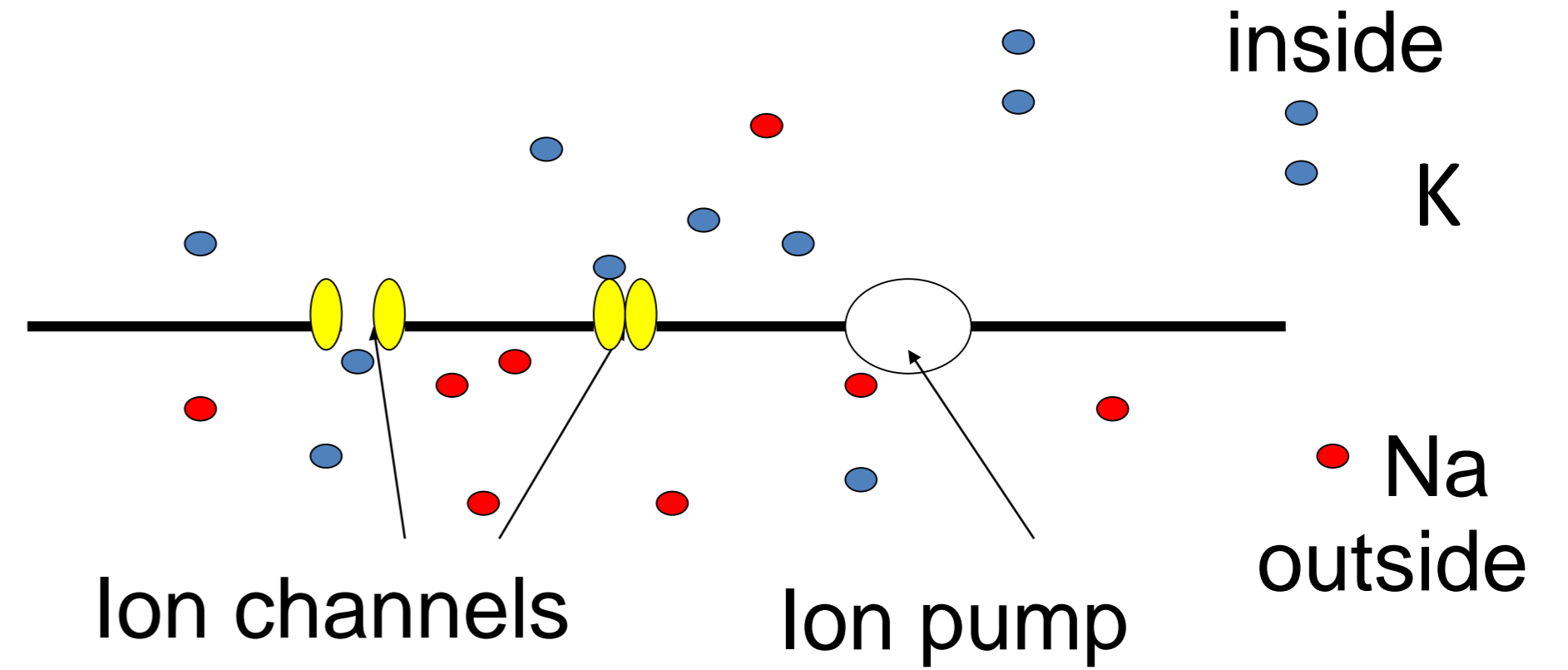
# Neuronal Dynamics – 2.2. Nernst equation

$n_1$  (inside)



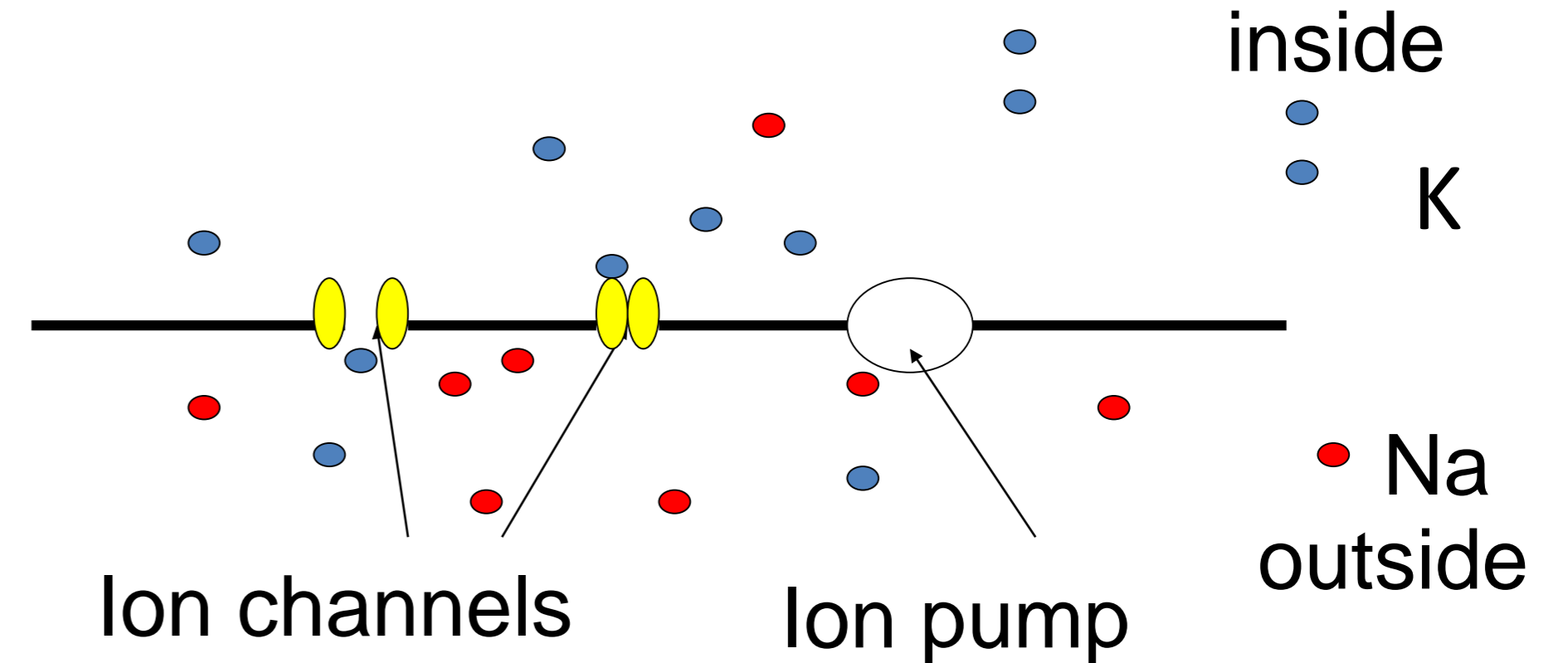
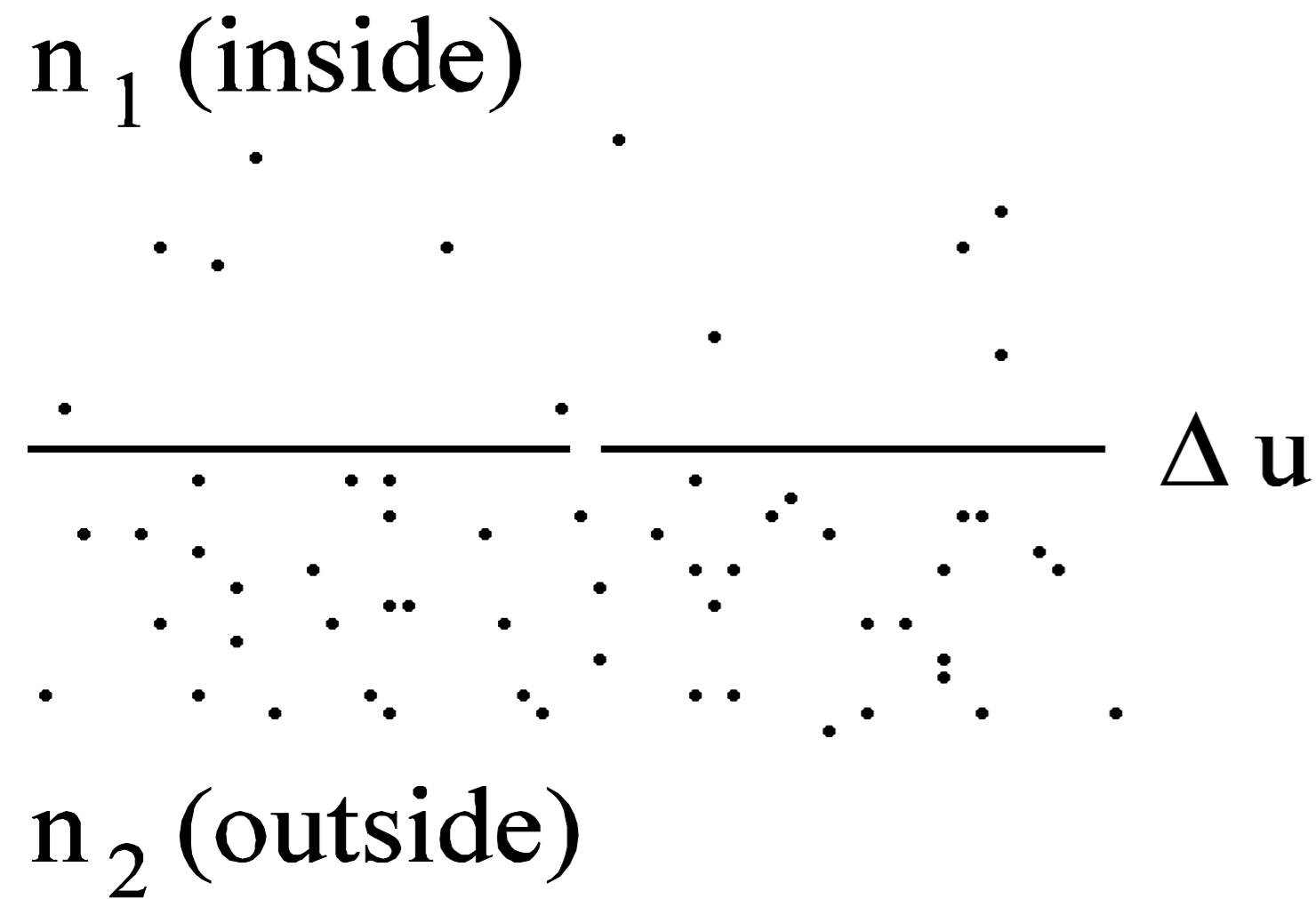
$n_2$  (outside)

$$n \propto e^{-\frac{E}{kT}}$$





# Neuronal Dynamics – 2.2. Nernst equation



$$\Delta u = u_1 - u_2 = \frac{-kT}{q} \ln \frac{n(u_1)}{n(u_2)}$$

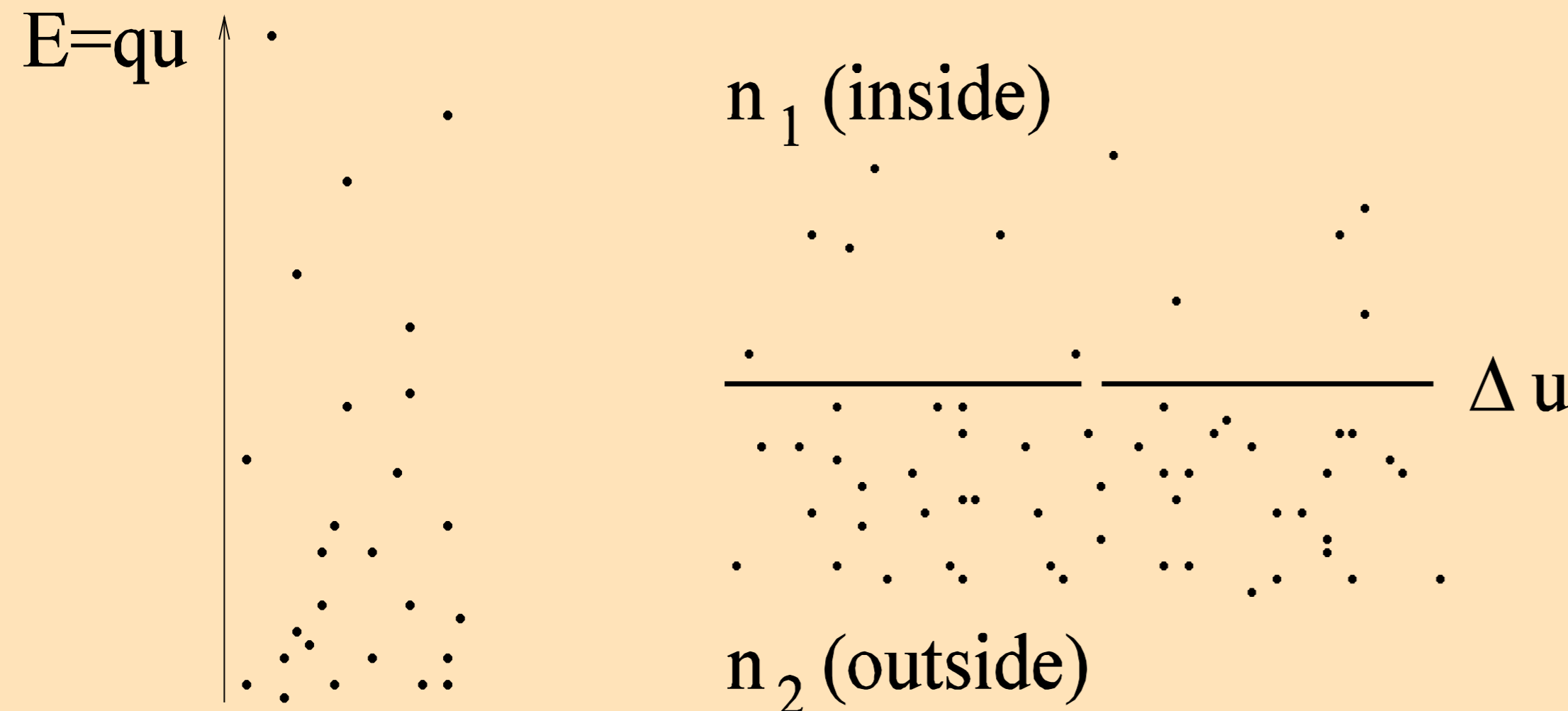
Reversal potential

**Concentration difference  $\Leftrightarrow$  voltage difference**

# Exercise 1.1– Reversal potential of ion channels

## Reversal potential

$$\Delta u = u_1 - u_2 = -\frac{kT}{q} \ln \frac{n(u_1)}{n(u_2)}$$



Calculate the reversal potential for Sodium

Potassium

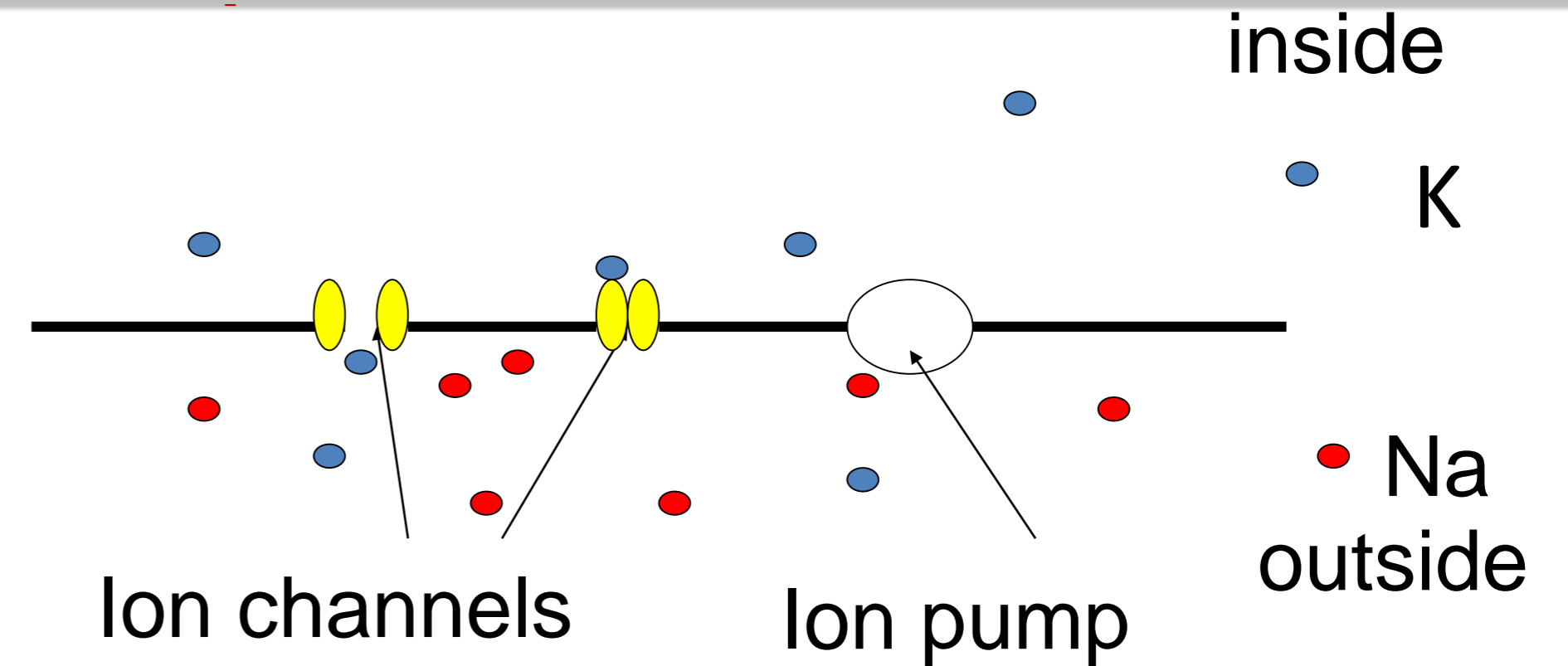
Calcium

given the concentrations

What happens if you change the temperature  $T$  from 37 to 18.5 degree?

***Start exercise at 9:35  
Next Lecture 9:48***

# Neuronal Dynamics – 2.2. Reversal potential



**Ion pump → Concentration difference**

Concentration difference ⇔ voltage difference

Reversal potential

Nernst equation

# Week 2 – part 3 : Hodgkin-Huxley Model



## Neuronal Dynamics: Computational Neuroscience of Single Neurons

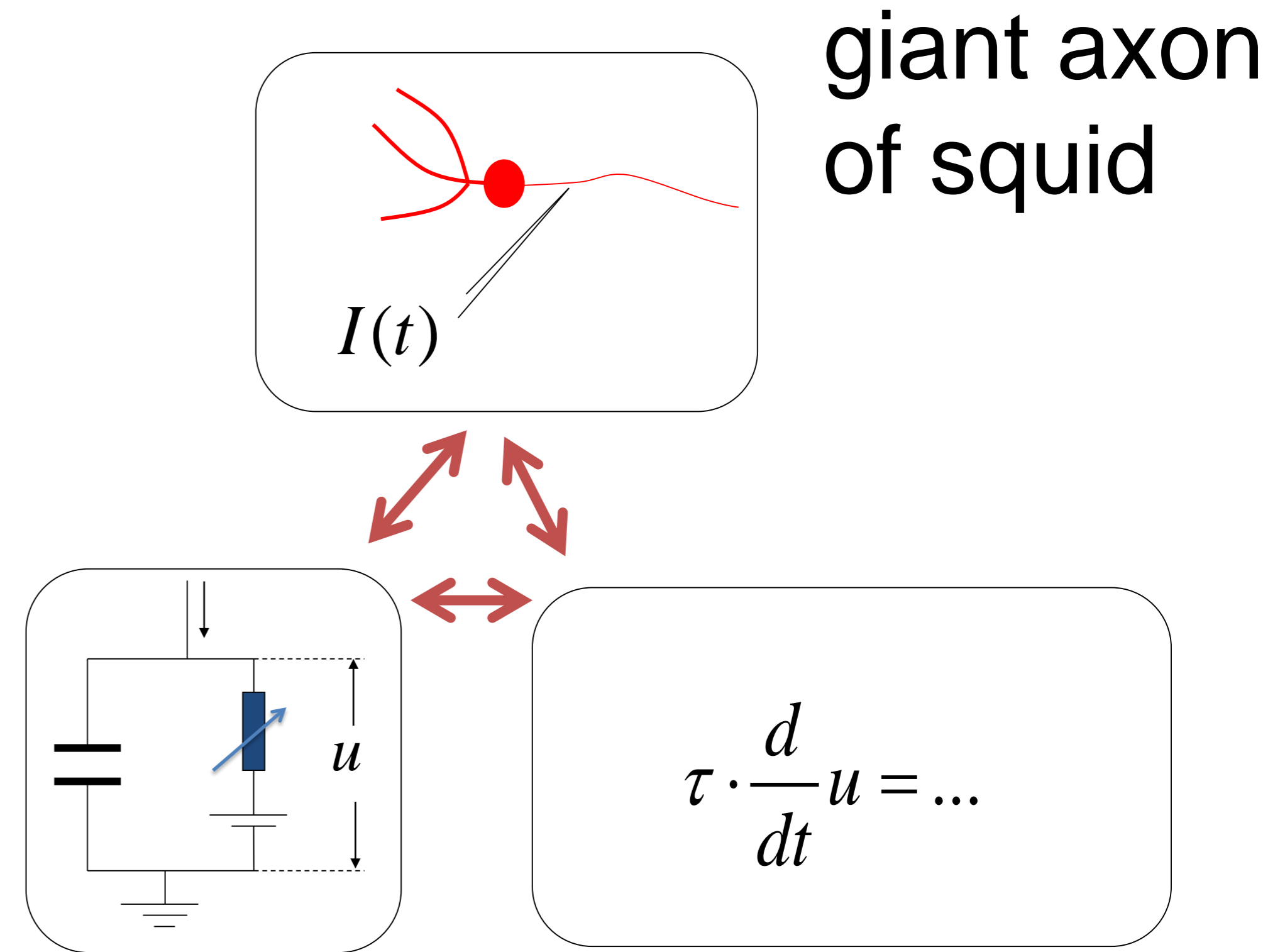
### Week 2 – Biophysical modeling: The Hodgkin-Huxley model

Wulfram Gerstner

EPFL, Lausanne, Switzerland

- ✓ 2.1 Biophysics of neurons
  - Overview
- ✓ 2.2 Reversal potential
  - Nernst equation
- 2.3 Hodgkin-Huxley Model**
- 2.4 Threshold in the Hodgkin-Huxley Model
  - where is the firing threshold?
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  - the zoo of ion channels

# Neuronal Dynamics – 2.3. Hodgkin-Huxley Model



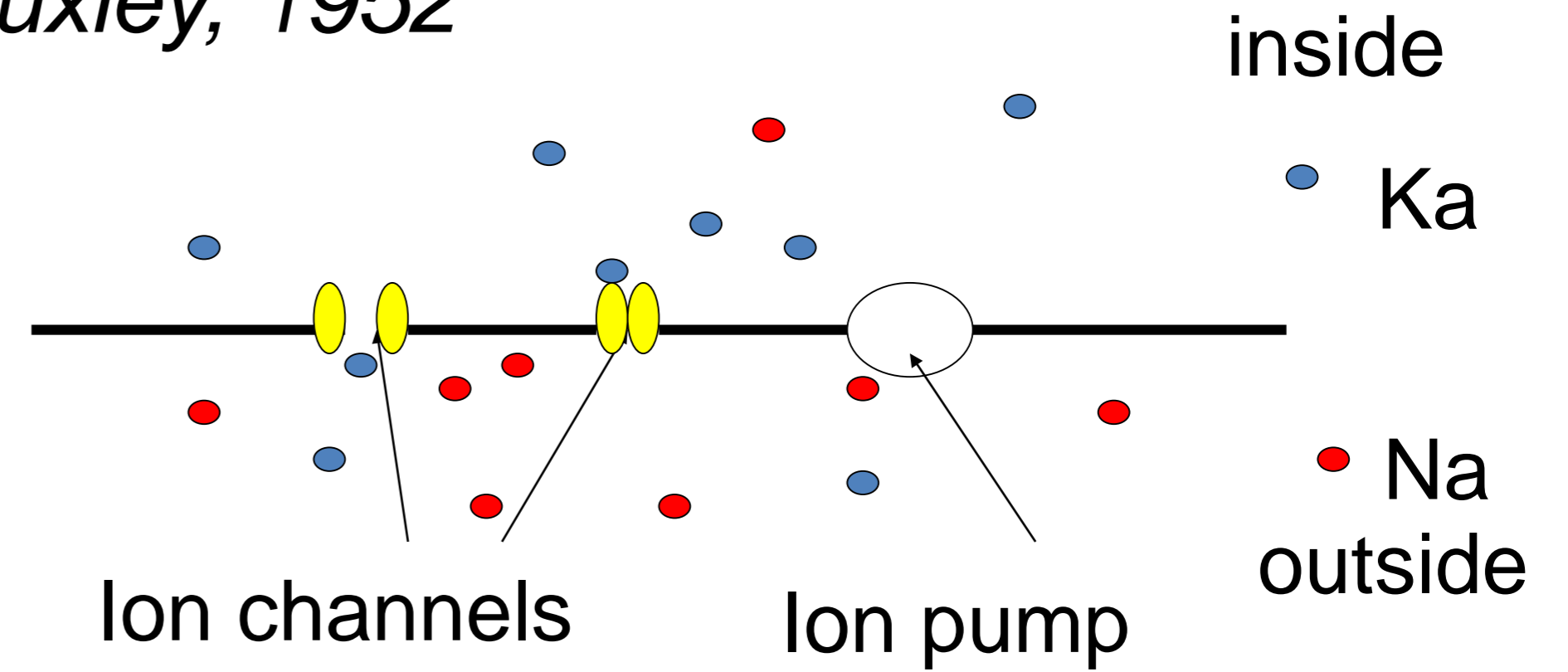
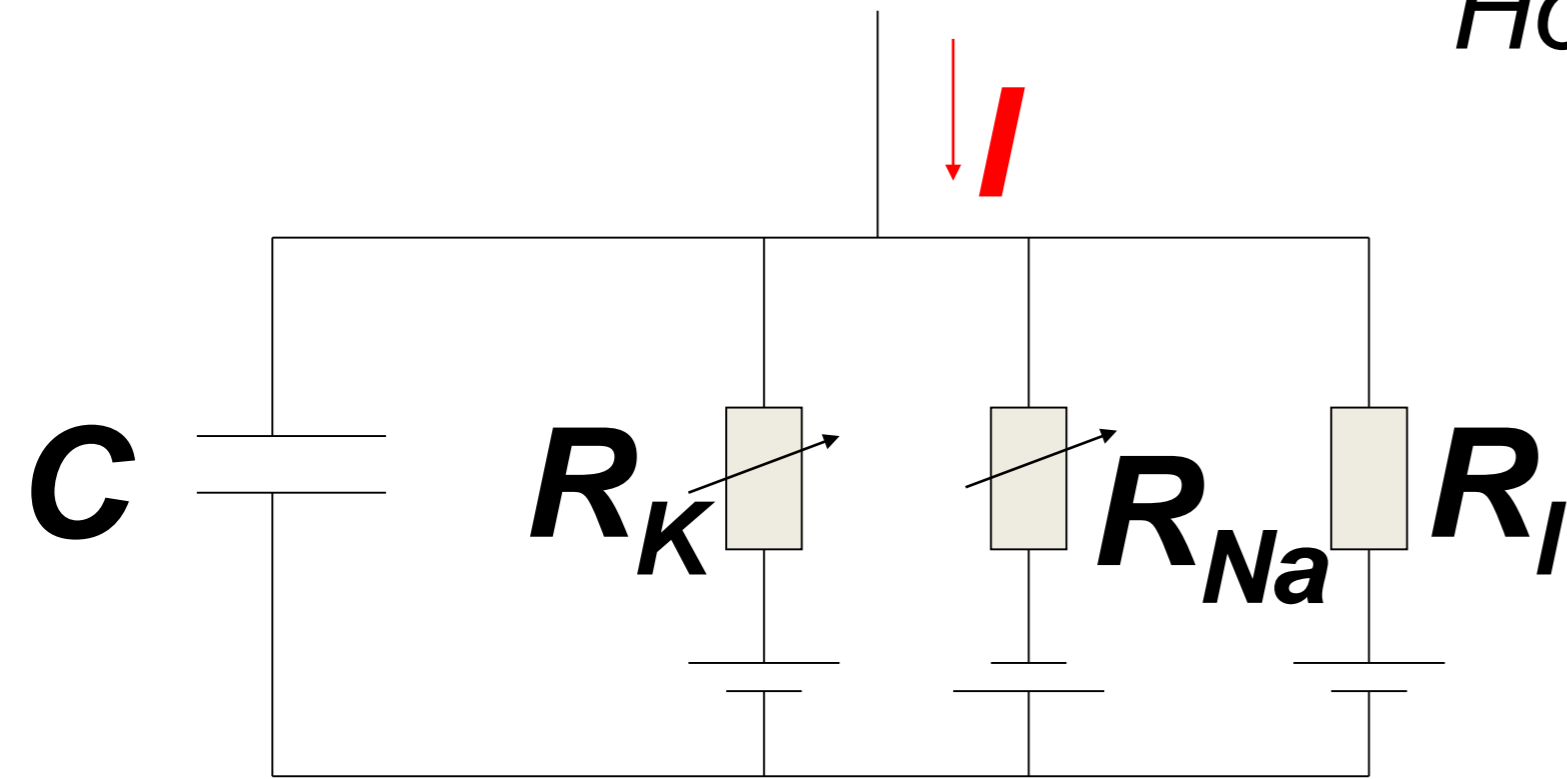
→ Hodgkin-Huxley model

*Hodgkin&Huxley (1952)*

*Nobel Prize 1963*

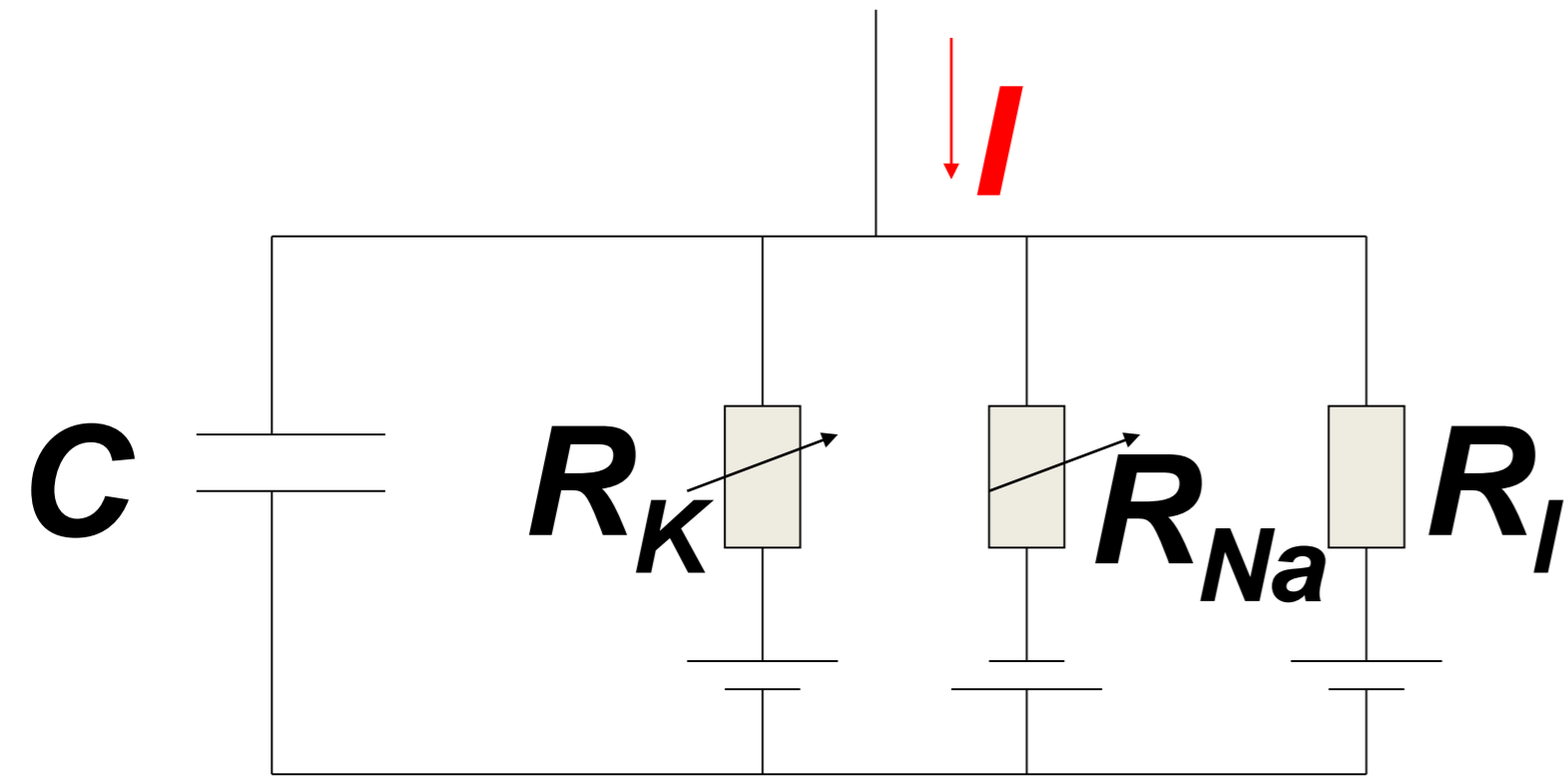
# Neuronal Dynamics – 2.3. Hodgkin-Huxley Model

*Hodgkin and Huxley, 1952*



*Mathematical derivation*

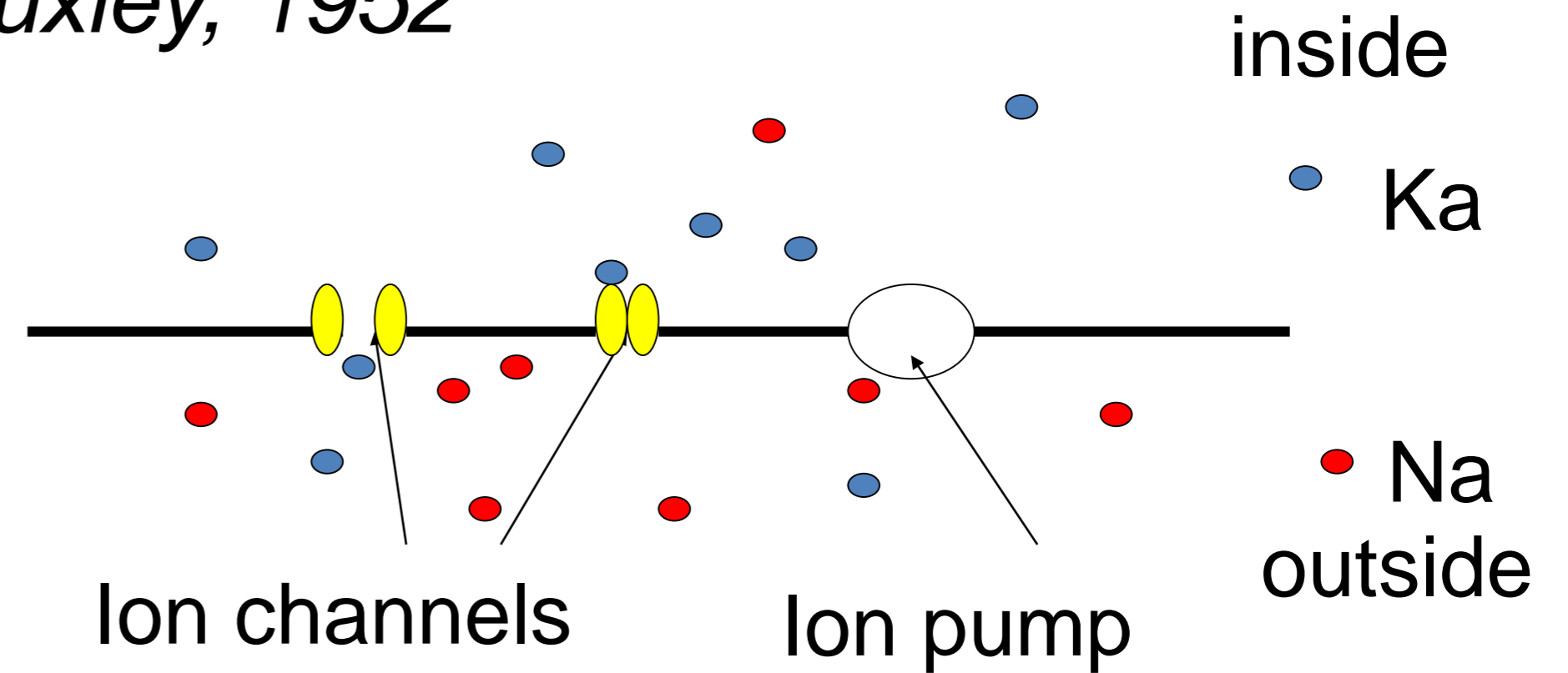
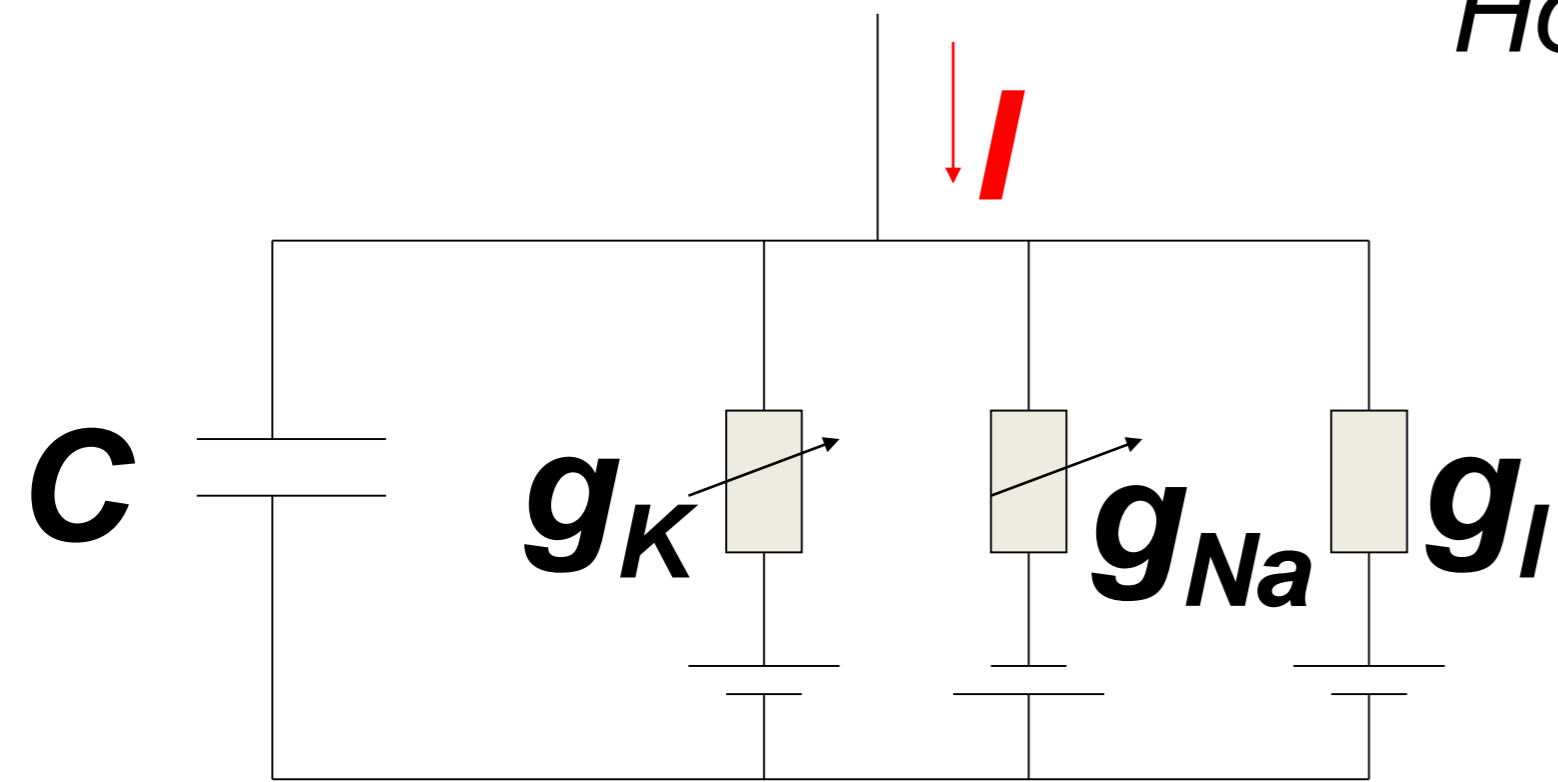
# Neuronal Dynamics – 2.3. Hodgkin-Huxley Model





# Neuronal Dynamics – 2.3. Hodgkin-Huxley Model

Hodgkin and Huxley, 1952



$$C \frac{du}{dt} = \underbrace{-g_{Na} m^3 h (u - E_{Na})}_{I_{Na}} - \underbrace{g_K n^4 (u - E_K)}_{I_K} - \underbrace{g_l (u - E_l)}_{I_{leak}} + I(t)$$

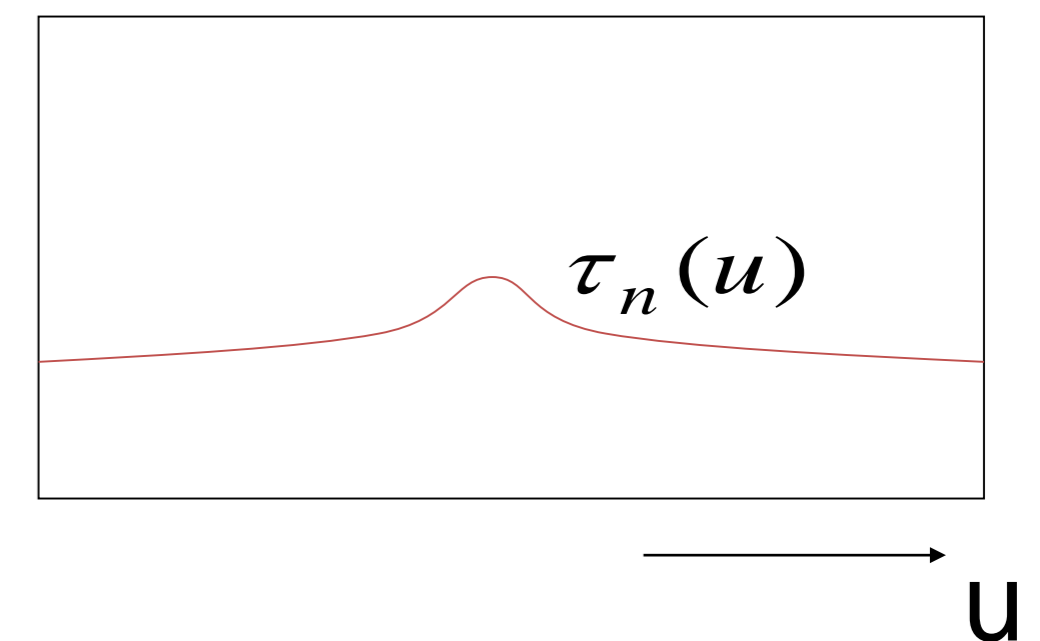
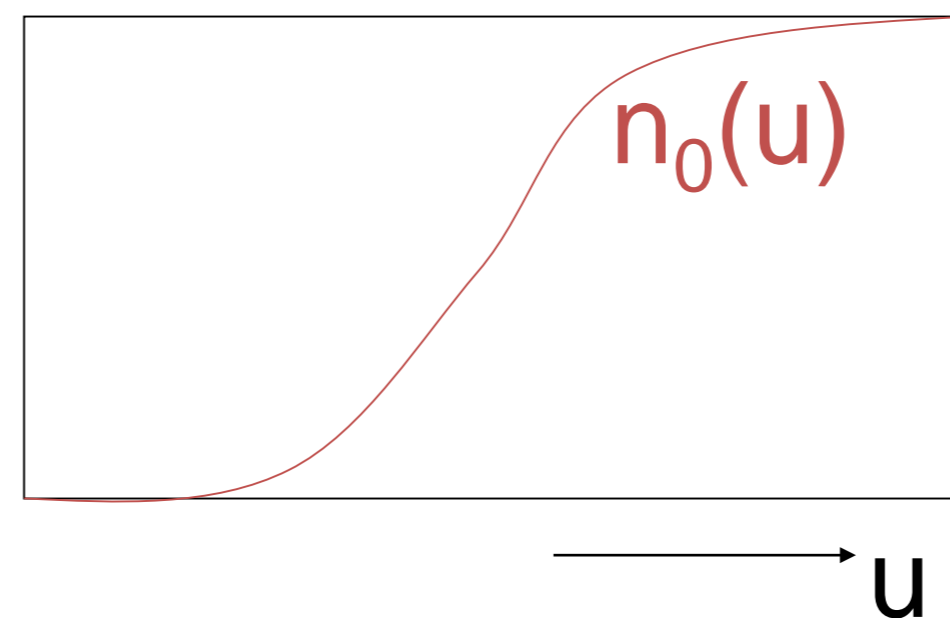
stimulus ↓

$$\frac{dh}{dt} = \frac{h - h_0(u)}{\tau_h(u)}$$

$$\frac{dm}{dt} = \frac{m - m_0(u)}{\tau_m(u)}$$

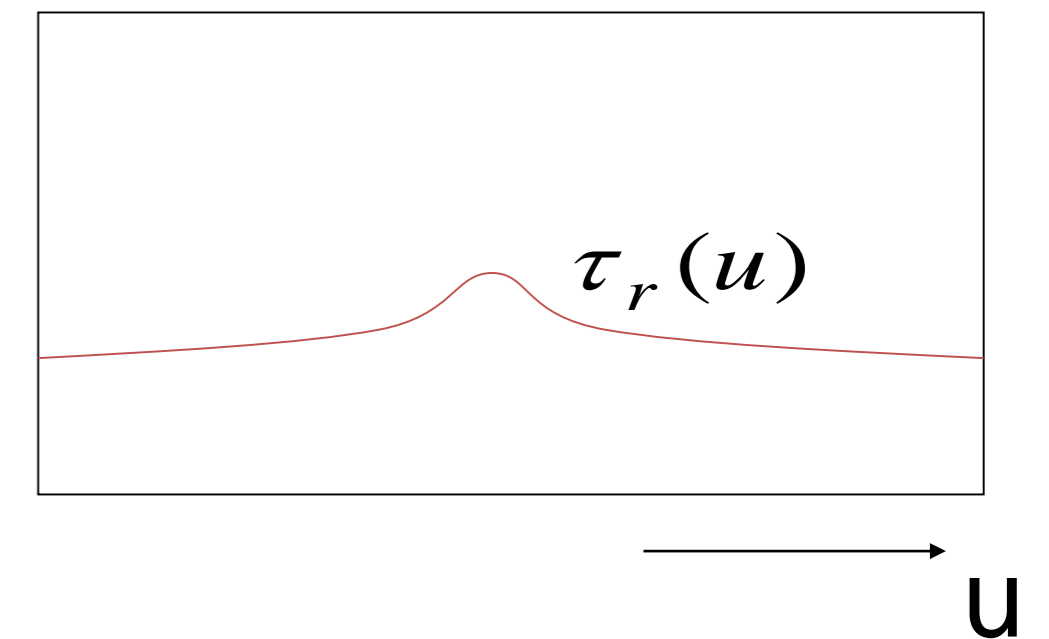
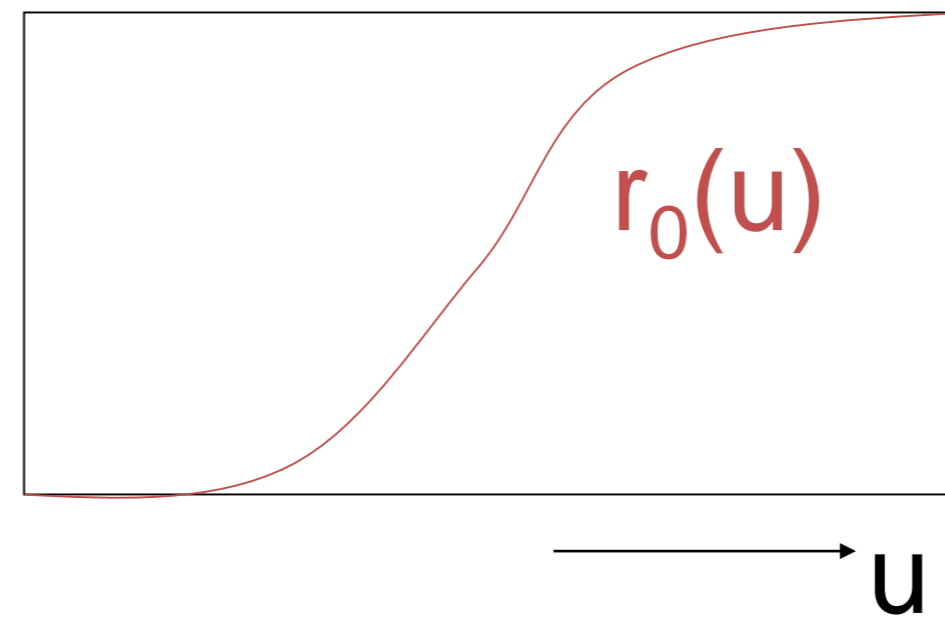
$$\frac{dm}{dt} = \frac{m - m_0(u)}{\tau_m(u)}$$

$$\frac{dm}{dt} = \frac{m - m_0(u)}{\tau_m(u)}$$



# Neuronal Dynamics – 2.3. Ion channel

$$C \frac{du}{dt} = - \sum_k I_{ion,k} + I(t)$$

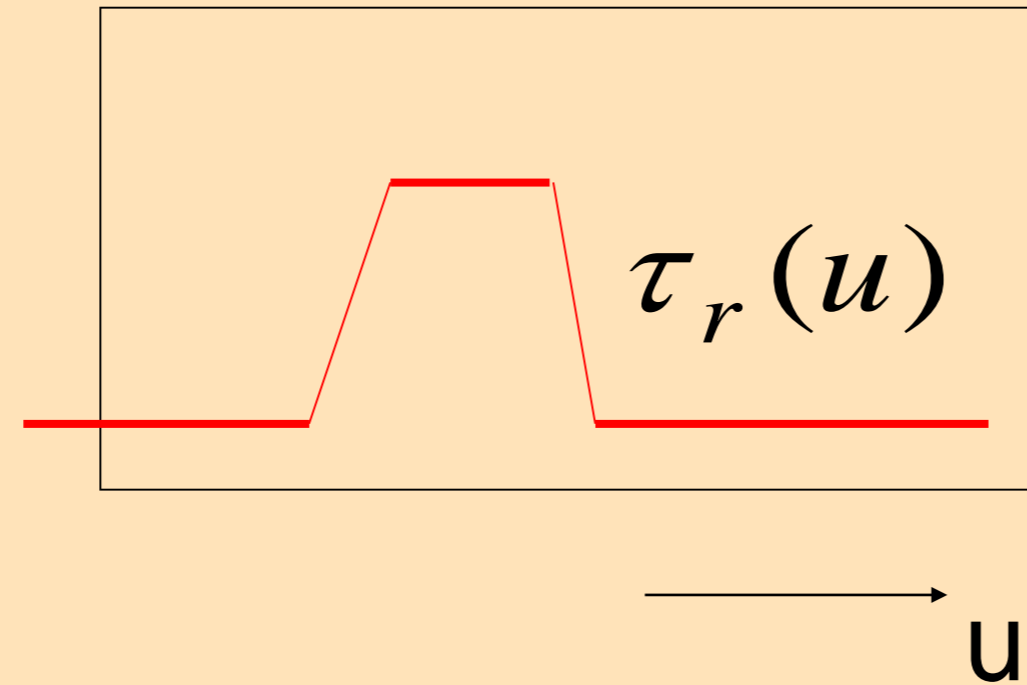
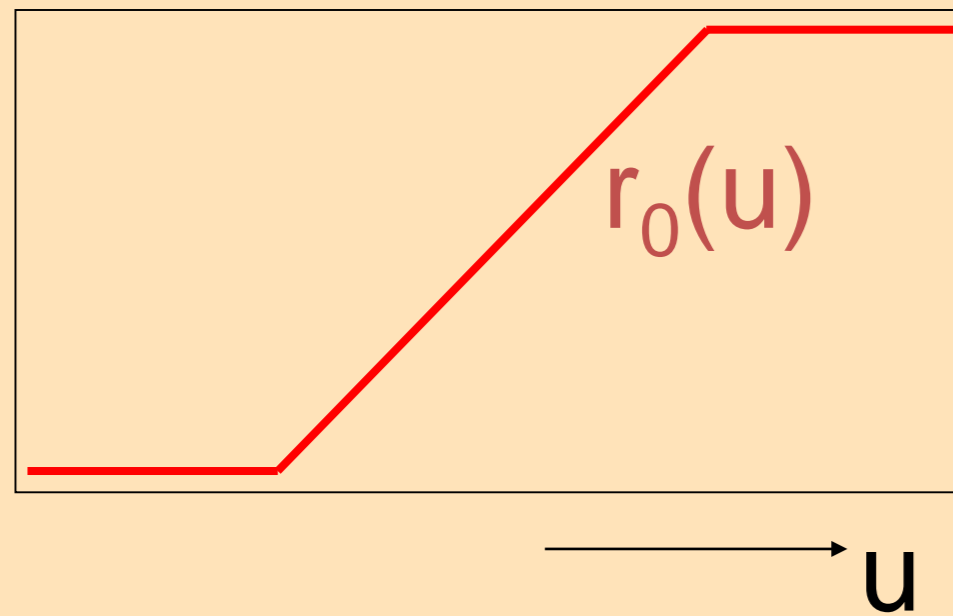


$$I_{ion} = -g_{ion} r^{n_1} s^{n_2} (u - E_{ion})$$

$$\frac{dr}{dt} = - \frac{r - r_0(u)}{\tau_r(u)}$$

$$\frac{ds}{dt} = - \frac{s - s_0(u)}{\tau_r(u)}$$

# Exercise 2 and 1.2 NOW!! - Ion channel



$$C \frac{du}{dt} = -g_{ion} r^{n_1} s^{n_2} (u - E_{ion}) + I(t)$$

$$\frac{dr}{dt} = -\frac{r - r_0(u)}{\tau_r(u)}$$

## Exercises

**1 and 2 NOW!**

**If finished, start  
Exercise 3.**

**This will be a preparation  
For Next lecture**

**At 11:15-12:05**

**Start Exercise 2 at 10:30.  
Continue with Exercise 1.2**

**Next lecture at:  
10H50**

# Week 2 – part 4: Threshold in the Hodgkin-Huxley Model



## Biological Modeling of Neural Networks

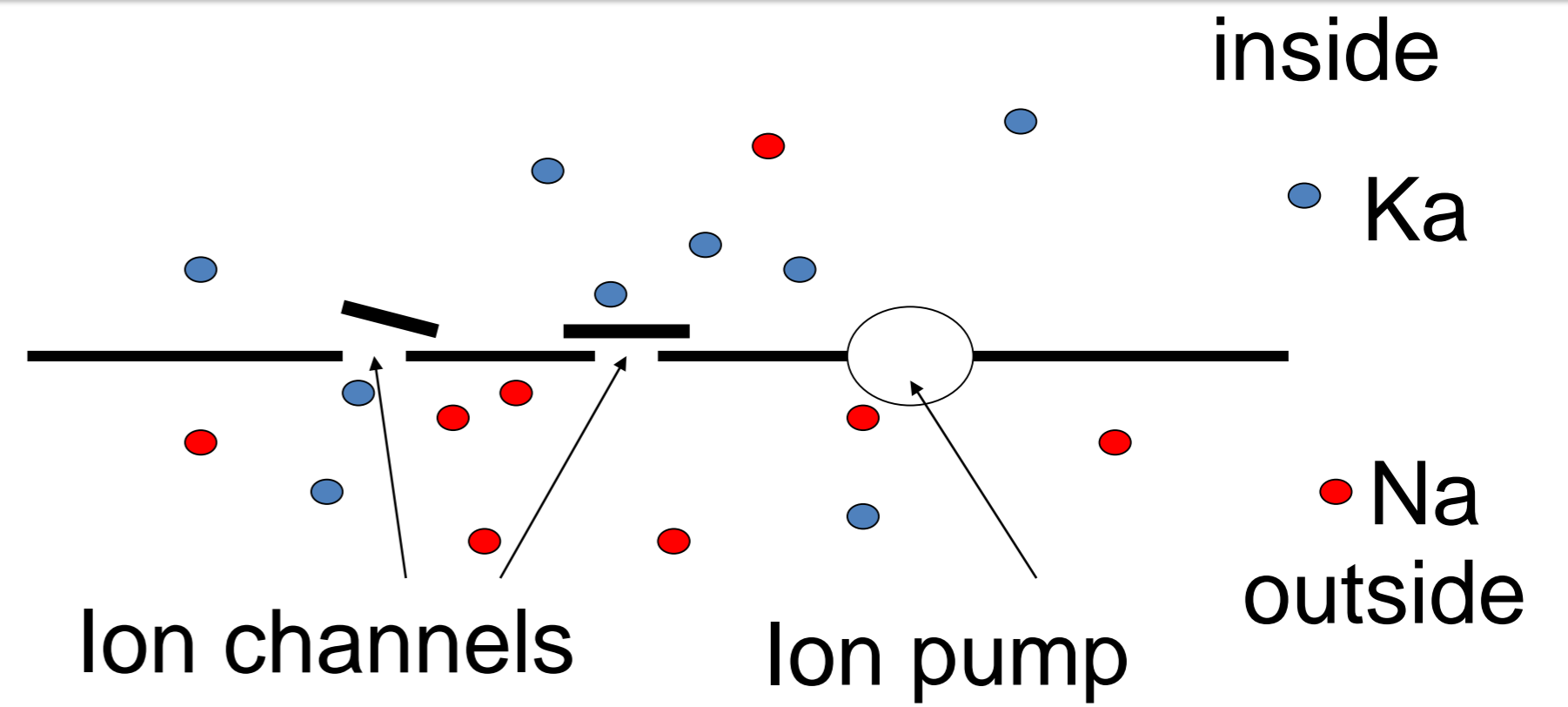
### Week 2 – Biophysical modeling: The Hodgkin-Huxley model

Wulfram Gerstner

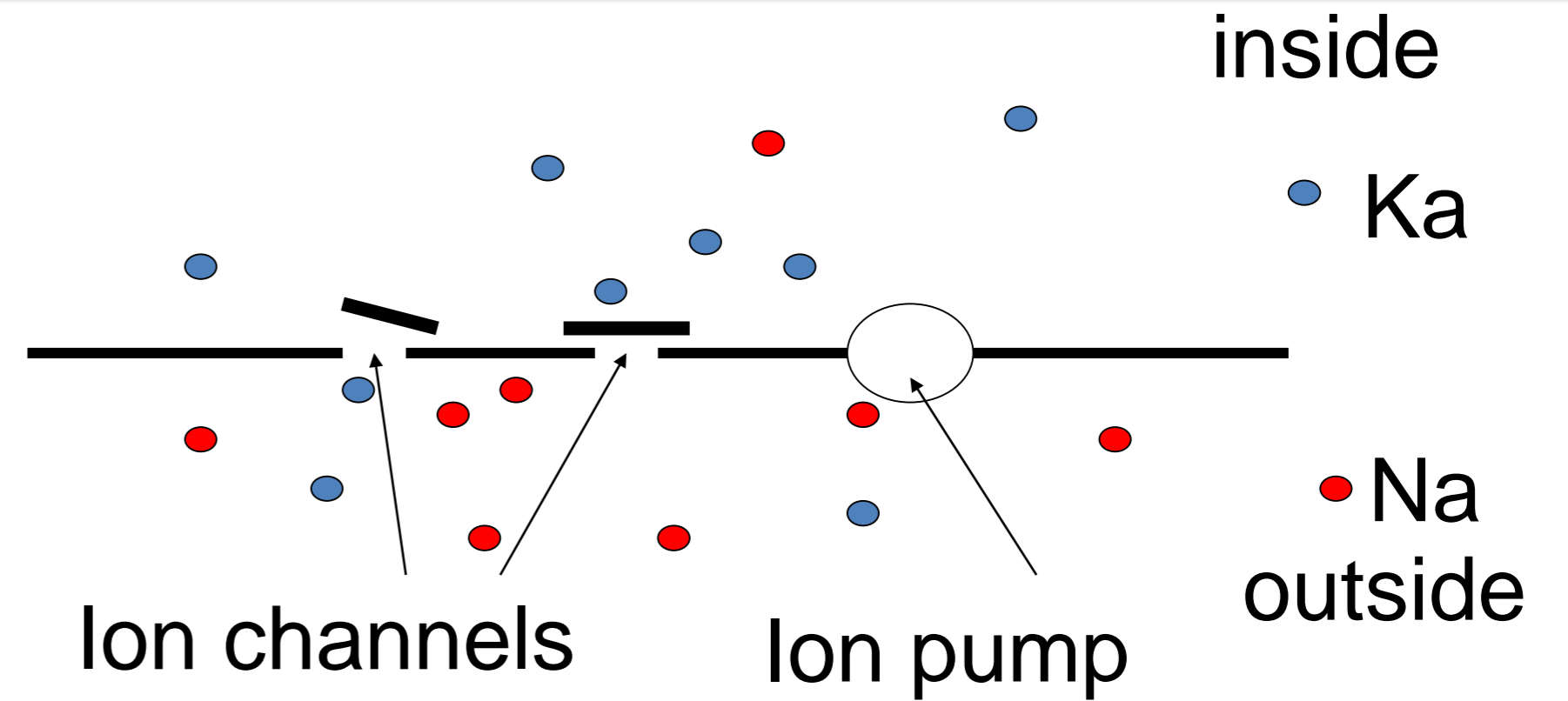
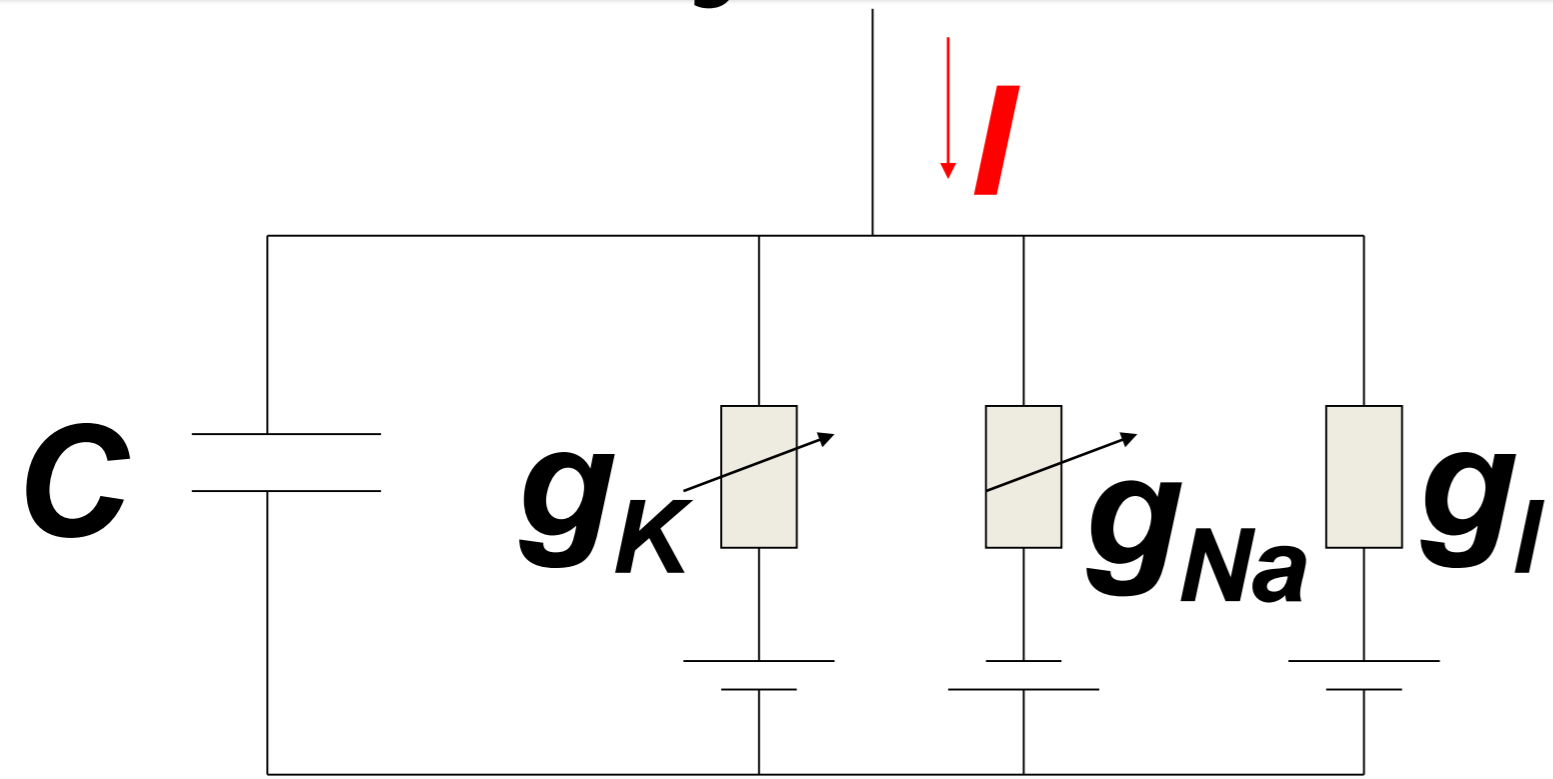
EPFL, Lausanne, Switzerland

- ✓ 2.1 Biophysics of neurons
  - Overview
- ✓ 2.2 Reversal potential
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- ✓ 2.3 Hodgkin-Huxley Model
- 2.4 Threshold in the Hodgkin-Huxley Model
  - where is the firing threshold?
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# Neuronal Dynamics – 2.4. Threshold in HH model



# Neuronal Dynamics – 2.4. Threshold in HH model



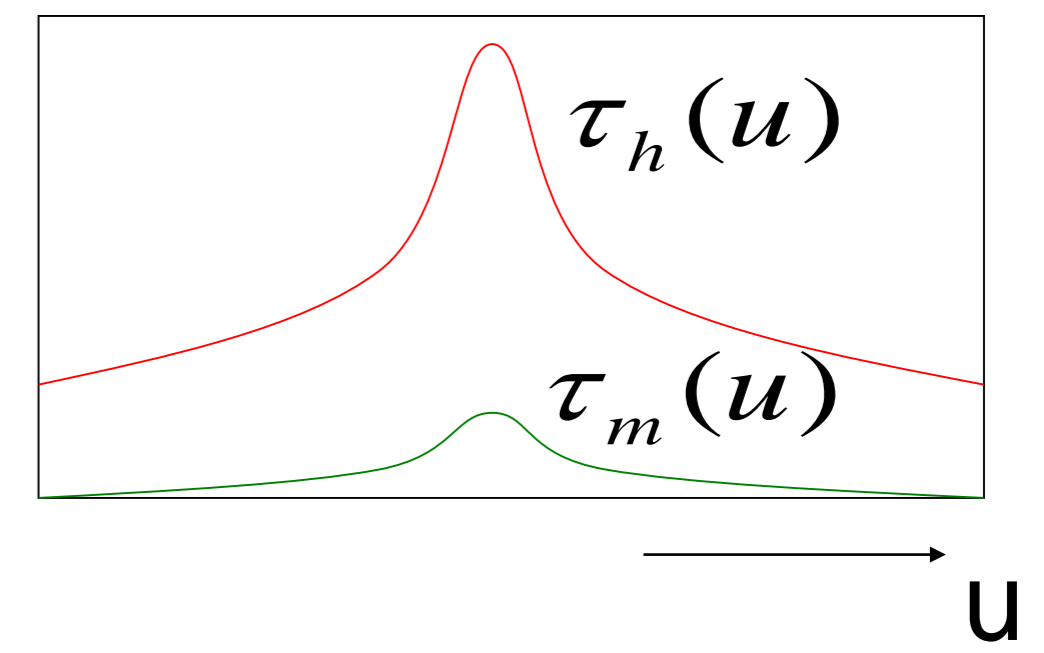
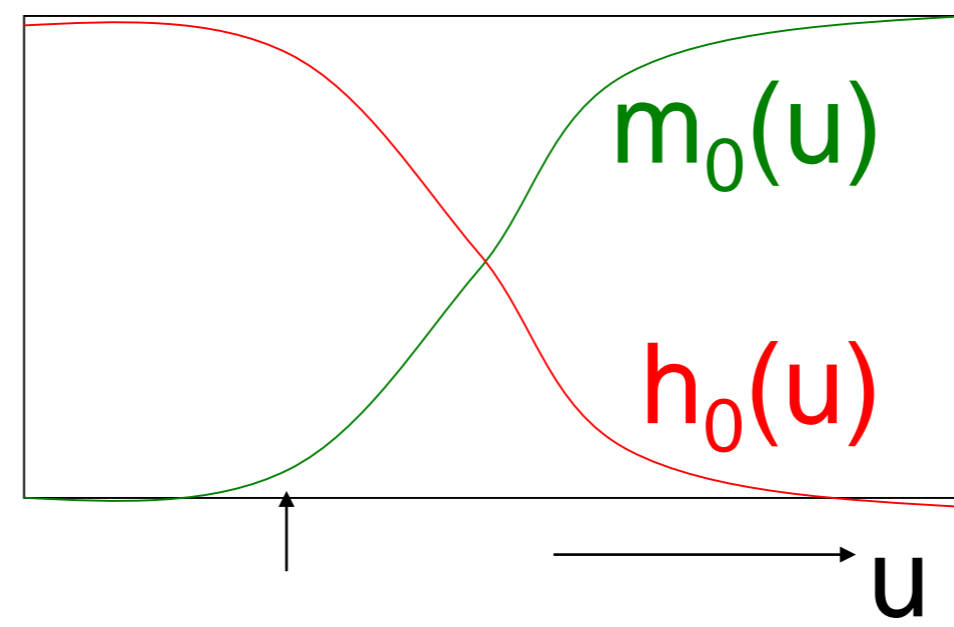
$$C \frac{du}{dt} = -g_{Na} m^3 h (u - E_{Na}) - g_l (u - E_l) + I(t)$$

*(Note: Red brackets in the original image group the sodium current term as  $I_{Na}$  and the potassium/leakage current term as  $I_K$ )*

**Where is the threshold for firing?**

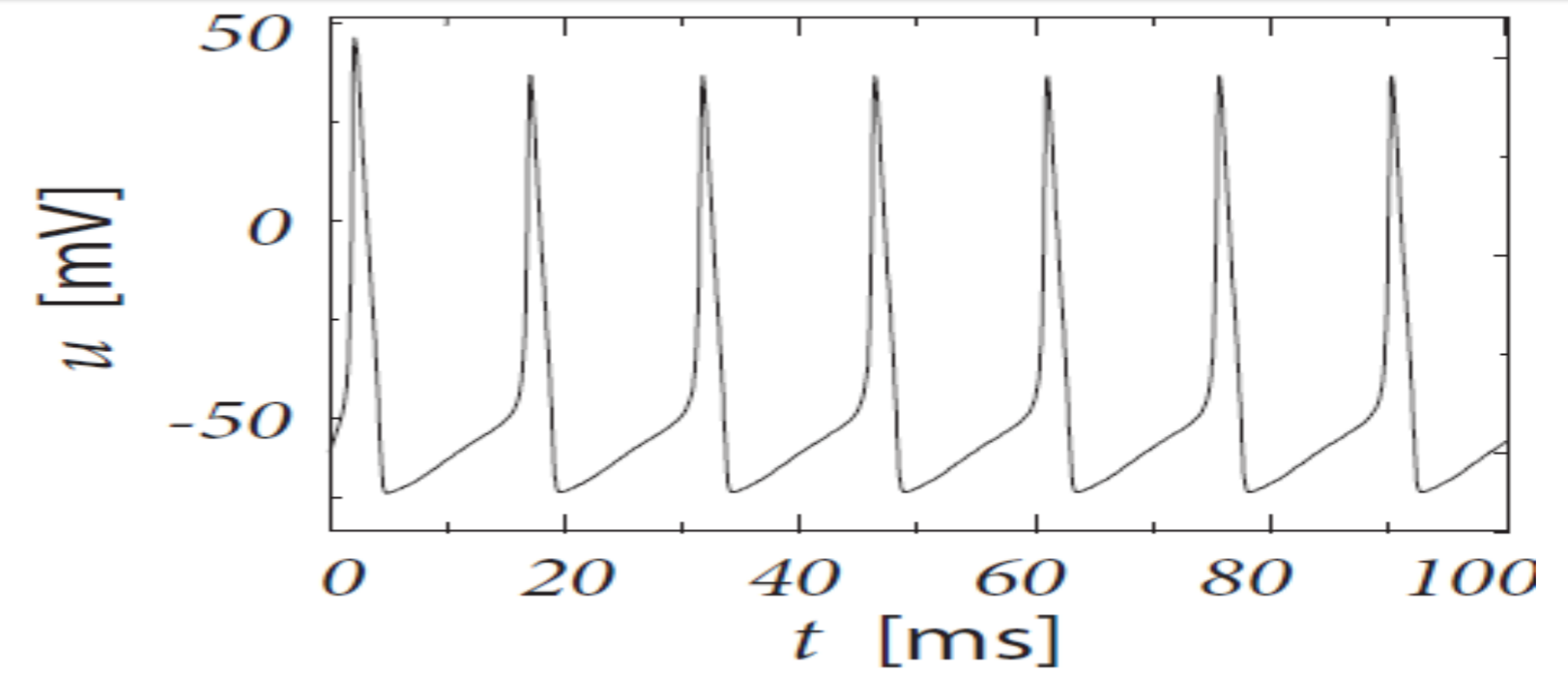
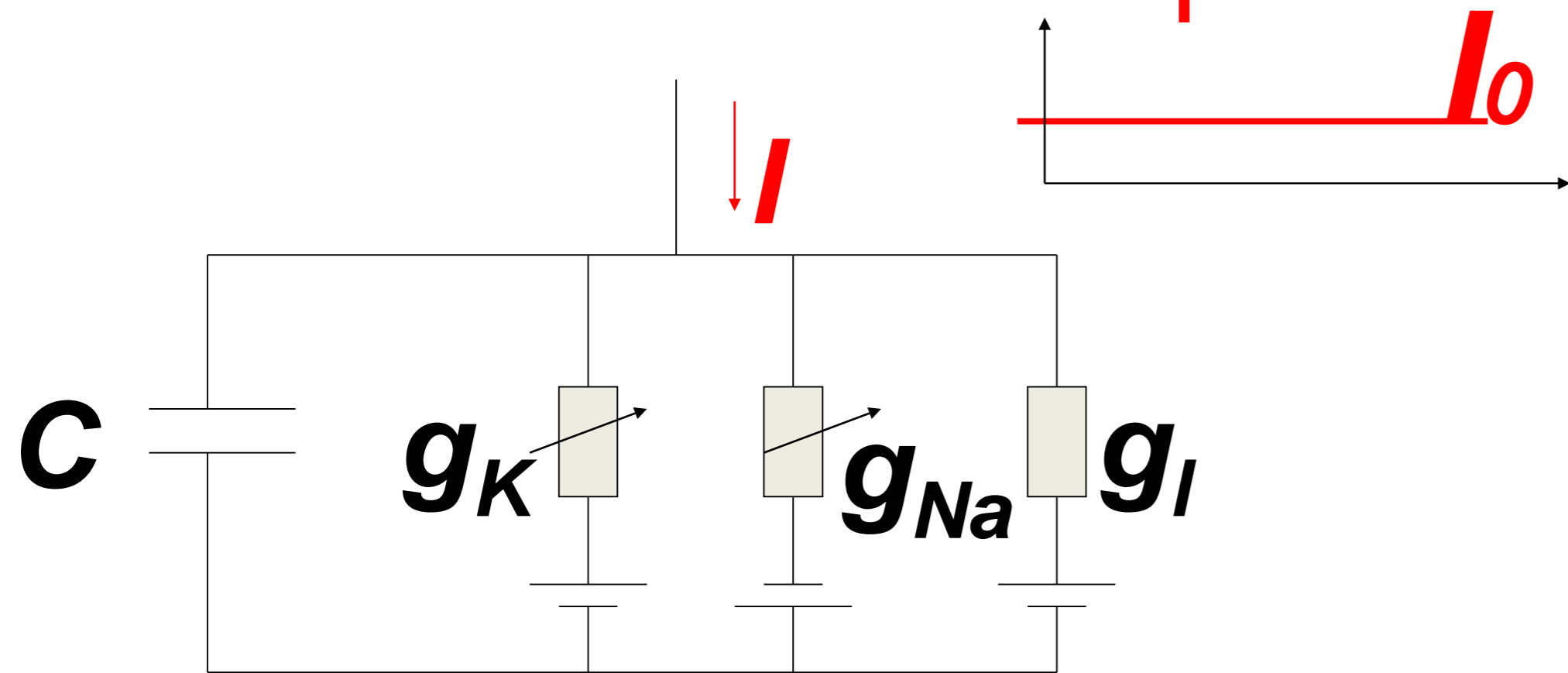
$$\frac{dm}{dt} = \frac{m - m_0(u)}{\tau_m(u)}$$

$$\frac{dh}{dt} = \frac{h - h_0(u)}{\tau_h(u)}$$



# Neuronal Dynamics – 2.4. Threshold in HH model

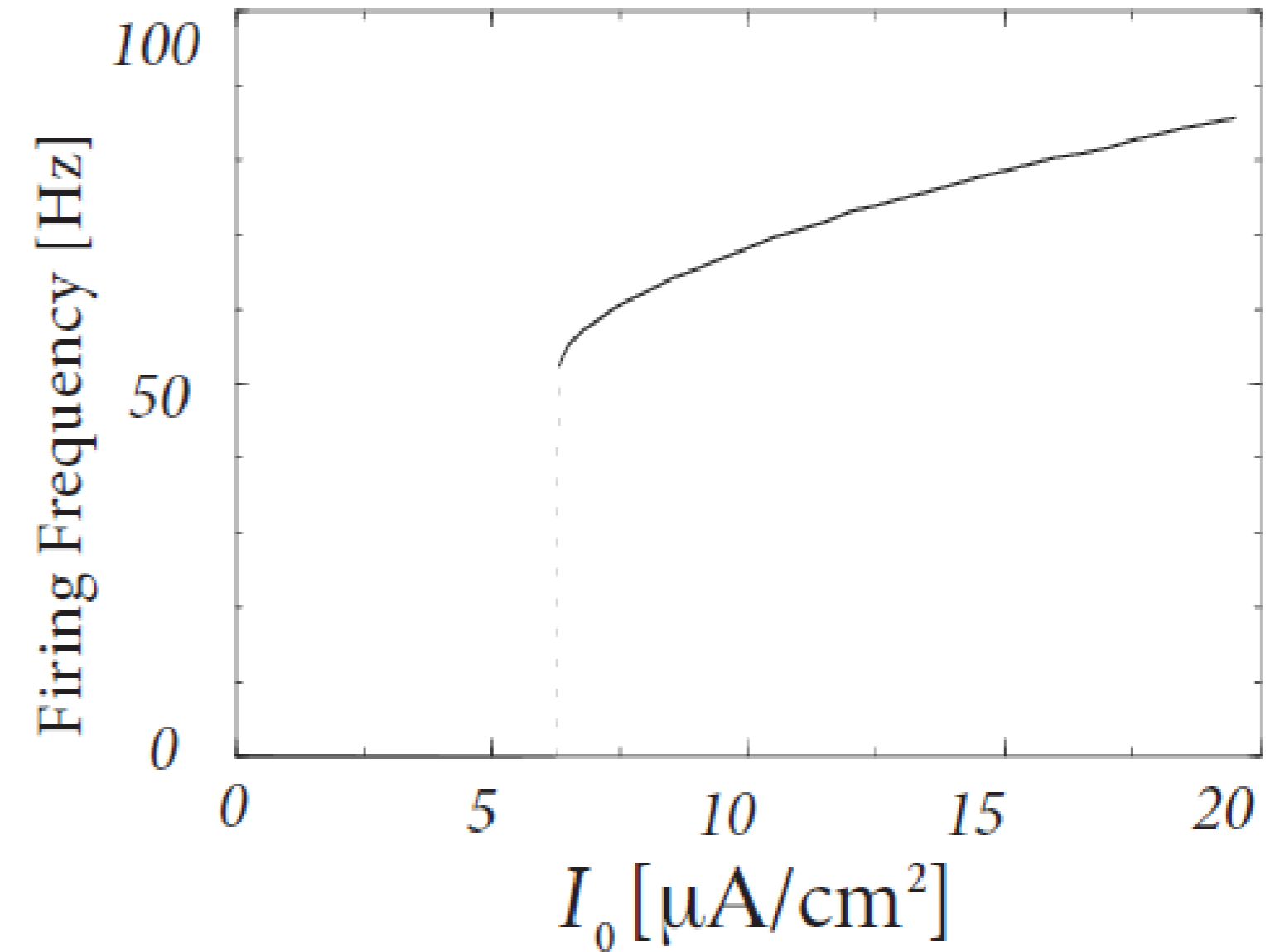
Constant current input



**Threshold?**

for repetitive firing

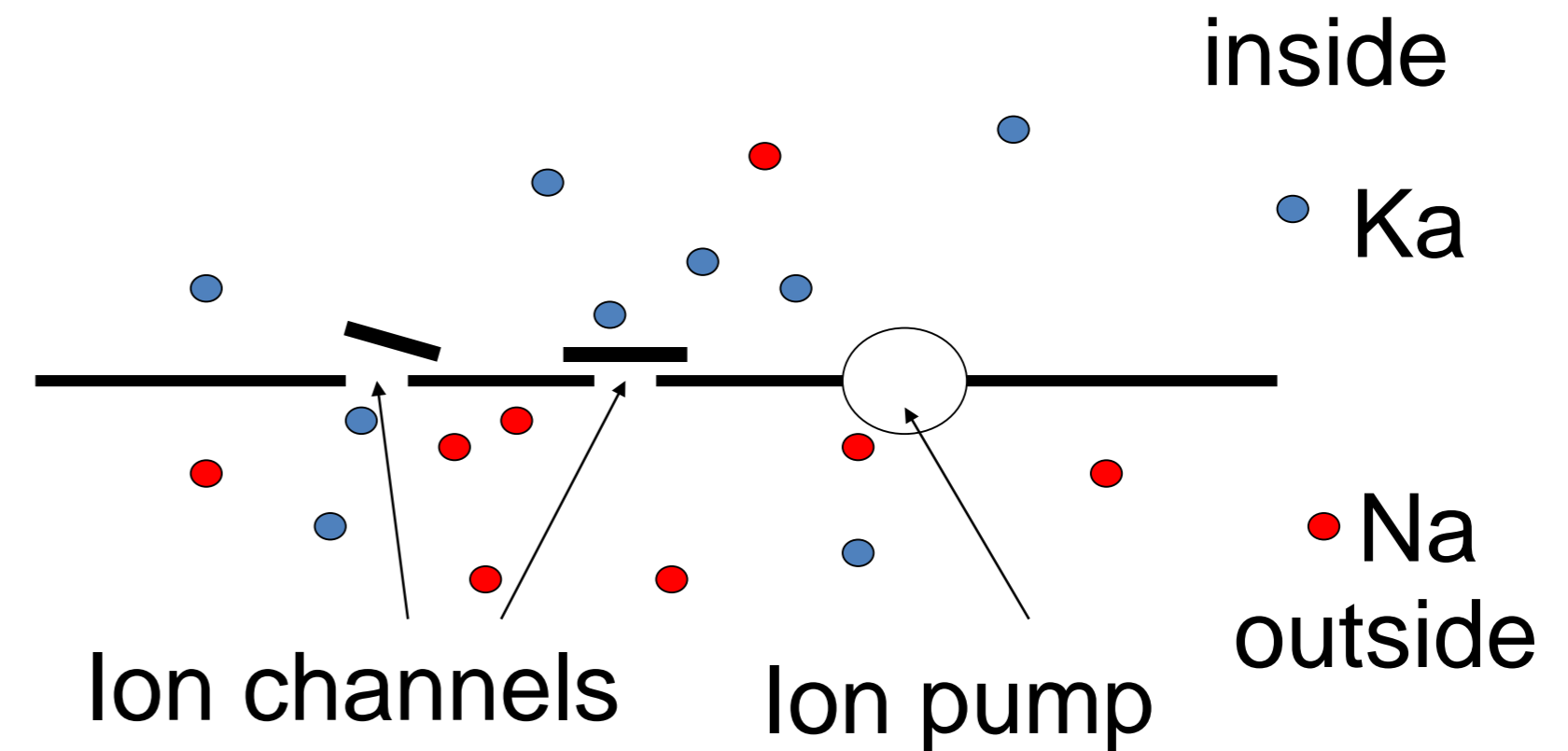
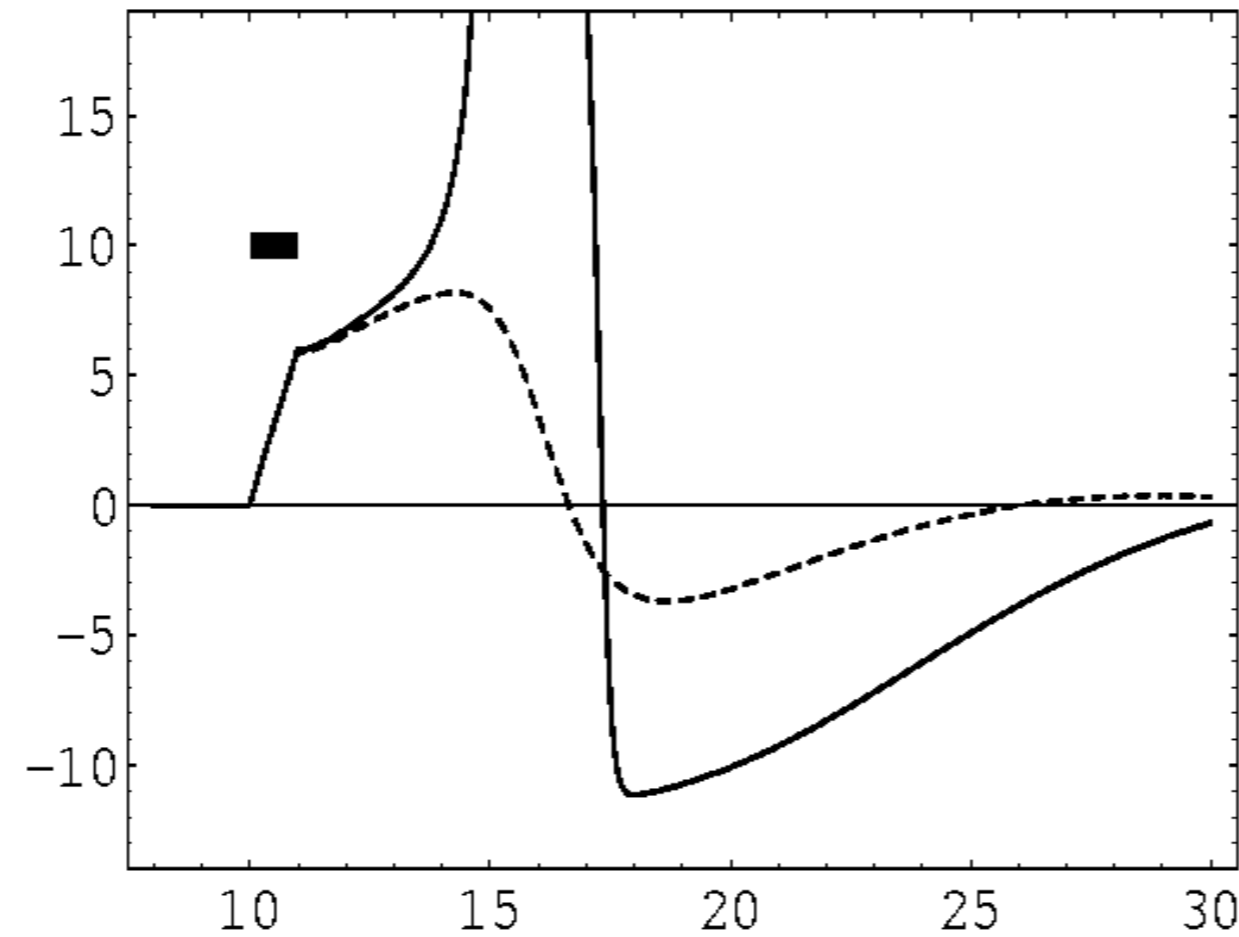
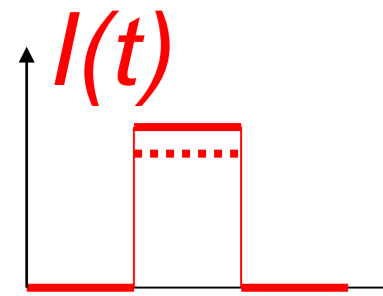
(**current** threshold)





# Neuronal Dynamics – 2.4. Threshold in HH model

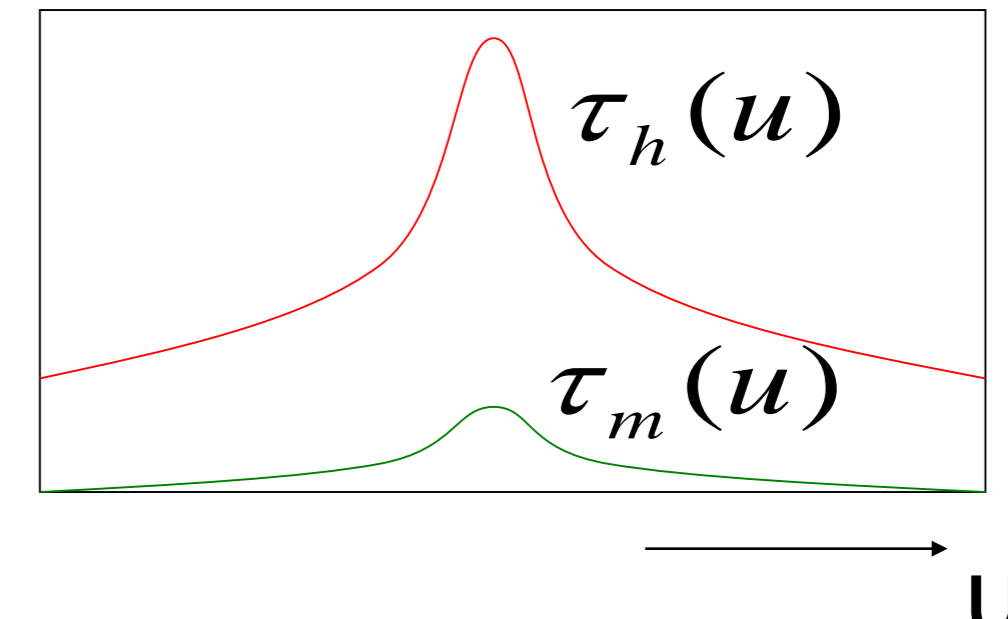
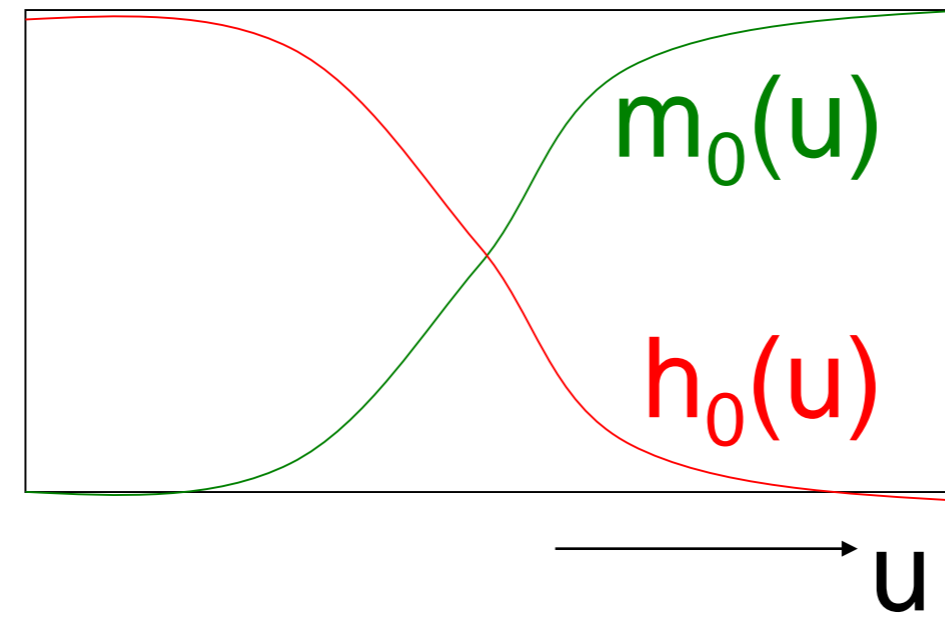
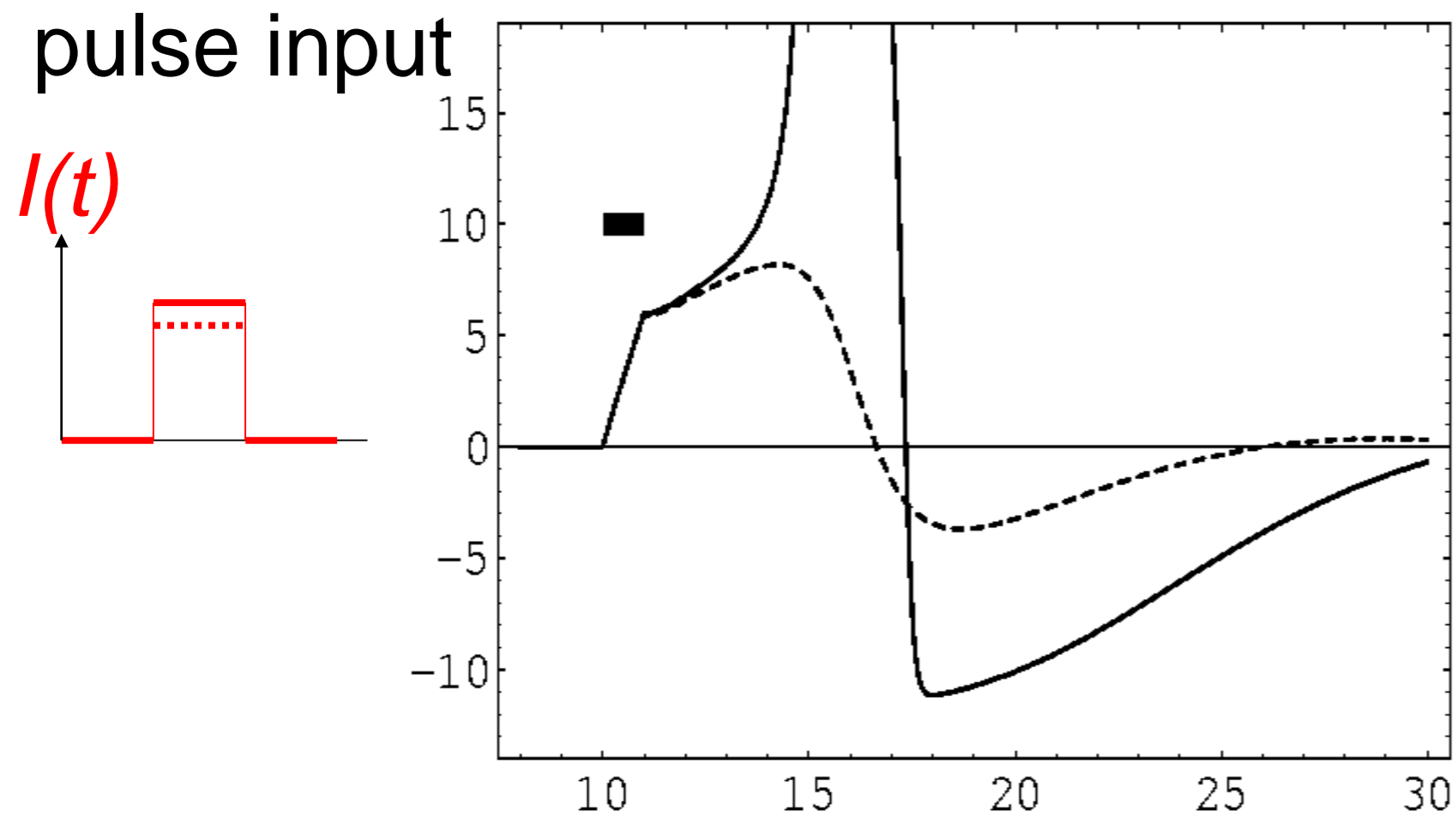
pulse input



Threshold?

- AP if amplitude 7.0 units
- No AP if amplitude 6.9 units  
(pulse with 1ms duration)  
(and pulse with 0.5 ms duration?)

# Neuronal Dynamics – 2.4. Threshold in HH model



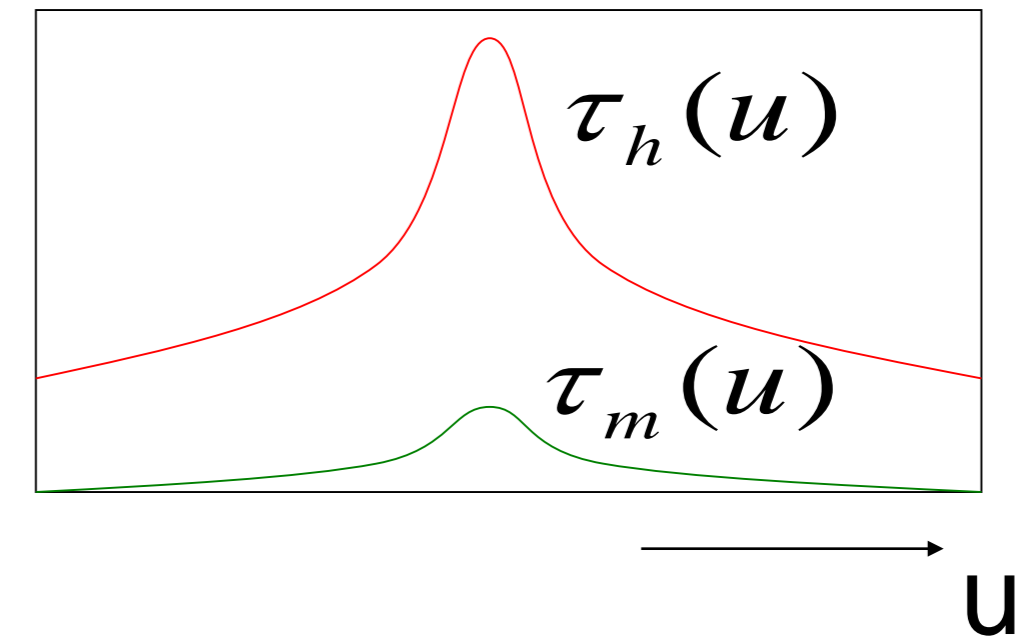
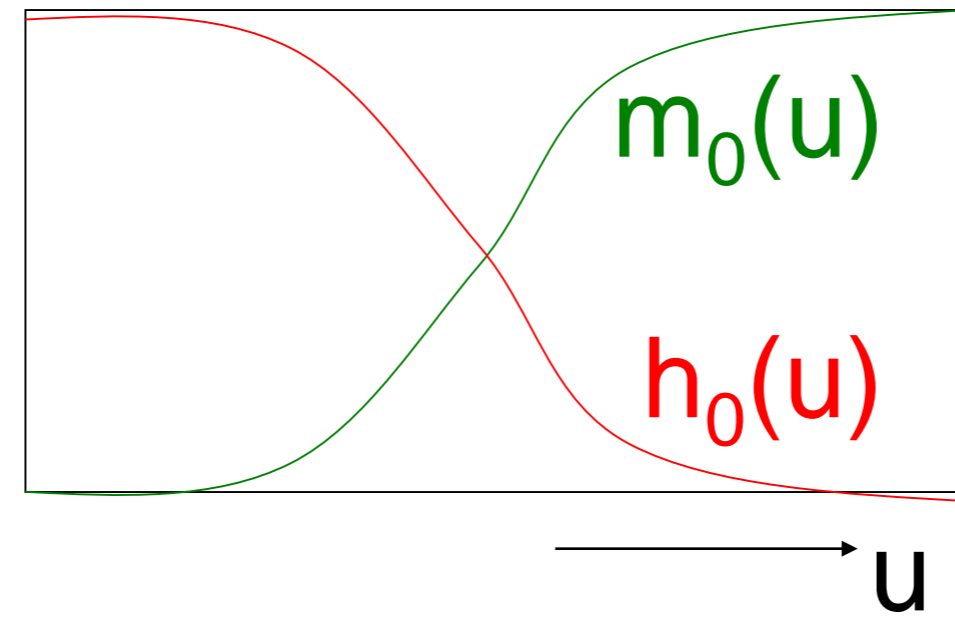
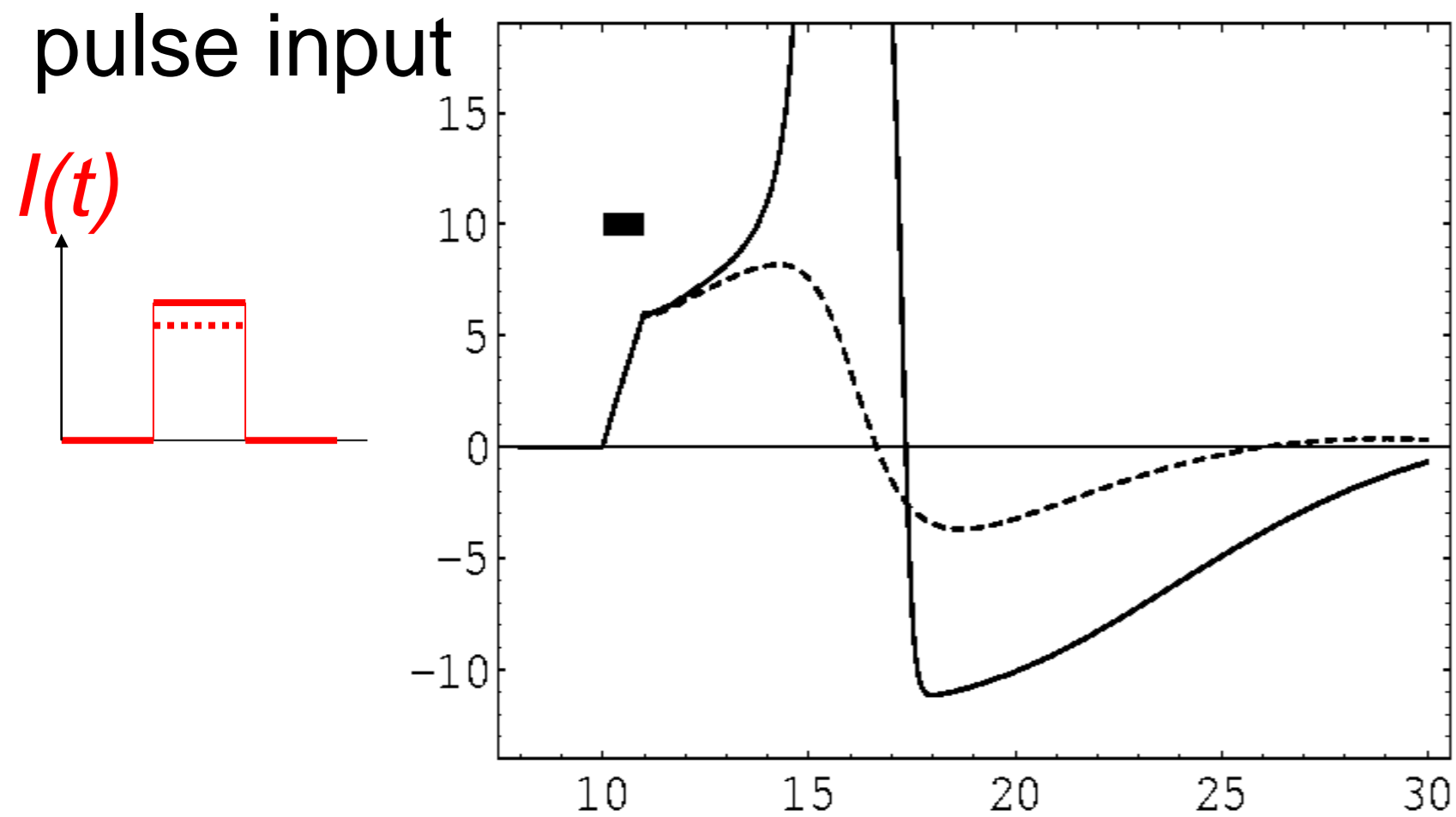
**Mathematical explanation**

$$C \frac{du}{dt} = \underbrace{-g_{Na} m^3 h}_{I_{Na}} (u - E_{Na}) - \underbrace{g_K n^4}_{I_K} (u - E_K) - \underbrace{g_l}_{I_{leak}} (u - E_l) + I(t) \quad \text{Stim.} \downarrow$$

$$\frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m(u)}$$

$$\frac{dh}{dt} = -\frac{h - h_0(u)}{\tau_h(u)}$$

# Neuronal Dynamics – 2.4. Threshold in HH model



Why start the explanation with  $m$  and not  $h$ ?

What about  $n$ ?

Where is the threshold?

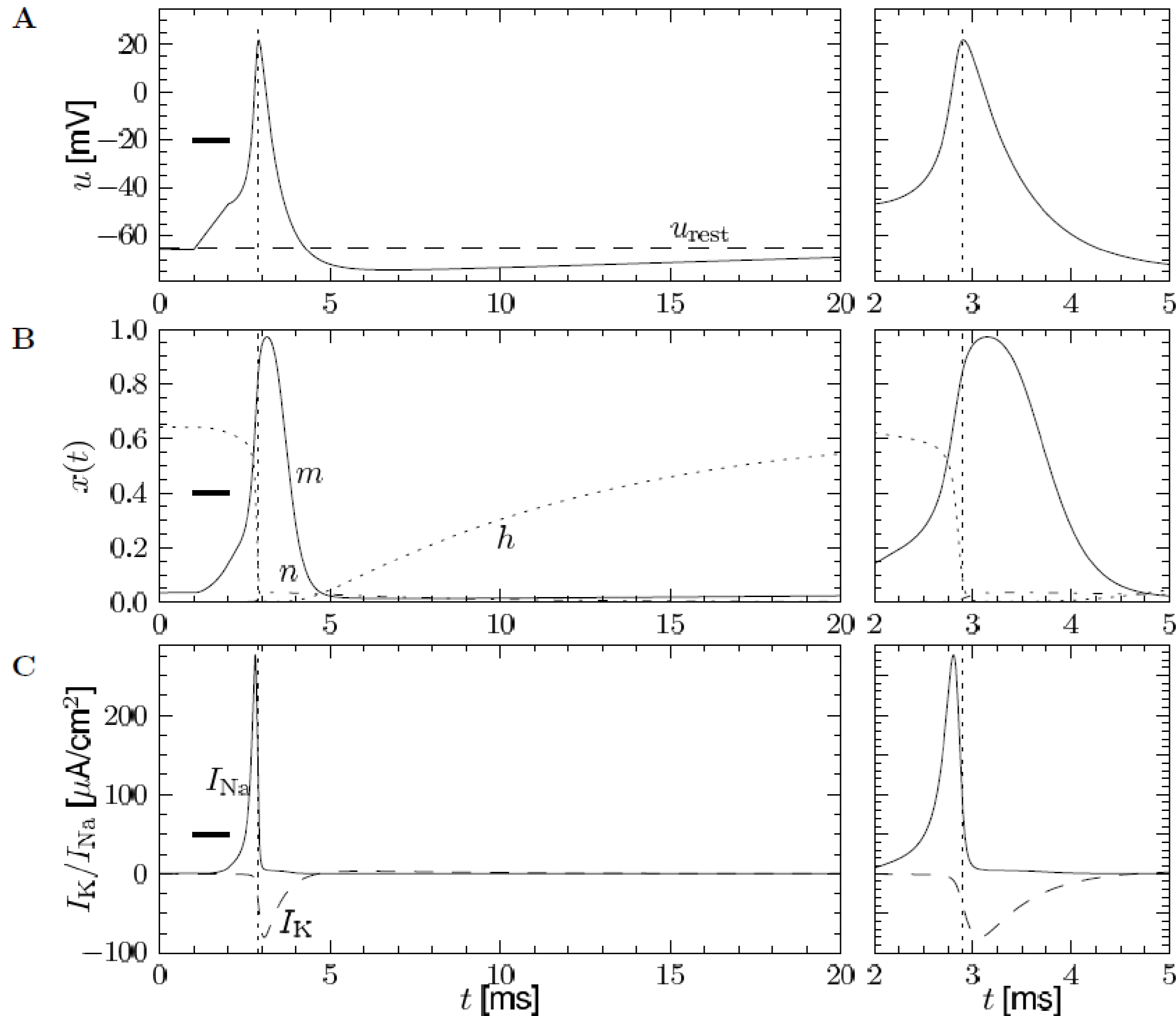
$$C \frac{du}{dt} = \underbrace{-g_{Na} m^3 h}_{I_{Na}} (u - E_{Na}) - \underbrace{g_K n^4}_{I_K} (u - E_K) - \underbrace{g_l}_{I_{leak}} (u - E_l) + I(t) \quad \text{Stim.} \downarrow$$

$$\frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m(u)}$$

$$\frac{dh}{dt} = -\frac{h - h_0(u)}{\tau_h(u)}$$

$$\frac{dn}{dt} = -\frac{n - n_0(u)}{\tau_n(u)}$$

# Neuronal Dynamics – 2.4. Threshold in HH model



$$C \frac{du}{dt} = -g_{Na} m^3 h (u - E_{Na}) - g_K n^4 (u - E_K) - g_l (u - E_l) + I(t)$$

# Neuronal Dynamics – 2.4. Threshold in HH model

First conclusion:

There is no strict threshold:

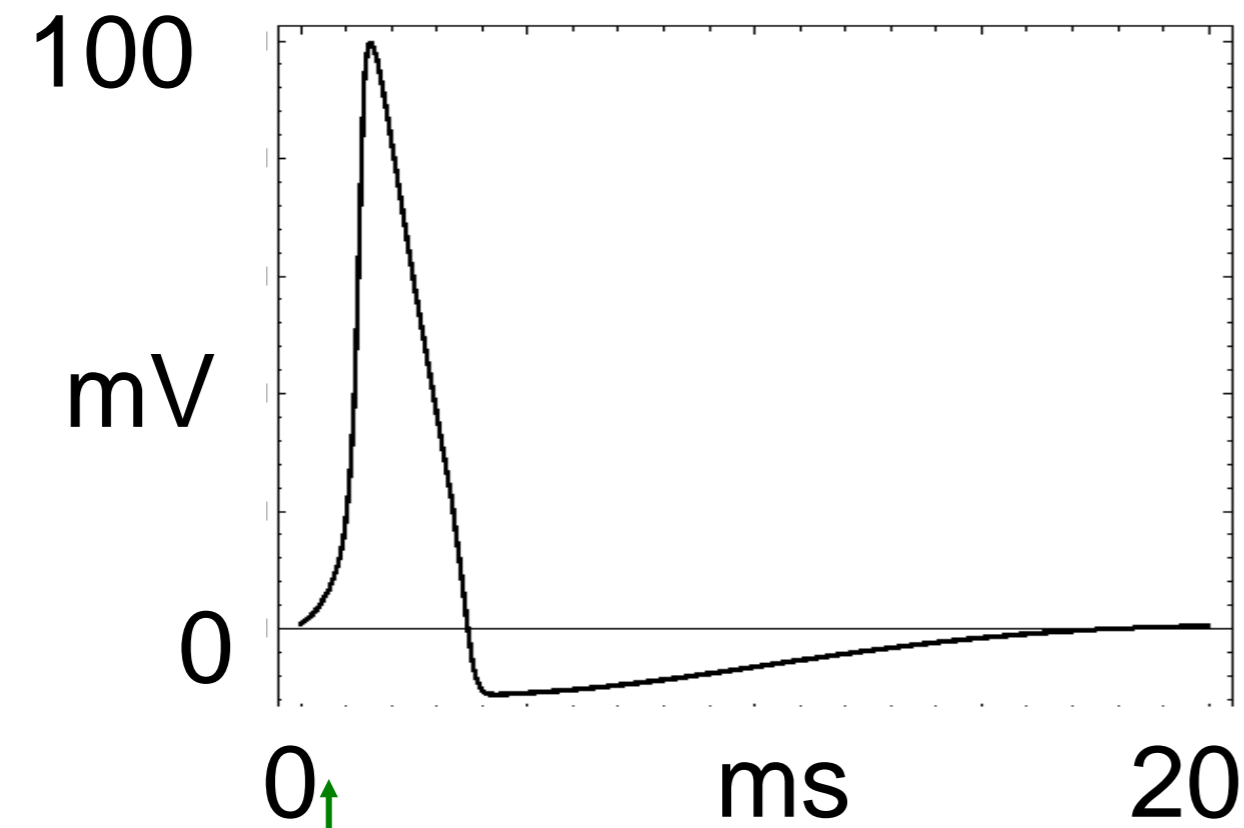
Coupled differential equations

*‘Effective’* threshold  
in simulations?

# Neuronal Dynamics – 2.4. Refractoriness in HH model

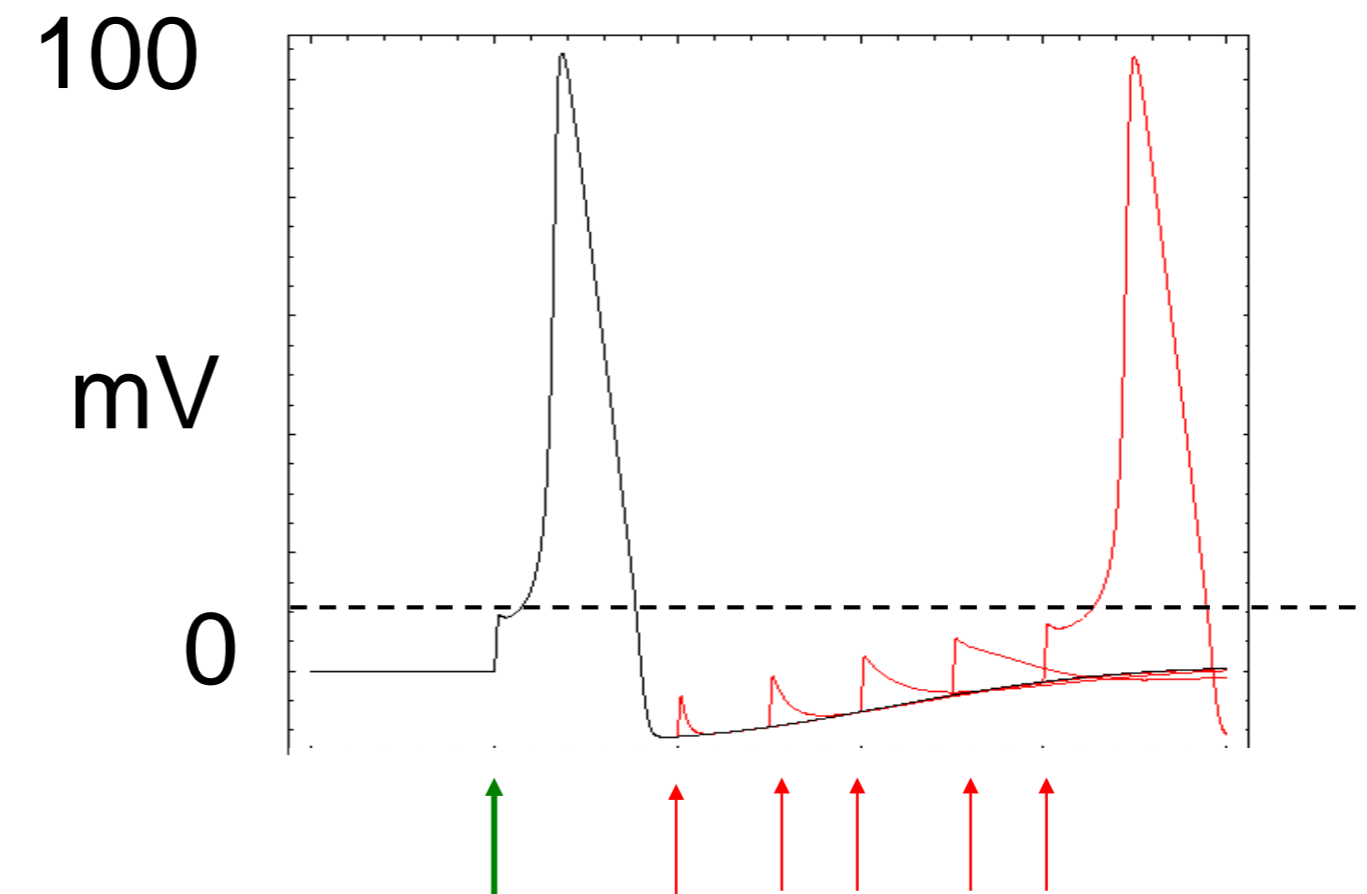
Where is the firing threshold?

## Action potential



Strong stimulus

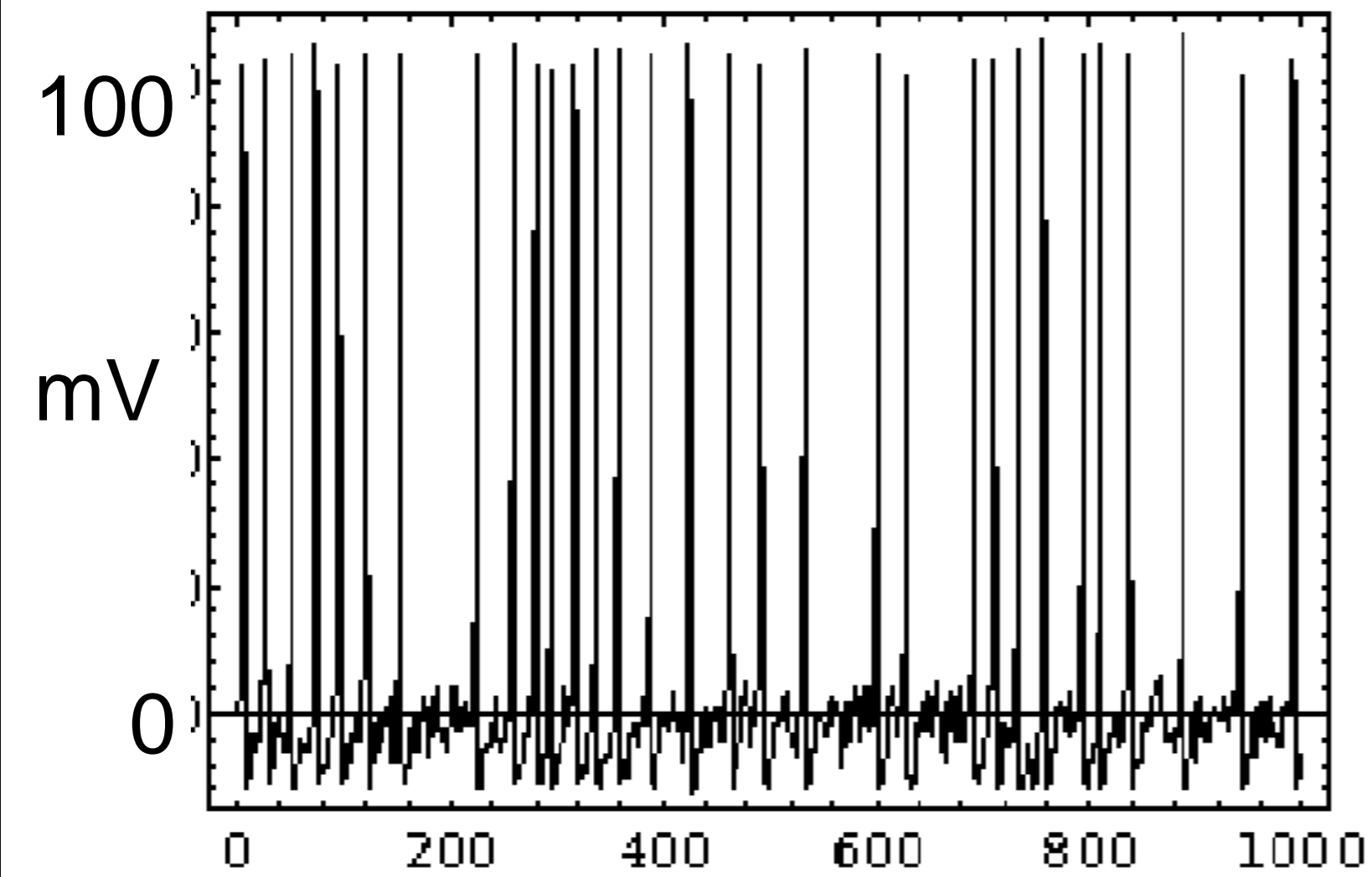
## refractoriness



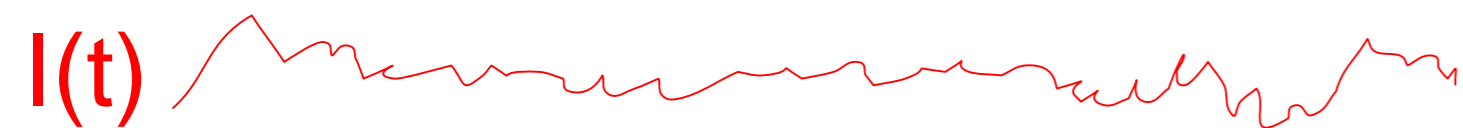
Strong stimulus  
strong stimuli

***Refractoriness!*** Harder to elicit a second spike

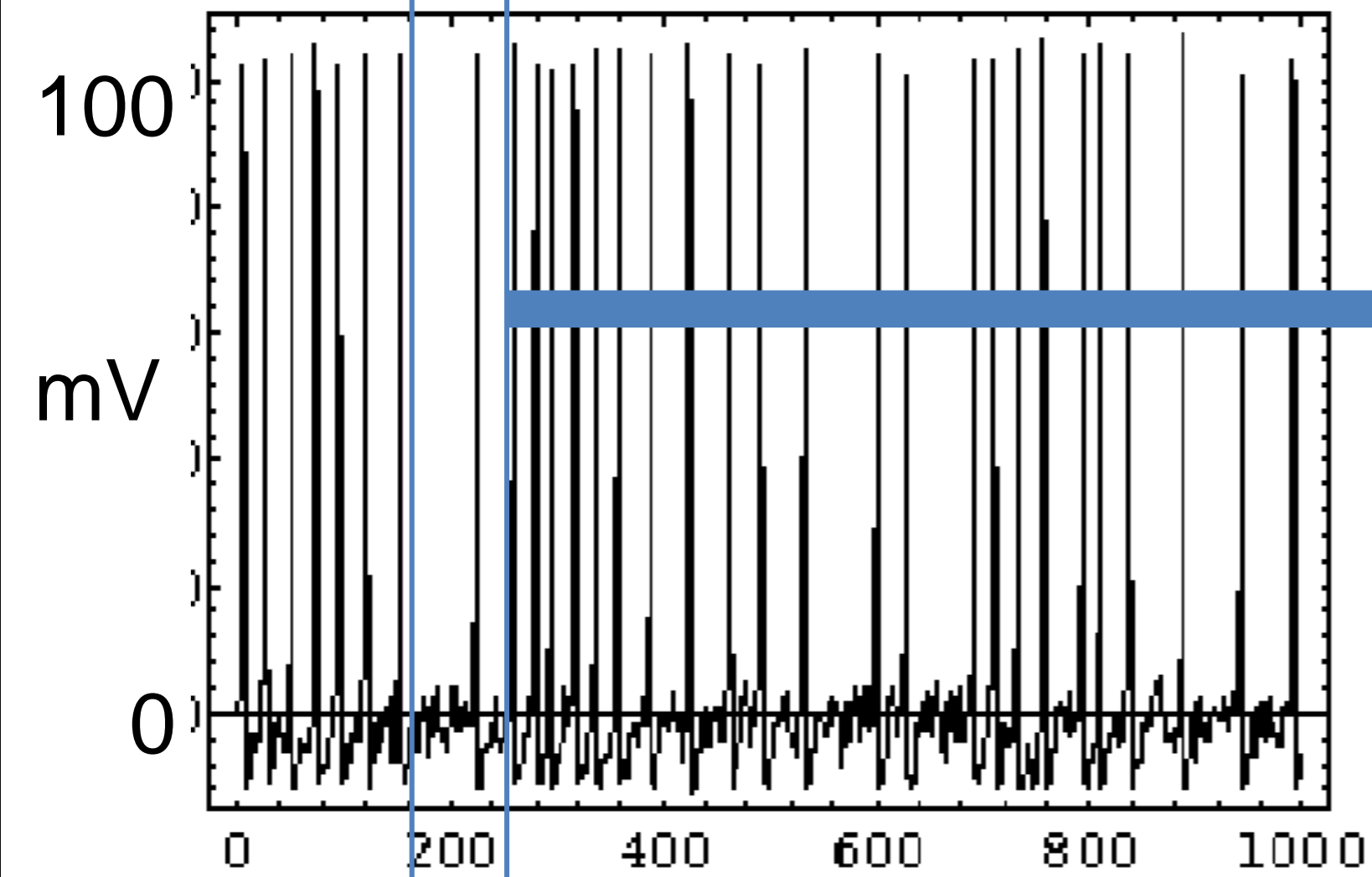
# Neuronal Dynamics – 2.4. Simulations of the HH model



Stimulation with  
time-dependent  
input current

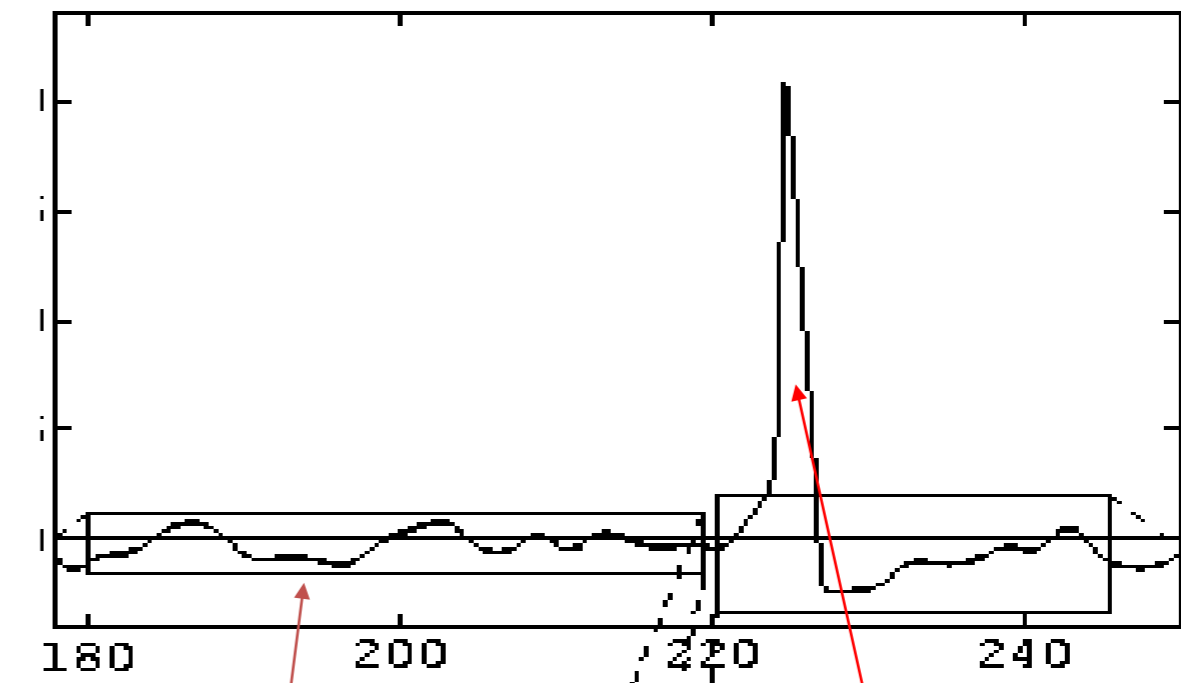


# Neuronal Dynamics – 2.4. Simulations of the HH model



$I(t)$

mV



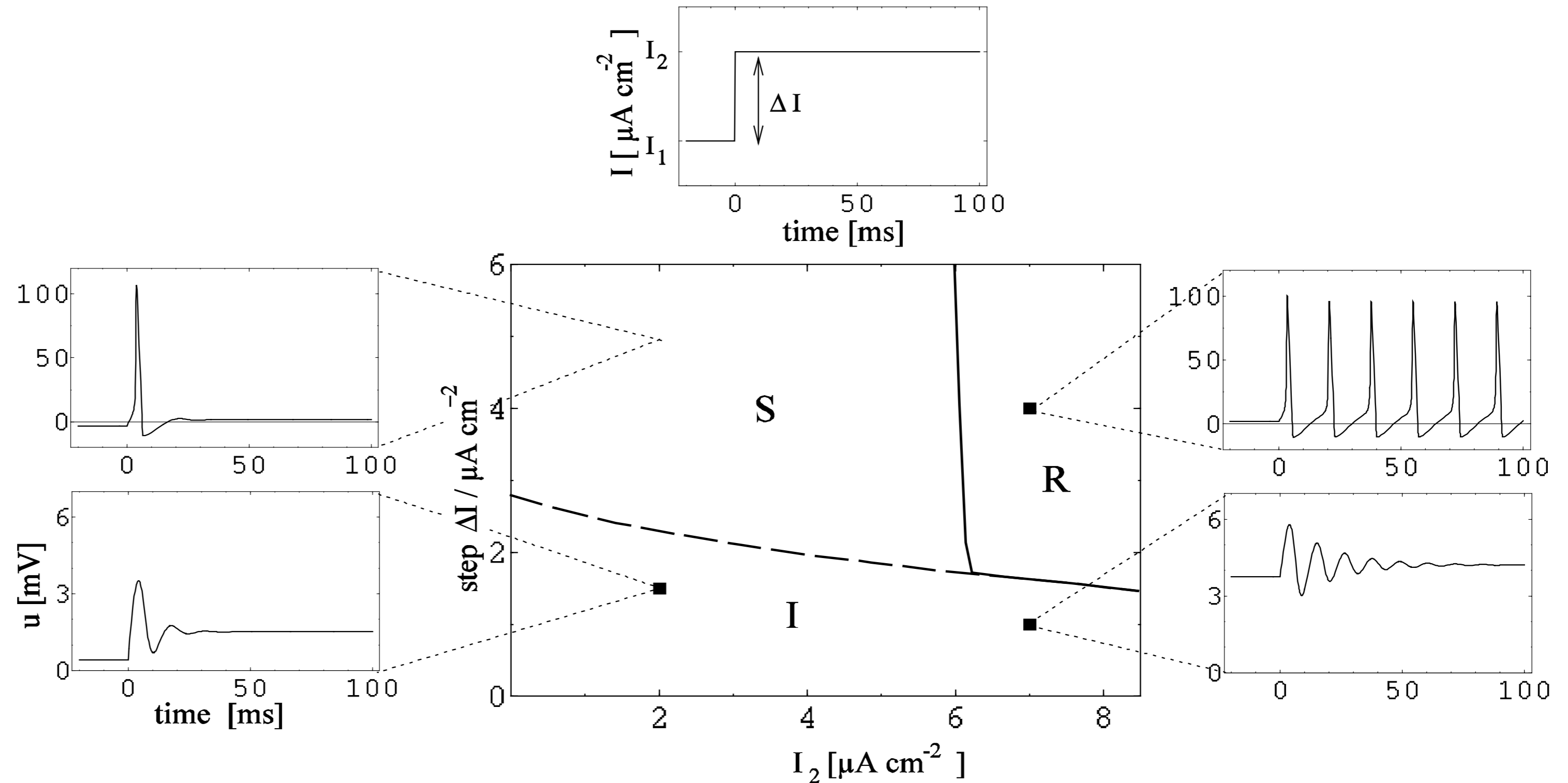
**Subthreshold response**

**Spike**



# Neuronal Dynamics – 2.4. Threshold in HH model

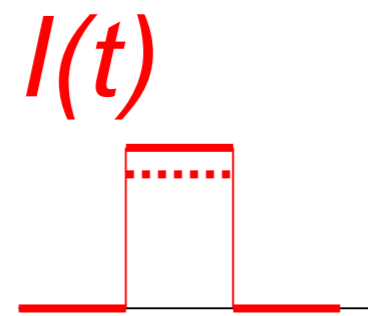
Step current input  $I_2$



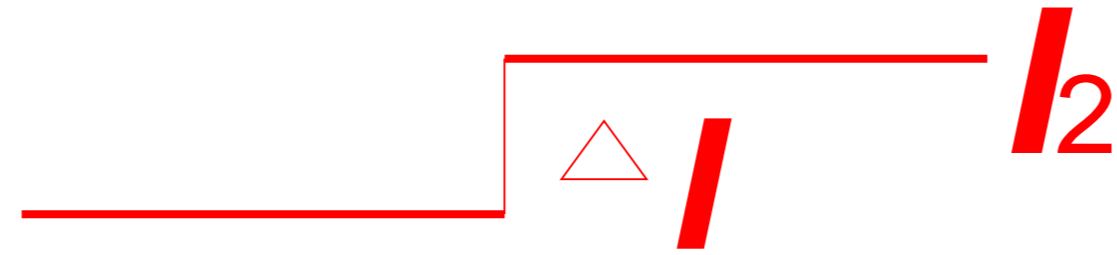
# Neuronal Dynamics – 2.4. Threshold in HH model

## Where is the firing threshold?

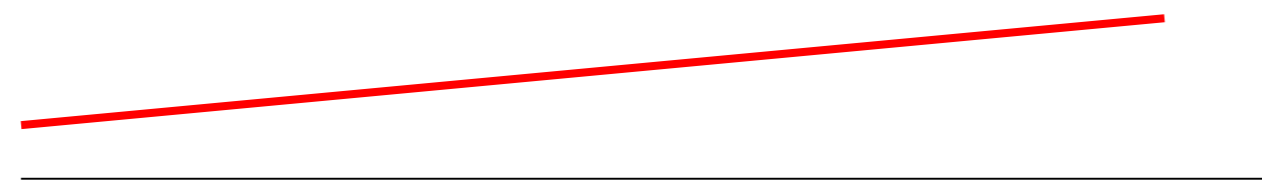
pulse input



step input



ramp input



**There is no threshold**

- no current threshold
- no voltage threshold

‘effective’ threshold

- depends on typical input

$$C \frac{du}{dt} = -g_{Na} m^3 h (u - E_{Na}) - \dots$$

# Neuronal Dynamics – 2.4. Type I and Type II

Hodgkin-Huxley model  
with other parameters  
(e.g. for cortical pyramidal  
Neuron )

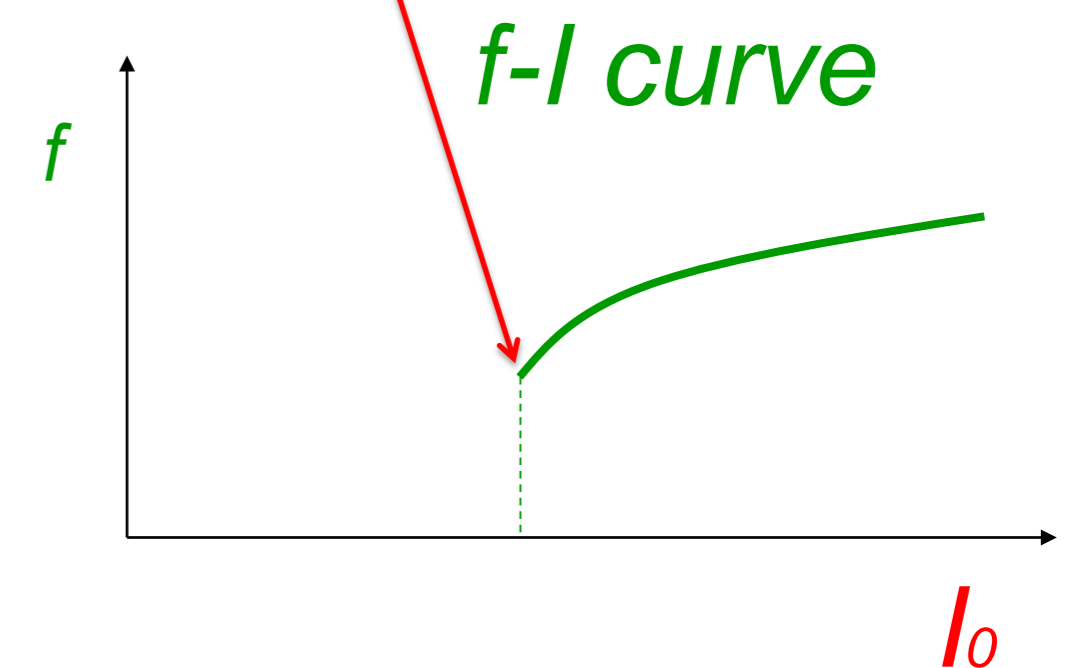
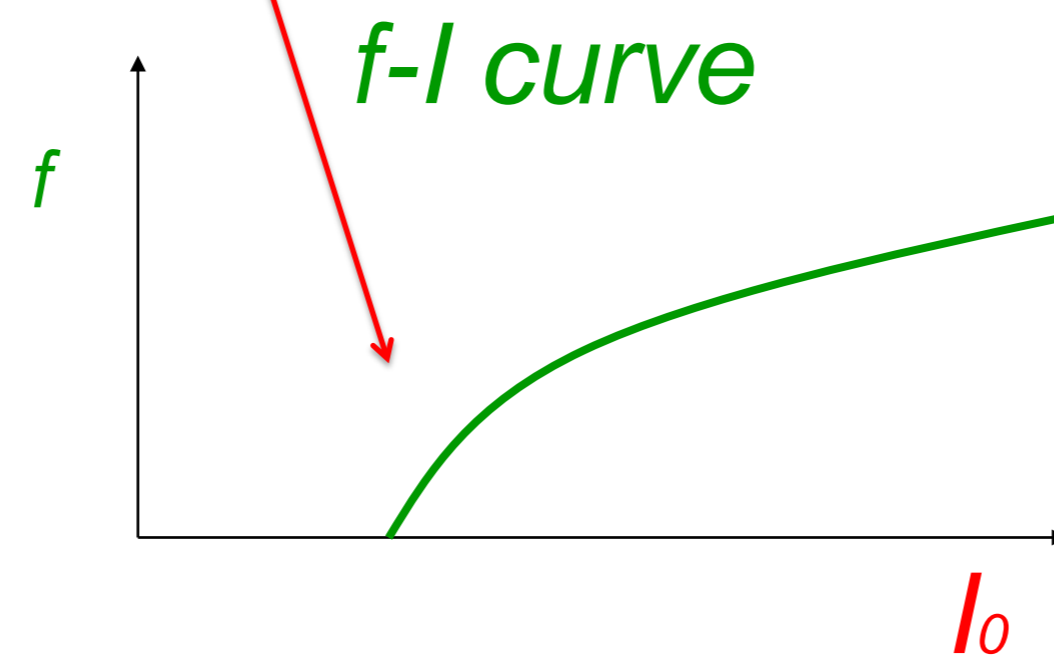
Hodgkin-Huxley model  
with standard parameters  
(giant axon of squid)

**Response at firing threshold?**

Type I

type II

ramp input/  
constant input



# Neuronal Dynamics – 2.4. Hodgkin-Huxley model

- 4 differential equations
- no explicit threshold
- effective threshold depends on stimulus
- BUT: voltage threshold good approximation**

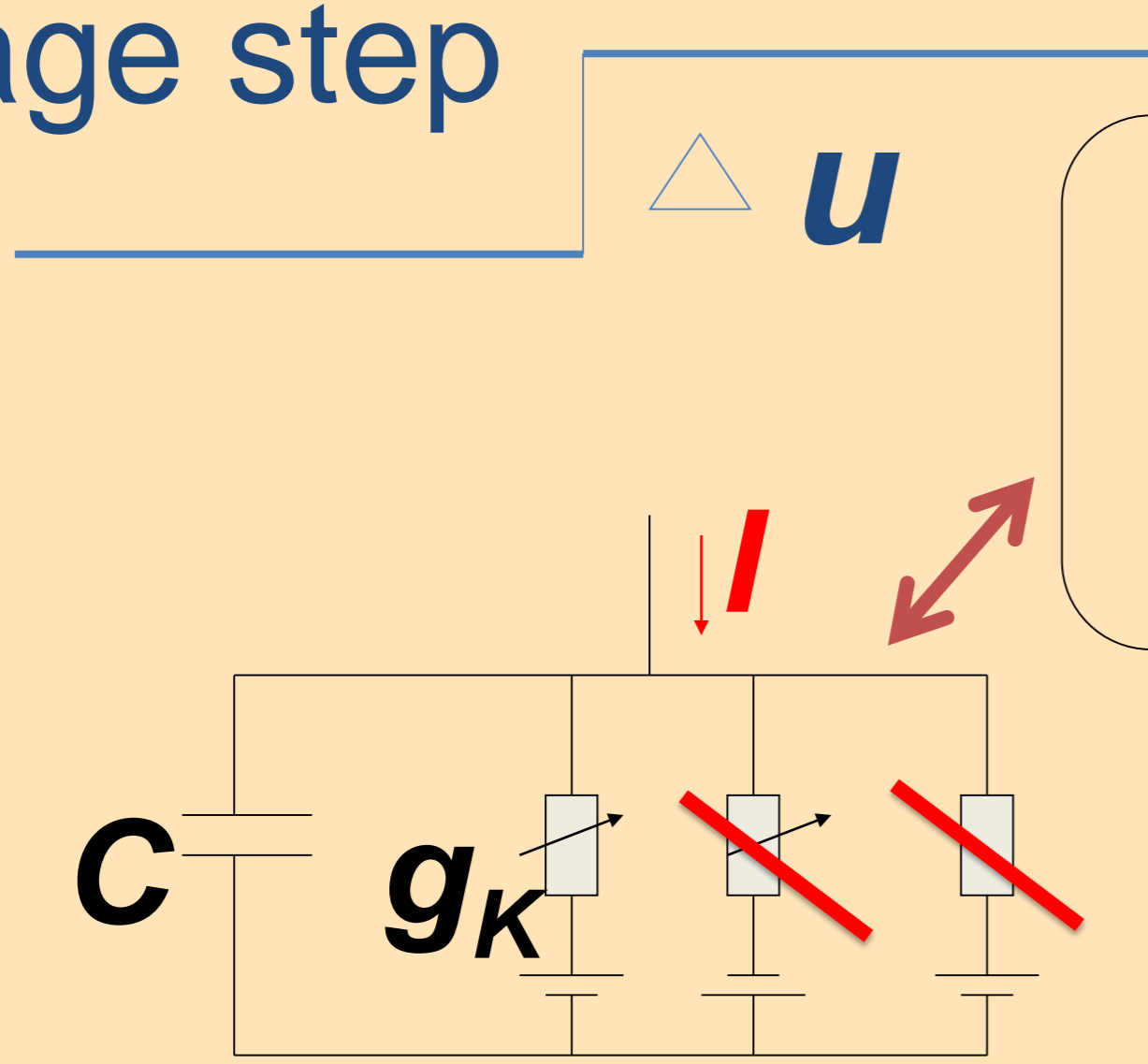
Giant axon of the squid

→ cortical neurons

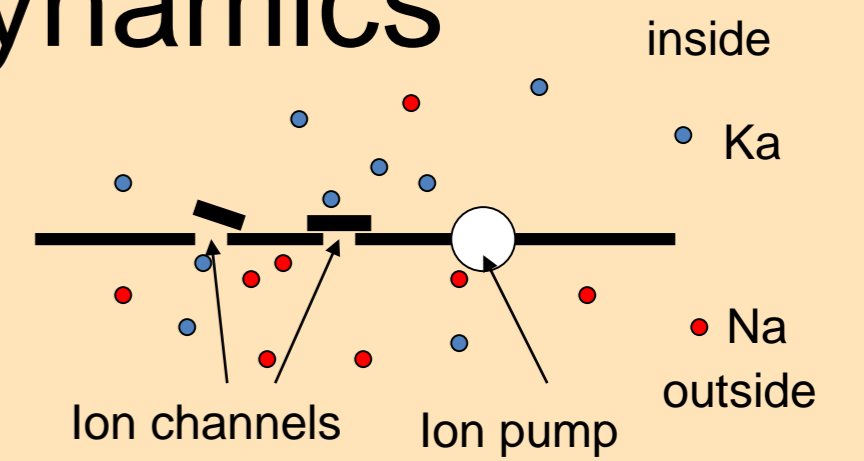
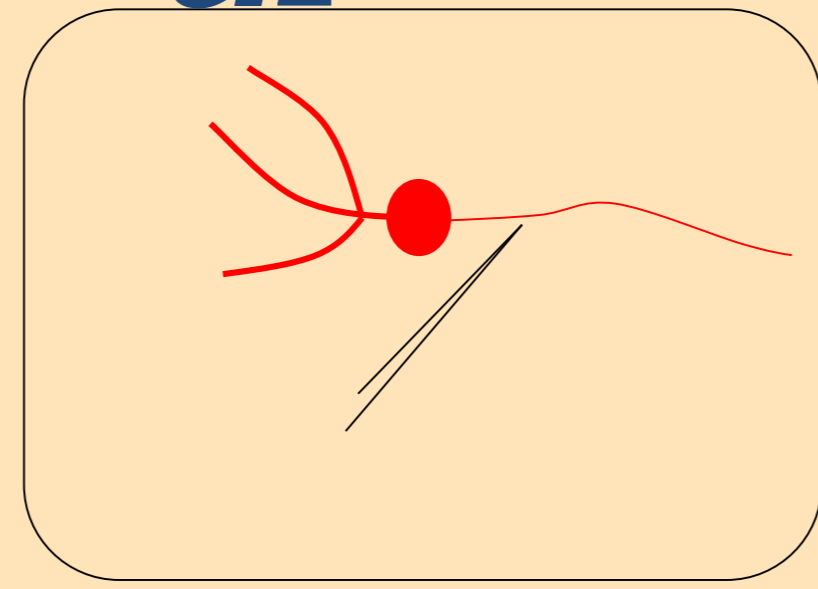
- Change of parameters
- More ion channels
- Same framework

# Exercise 3.1-3.3 – Hodgkin-Huxley – ion channel dynamics

voltage step



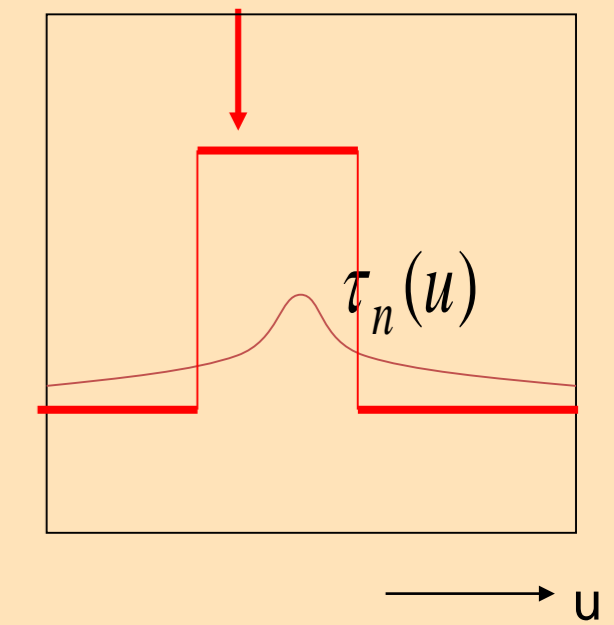
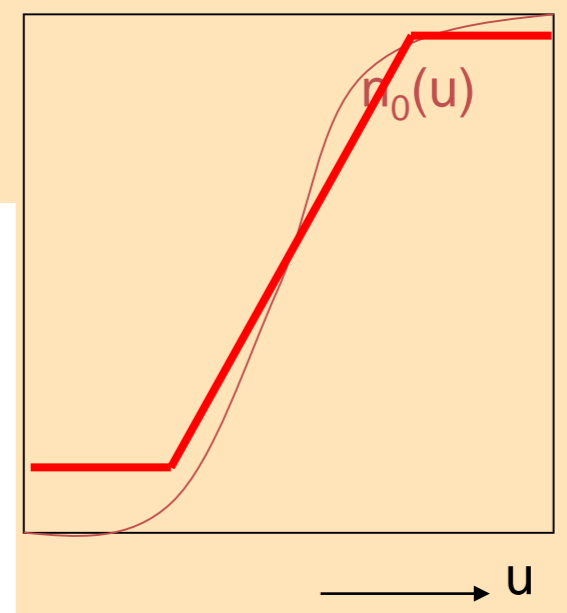
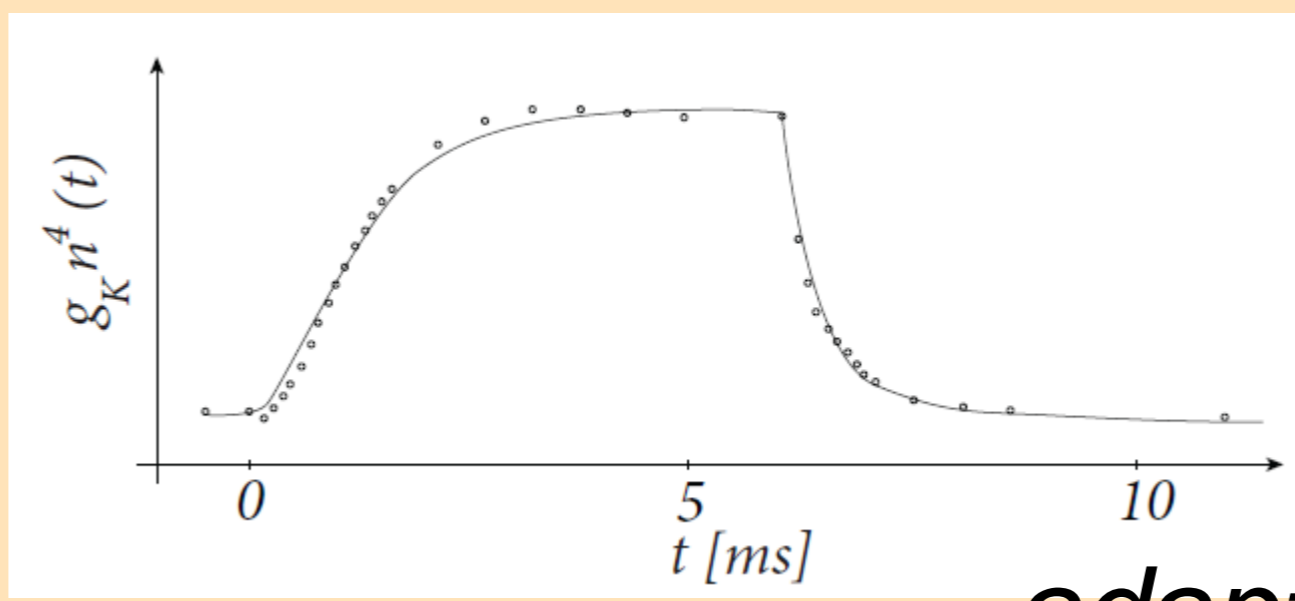
$U_2$  Determine ion channel dynamics



$$C \frac{du}{dt} = -g_K n^4 (u - E_K) + I(t)$$

stimulus

apply voltage step



Start Exercise 3 at 11:33  
Next Lecture at:  
11.48

*adapted from  
Hodgkin&Huxley 1952*

# Week 2 – part 5: Detailed Biophysical Models



## Biological Modeling of Neural Networks

### Week 2 – Biophysical modeling: The Hodgkin-Huxley model

Wulfram Gerstner

EPFL, Lausanne, Switzerland

#### ✓ 2.1 Biophysics of neurons

- Overview

#### ✓ 2.2 Reversal potential

- Nernst equation

#### ✓ 2.3 Hodgkin-Huxley Model

#### ✓ 2.4 Threshold in the Hodgkin-Huxley Model

- where is the firing threshold?

#### 2.5. Detailed biophysical models

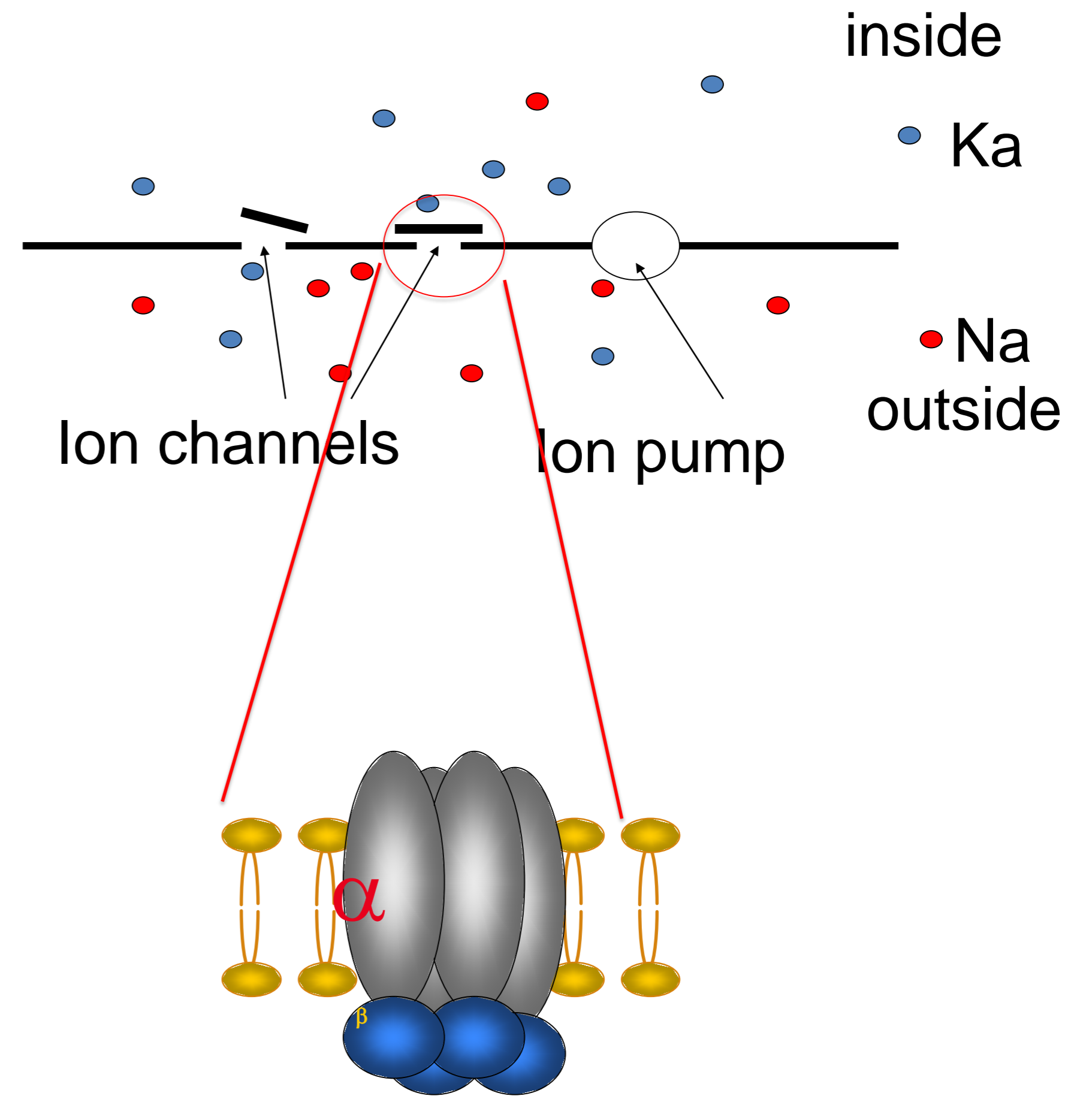
- the zoo of ion channels

# Neuronal Dynamics – 2.5 Biophysical models

*There are about 200  
identified ion channels*

<http://channelpedia.epfl.ch/>

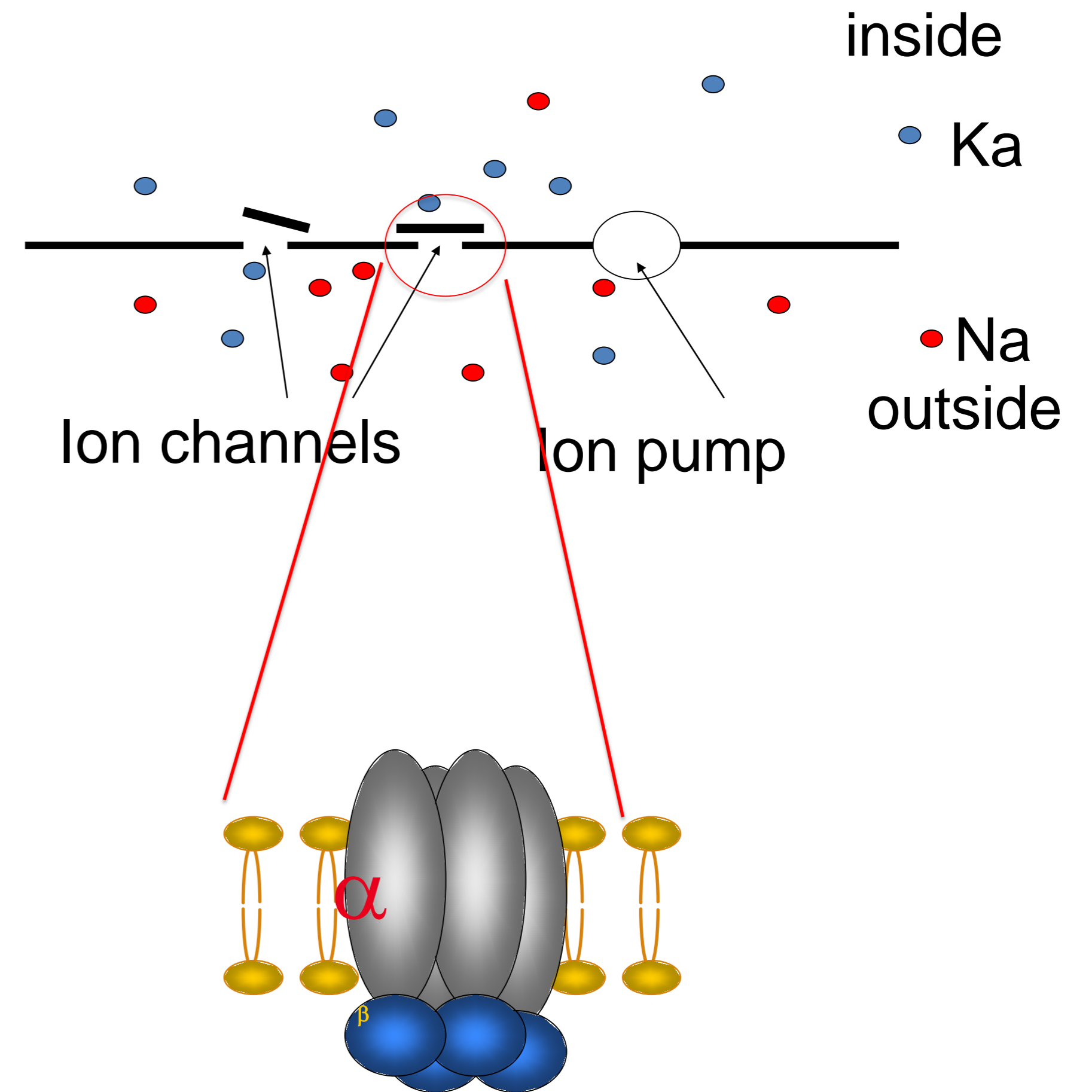
Hodgkin-Huxley model  
Provides flexible framework



# Neuronal Dynamics – 2.5 Biophysical models

Individual ion channels can be measured.

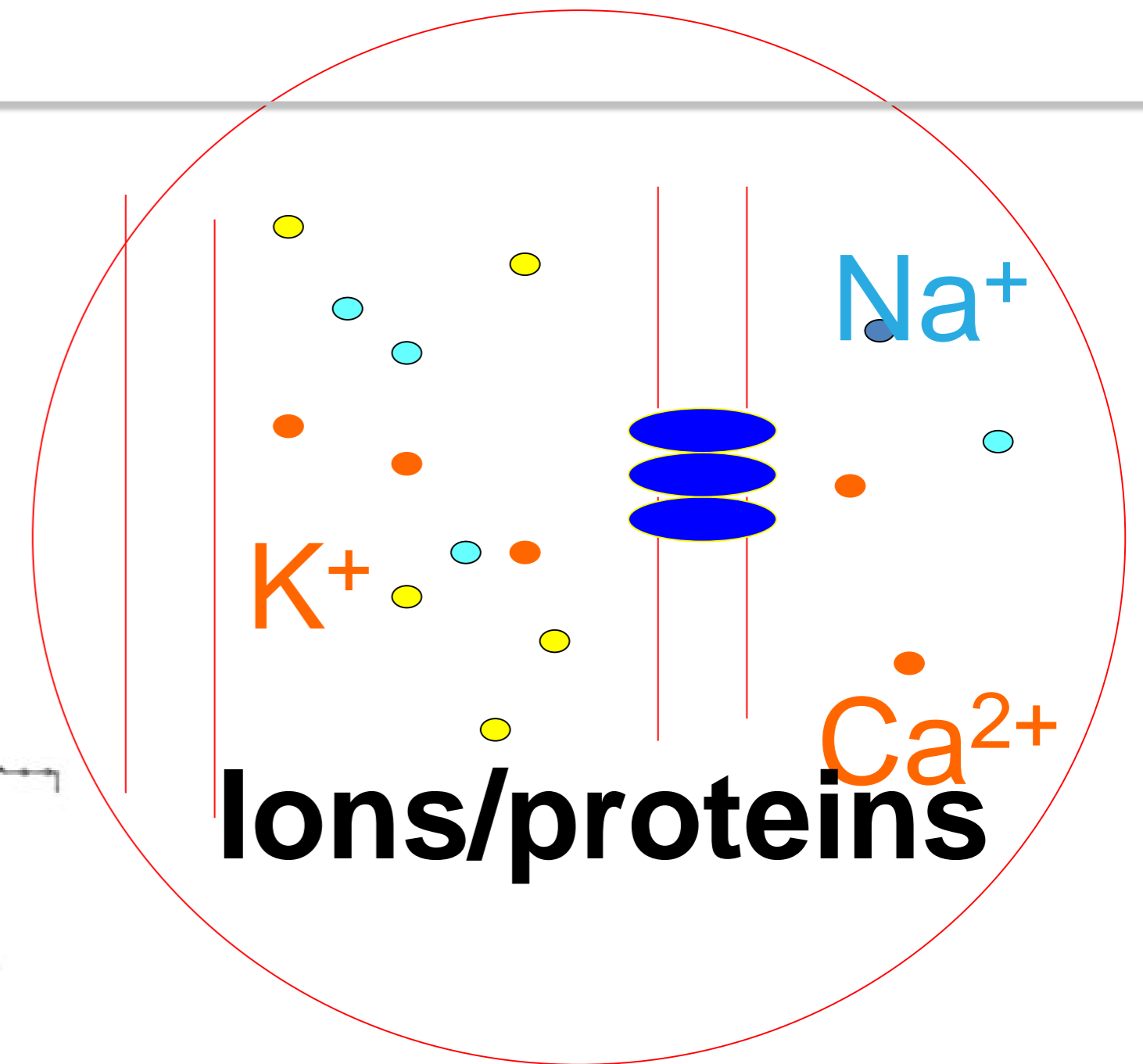
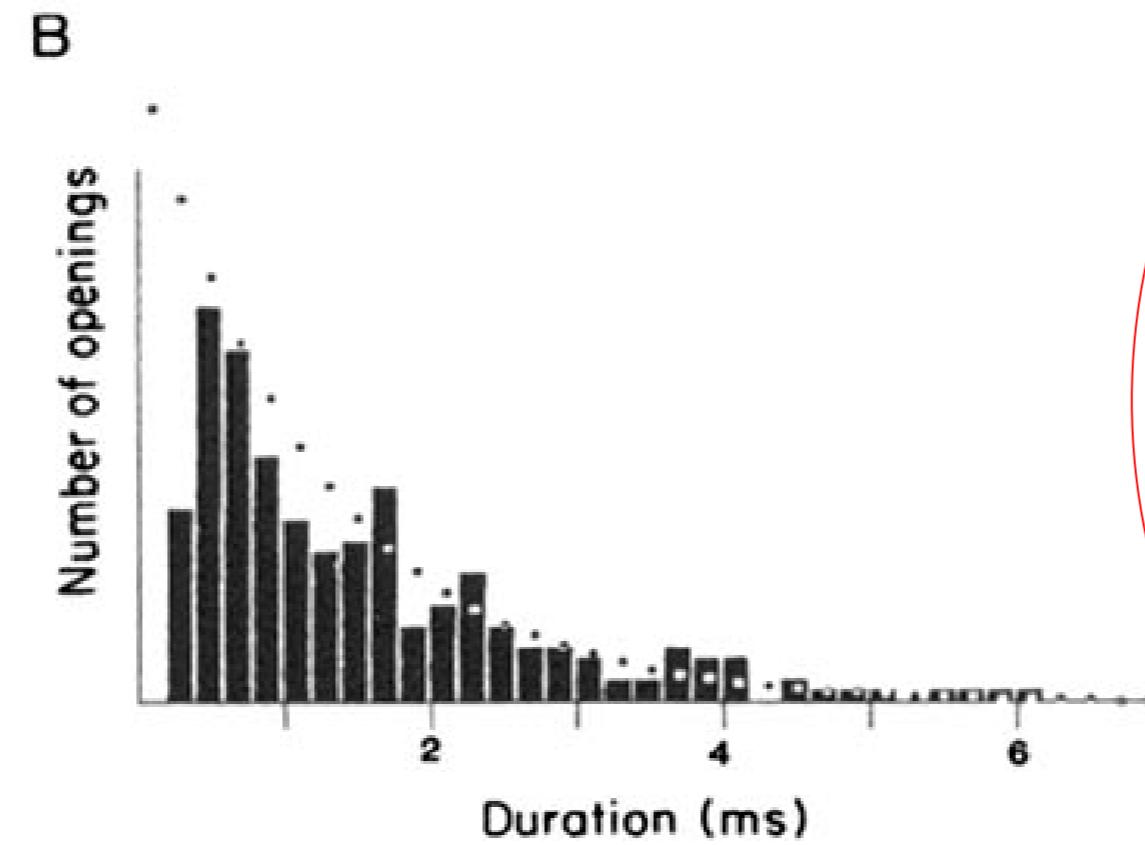
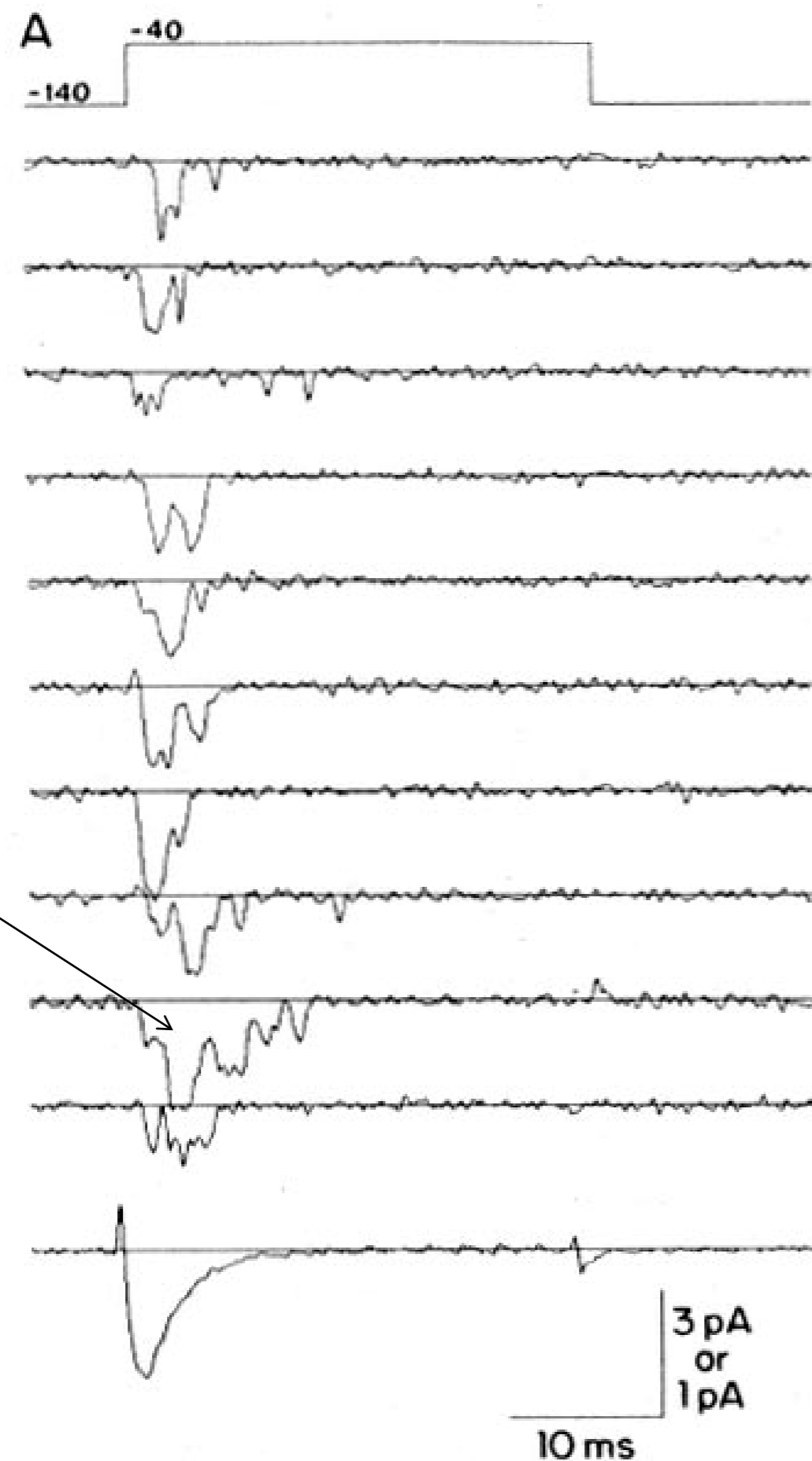
Opening and closing is stochastic





# Neuronal Dynamics – 2.5 Ion channels

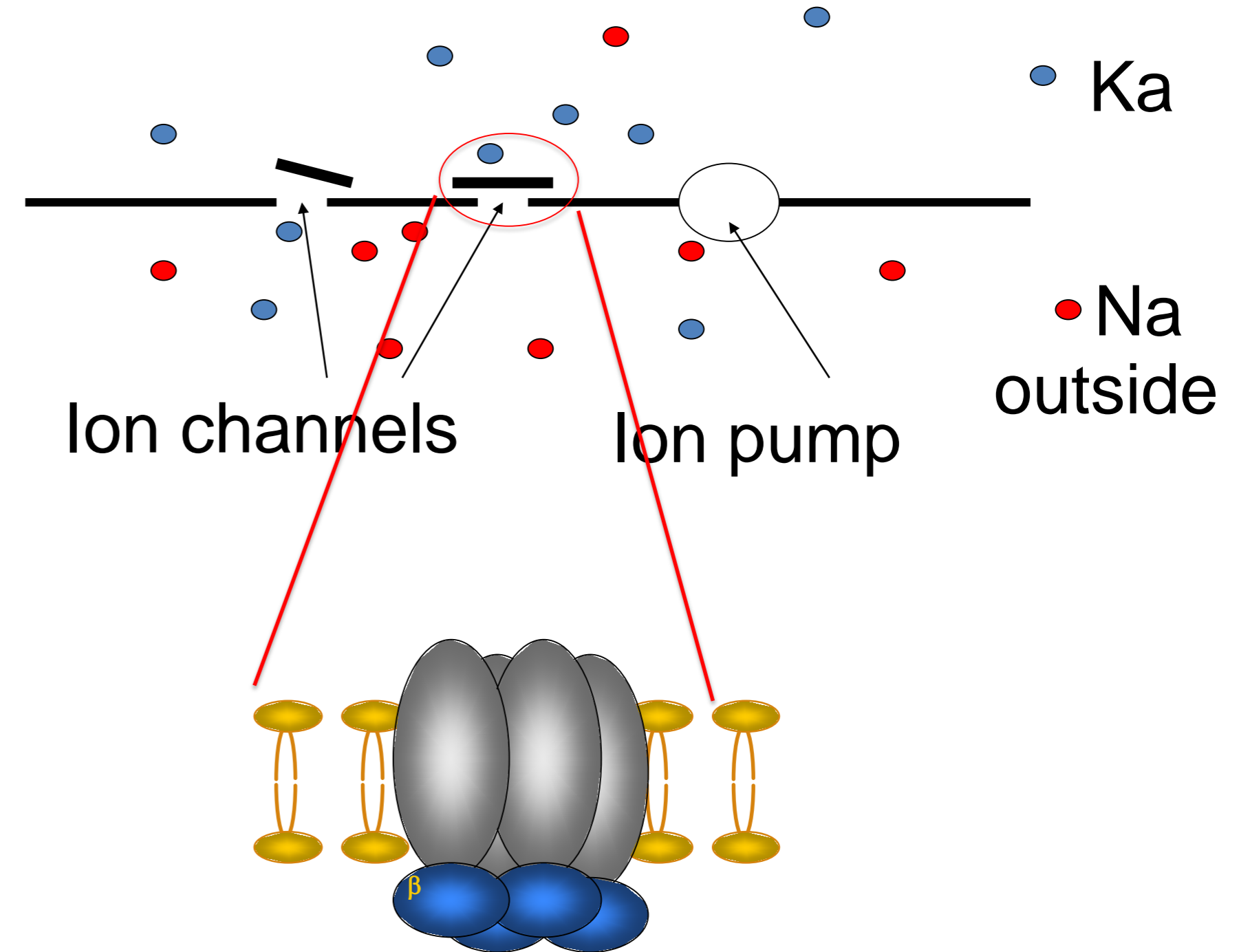
Steps:  
Different number  
of channels



Na<sup>+</sup> channel from rat heart (*Patlak and Ortiz 1985*)  
**A** traces from a patch containing several channels.  
Bottom: average gives current time course.  
**B**. Opening times of single channel events

# Neuronal Dynamics – 2.5 Biophysical models

Hodgkin-Huxley:  
-Cambridge lab  
-Plymouth lab



Hodgkin-Huxley model  
provides flexible framework

*Hodgkin&Huxley (1952)*  
*Nobel Prize 1963*

# Exercise 4 – Hodgkin-Huxley model – gating dynamics

A) Often the gating dynamics is formulated as

$$\frac{dm}{dt} = \alpha_m(u)(1-m) - \beta_m(u)m \qquad \frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m(u)}$$

Calculate  $m_0(u)$  and  $\tau_m(u)$

B) Assume a form  $\alpha_m(u) = \beta_m(u) = \frac{1}{1 - \exp[-(u + a) / b]}$

How are  $a$  and  $b$  related to  $\gamma$  and  $\theta$  in the equations

$$\frac{dm}{dt} = -\frac{m - m_0(u)}{\tau_m(u)}$$



$$m_0(u) = 0.5\{1 + \tanh[\gamma(u - \theta)]\}$$

C) What is the time constant  $\tau_m(u)$  ?

# Biological Modeling of Neural Networks

TA in 2019:

*Chiara Gastaldi*

*Martin Barry*

*Noé Gallice*

Now Computer Exercises:

Play with Hodgkin-Huxley model

*The End*

# Week 2 – References and Suggested Reading

**Reading:** W. Gerstner, W.M. Kistler, R. Naud and L. Paninski, *Neuronal Dynamics: from single neurons to networks and models of cognition*. Chapter 2: *The Hodgkin-Huxley Model*, Cambridge Univ. Press, 2014

- Hodgkin, A. L. and Huxley, A. F. (1952). *A quantitative description of membrane current and its application to conduction and excitation in nerve*. J Physiol, 117(4):500-544.
- Ranjan, R., et al. (2011). *Channelpedia: an integrative and interactive database for ion channels*. Front Neuroinform, 5:36.
- Toledo-Rodriguez, M., Blumenfeld, B., Wu, C., Luo, J., Attali, B., Goodman, P., and Markram, H. (2004). *Correlation maps allow neuronal electrical properties to be predicted from single-cell gene expression profiles in rat neocortex*. Cerebral Cortex, 14:1310-1327.
- Yamada, W. M., Koch, C., and Adams, P. R. (1989). *Multiple channels and calcium dynamics*. In Koch, C. and Segev, I., editors, *Methods in neuronal modeling*, MIT Press.
- Aracri, P., et al. (2006). *Layer-specific properties of the persistent sodium current in sensorimotor cortex*. Journal of Neurophysiol., 95(6):3460-3468.