

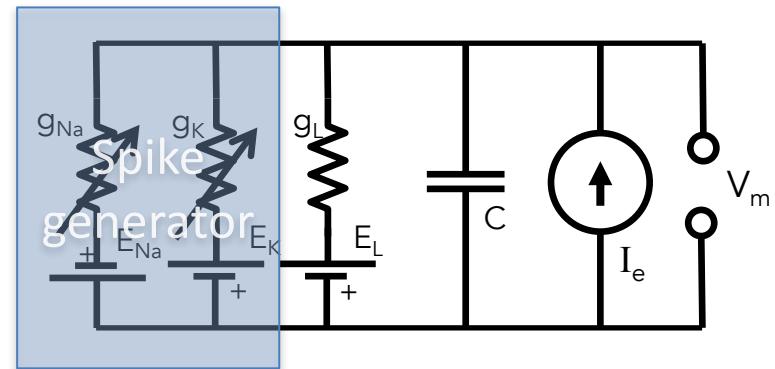
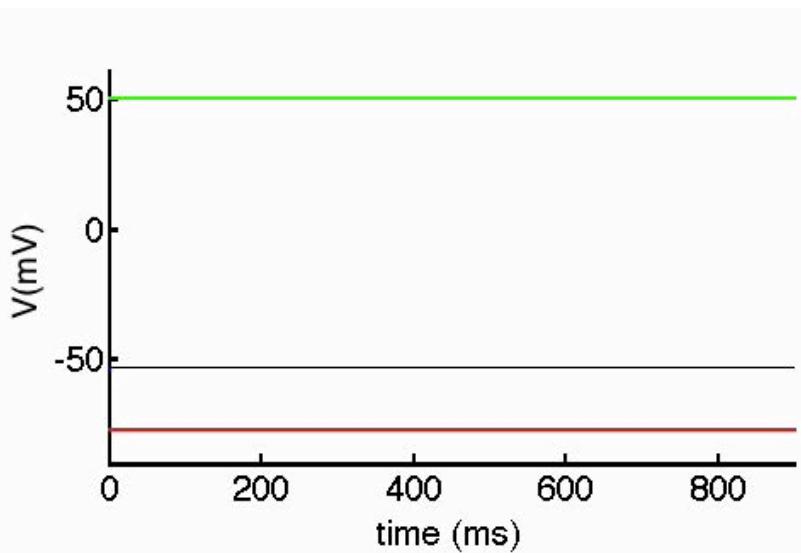
# Introduction to Neural Computation

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Prof. Michale Fee  
MIT BCS 9.40 — 2018  
Lecture 4

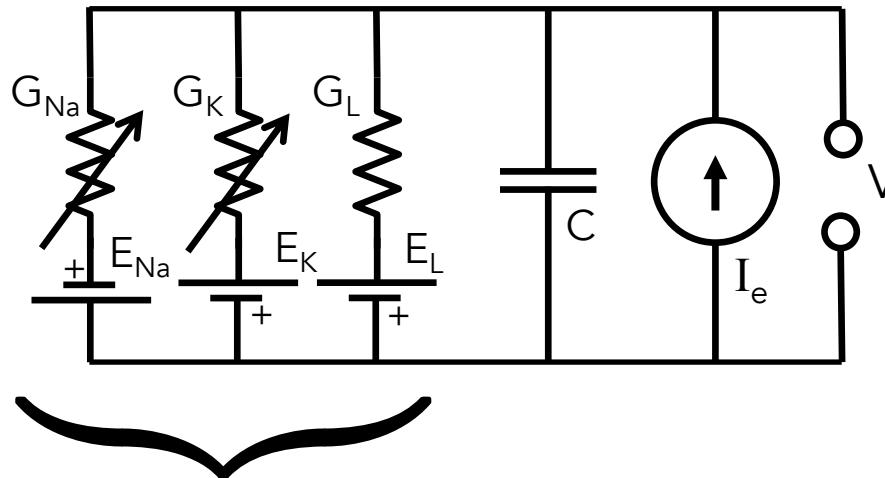
# Hodgkin-Huxley model of action potential generation

Voltage and time-dependent ion channels are the 'knobs' that control membrane potential.



Removed due to copyright restrictions: Figure 1a: The first intracellular recording of an action potential, from squid axon. Häusser, M. "The Hodgkin-Huxley theory of the action potential." *Nature Neuroscience* 3 (2000).

# Hodgkin-Huxley model of action potential generation



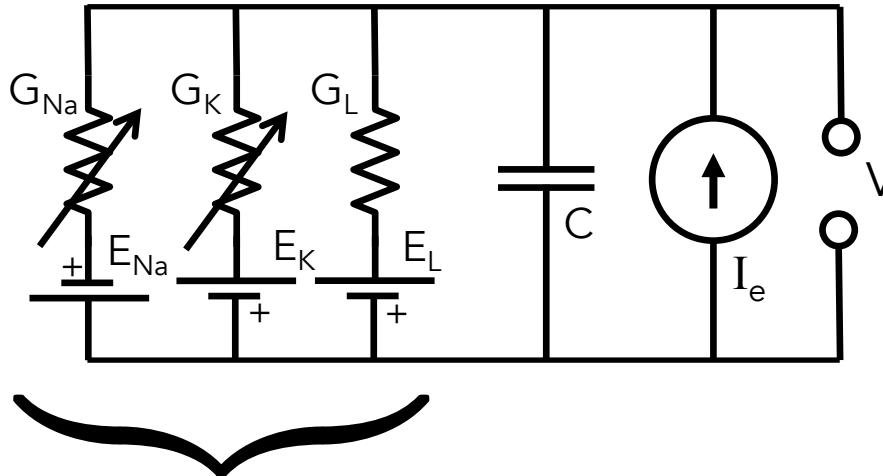
$$I_m = I_{Na} + I_K + I_L$$

This is the total membrane ionic current, and it includes the contribution from —sodium channels, potassium channels and a 'leak' conductance.

The equation for our HH model neuron is

$$I_m(t) + C \frac{dV(t)}{dt} = I_e(t)$$

# Hodgkin-Huxley model of action potential generation



$$I_m = I_{Na} + I_K + I_L \quad I_m(t) + C \frac{dV(t)}{dt} = I_e(t)$$

$$I_{Na} = G_{Na}(V,t)(V - E_{Na}) \quad I_K = G_K(V,t)(V - E_K) \quad I_L = G_L(V - E_L)$$

We can see that the membrane potential depends on current  
...which depends on all the conductances  
...which depend on the membrane potential

# We are going to write down an algorithm for how a neuron spikes!

Start with  $V_m$  :

→ Compute voltage-dependent parameters using  $V_m$

Compute conductance using voltage-dependent parameters

Compute sodium and potassium current from conductances

Compute total membrane current

Compute  $V_\infty$  and  $\tau_{mem}$

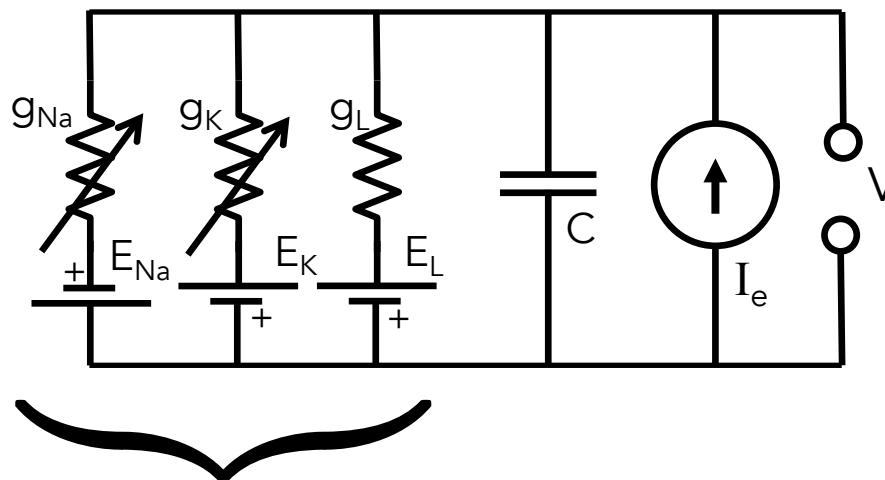
Compute membrane potential

$$V_m + \tau \frac{dV_m}{dt} = V_\infty$$

# Learning objectives for Lecture 2

- To be able to draw the circuit diagram of the HH model
- Understand what a voltage clamp is and how it works
- Be able to plot the voltage and time dependence of the potassium current and conductance
- Be able to explain the time and voltage dependence of the potassium conductance in terms of Hodgkin-Huxley gating variables

# Outline of HH model



$$I_m = I_{Na} + I_K + I_L$$

$$I_X = G_X \cdot (V - E_X)$$

$$I_{Na} = G_{Na}(V,t)(V - E_{Na})$$

$$I_K = G_K(V,t)(V - E_K)$$

$$I_L = G_L(V - E_L)$$

The sodium conductance is time-dependent and voltage-dependent

The potassium conductance is time-dependent and voltage-dependent

The leak conductance is neither time-dependent nor voltage-dependent

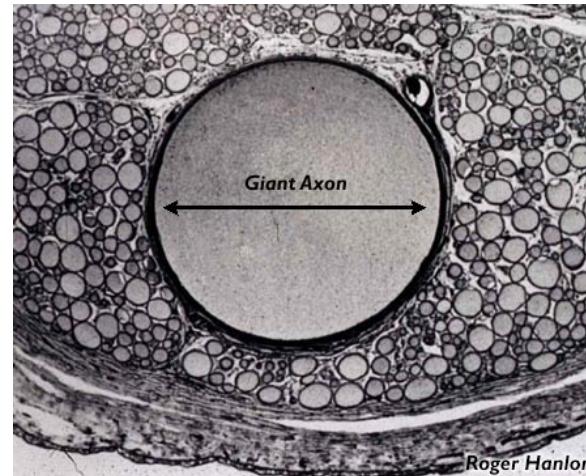
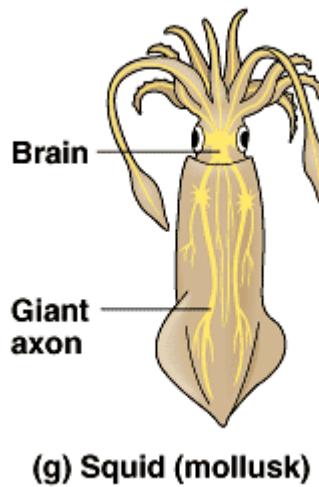
$$E_{Na} = +55mV$$

$$E_K = -75mV$$

$$E_L = -50mV$$

# Voltage and Time dependence

- Voltage and time-dependent ion channel **conductances** are the 'knobs' that control membrane potential.
- H&H studied the properties of K and Na channels in the squid giant axon. In particular they wanted to study the voltage and time dependence of the K and Na channels.



1mm diameter!

Squid diagram from [The CellularScale](#).  
License CC BY-NC-SA.

Image of squid giant axon © Kay Cooper and Roger Hanlon. Used with permission.

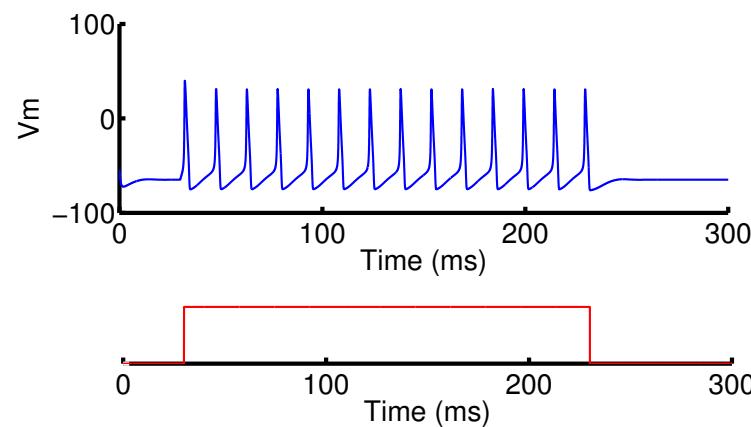
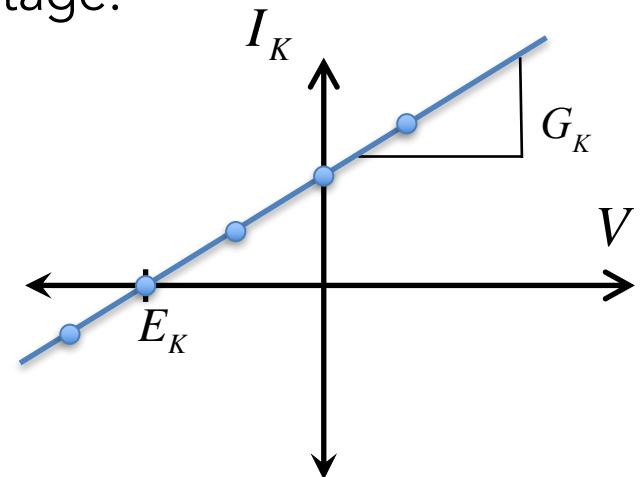
Most axons in our brain are 1um dia

# Outline of HH model

The best way study the time and voltage dependent **conductance** of ionic channels is to suddenly 'set' the voltage at different values and measure the current required to hold that voltage.

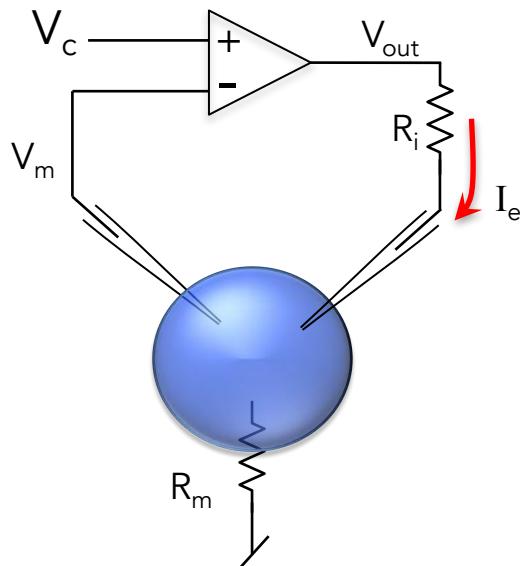
Then plot the I-V curve.

This not easy, because as soon as you depolarize the axon, the axon begins to spike!

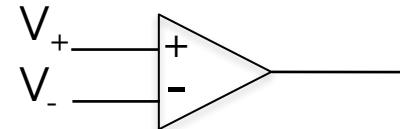


# Voltage Clamp

A voltage clamp is a device that holds the membrane potential of a cell to any desired 'command' voltage  $V_c$ , and measures the current required to hold that voltage.



The key component is an operational amplifier (op-amp)



The basic equation of an op-amp is:

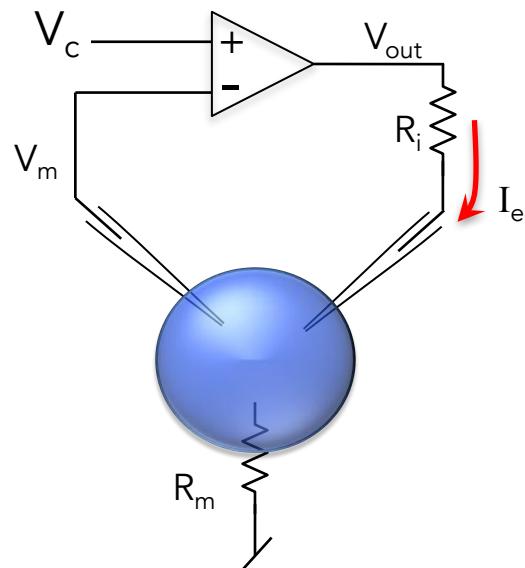
$$V_{out} = G(V_+ - V_-)$$

where  $G$  is the gain, typically  $\sim 10^5$  or  $10^6$ ,

Don't get confused here.  $G$  is gain, not conductance!

# Voltage Clamp

$$V_{out} = G(V_c - V_m)$$



$$\text{If } V_m < V_c \text{ then } V_{out} \gg 0$$

- Drives current into neuron
- Increases membrane potential

$$\text{If } V_m > V_c \text{ then } V_{out} \ll 0$$

- Pulls current out of neuron
- Decreases membrane potential

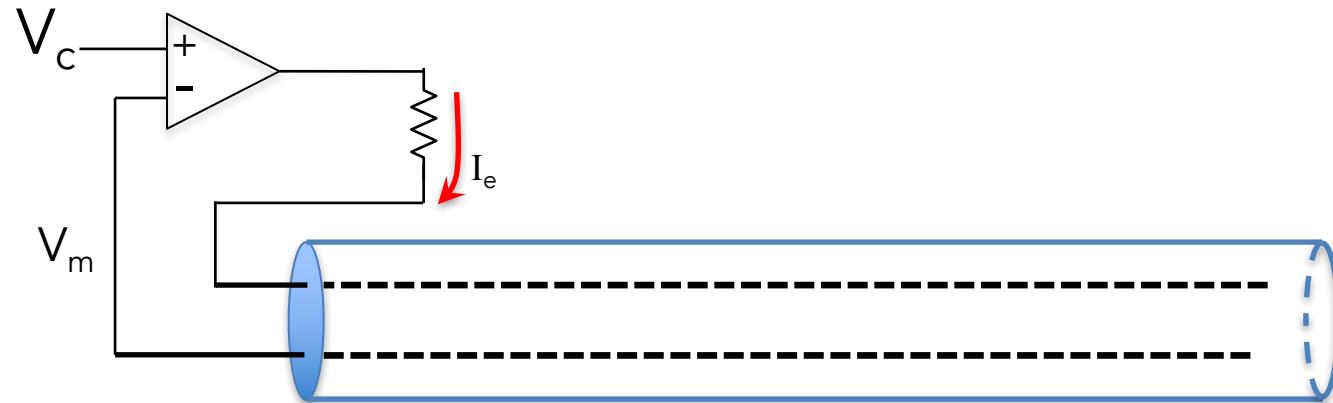
This is called 'negative feedback'.

- It is easy to show that, for large gain:

$$V_m \simeq V_c$$

- Thus, the voltage clamp circuit drives whatever current ( $I_e$ ) is necessary to 'clamp' the voltage of the neuron to the command voltage.
- During a voltage clamp experiment, we step the  $V_c$  around within the voltage range of interest and measure  $I_e$ .

# Ionic currents

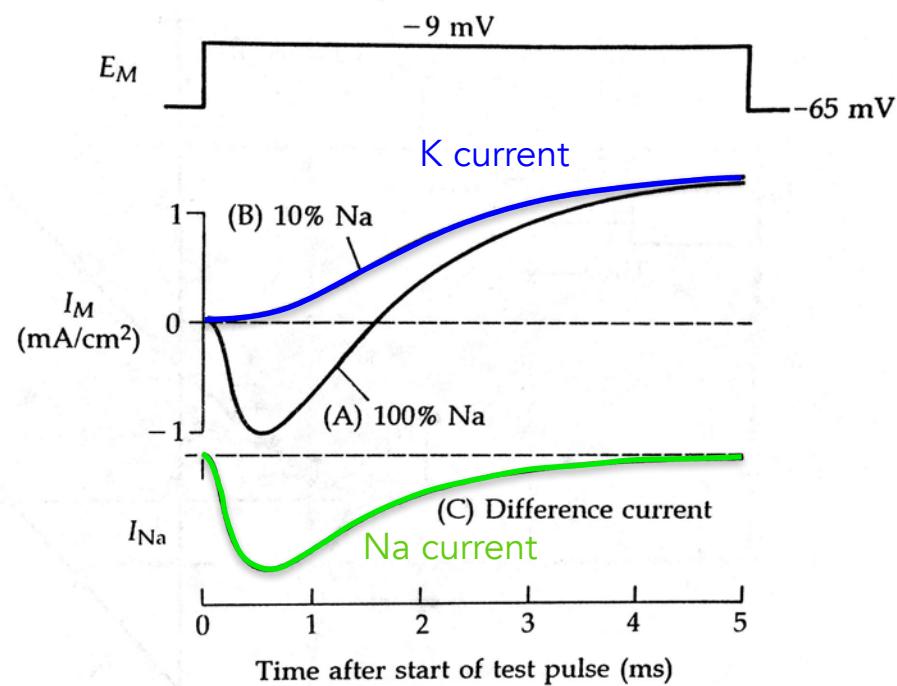


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# Ionic currents

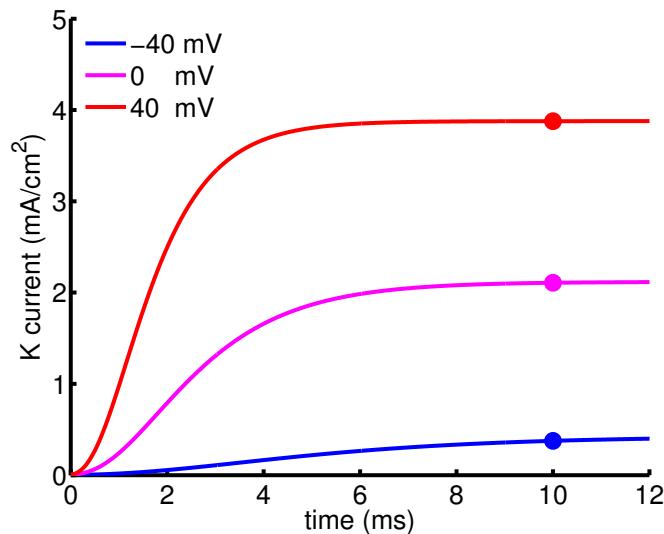
How do we figure out the contribution of Na and the contribution of K?

Ionic substitution (e.g. replace NaCl with choline chloride)

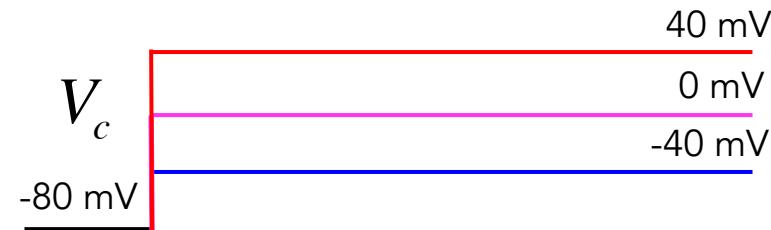
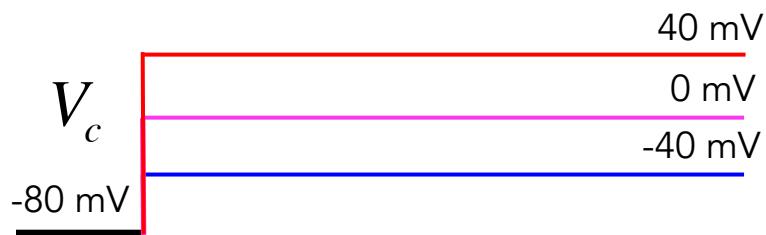
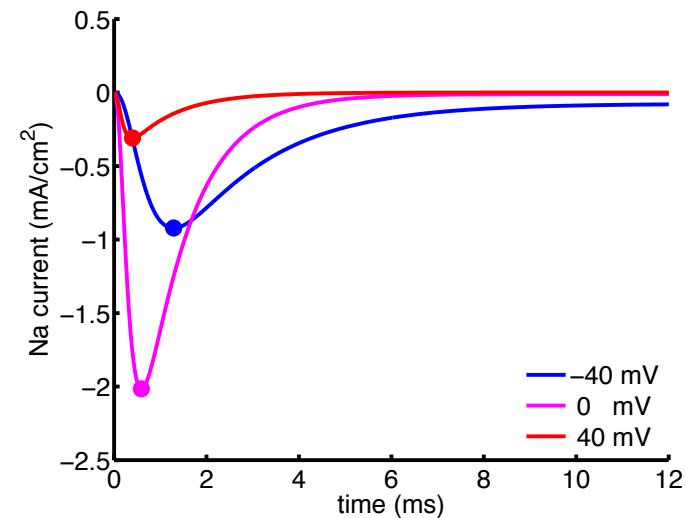


# Ionic currents

K current

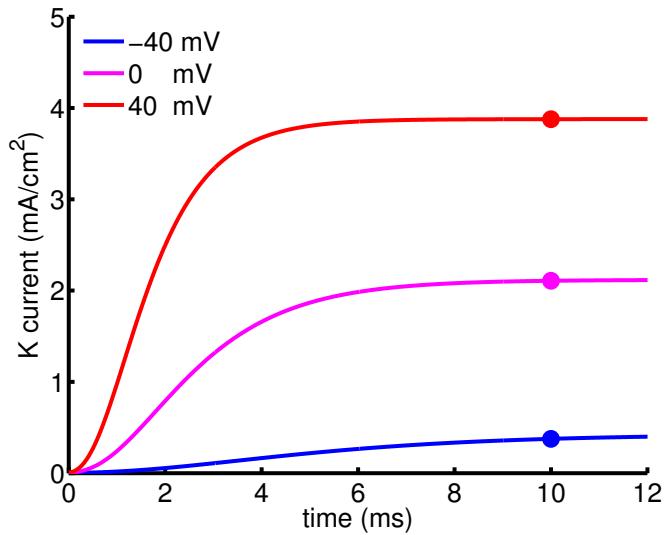


Na current

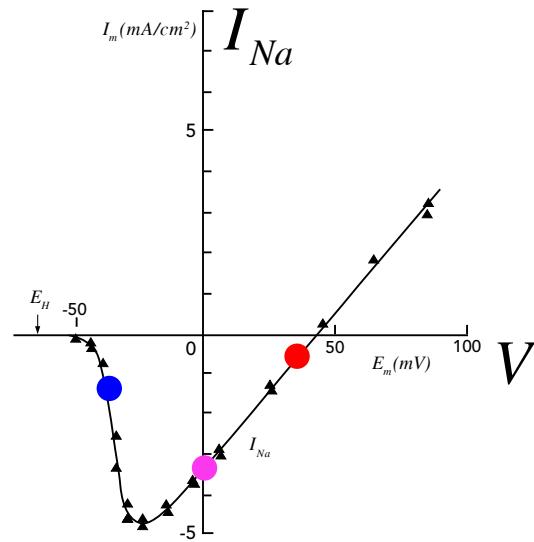
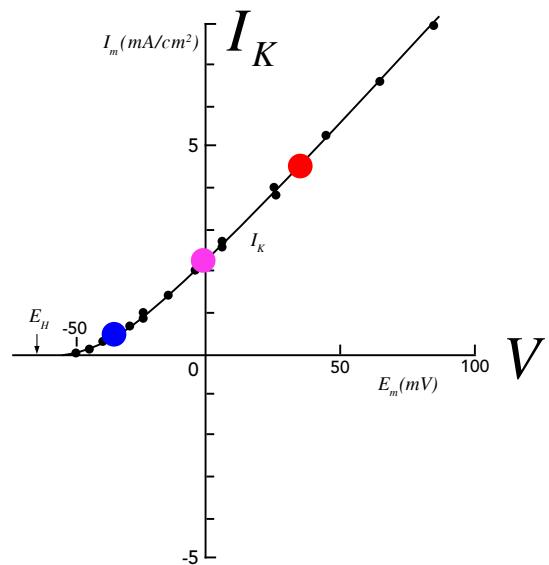
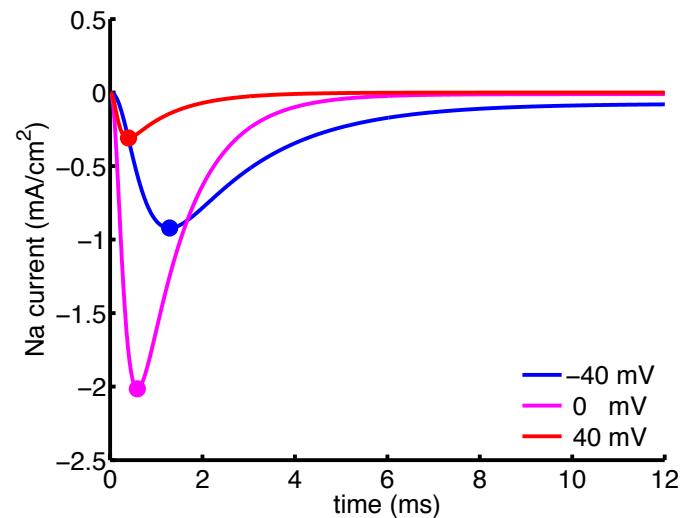


# Ionic currents

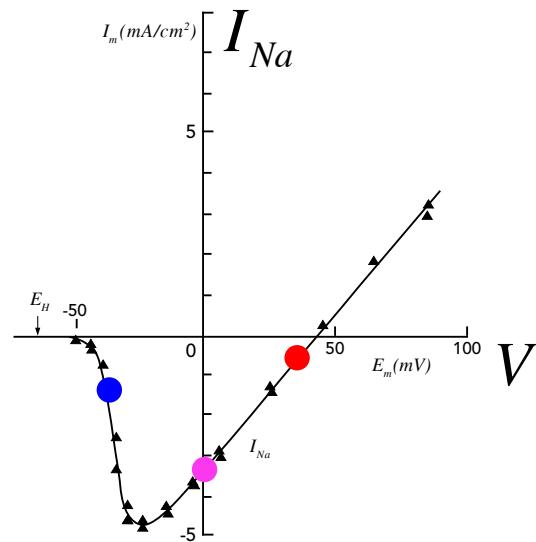
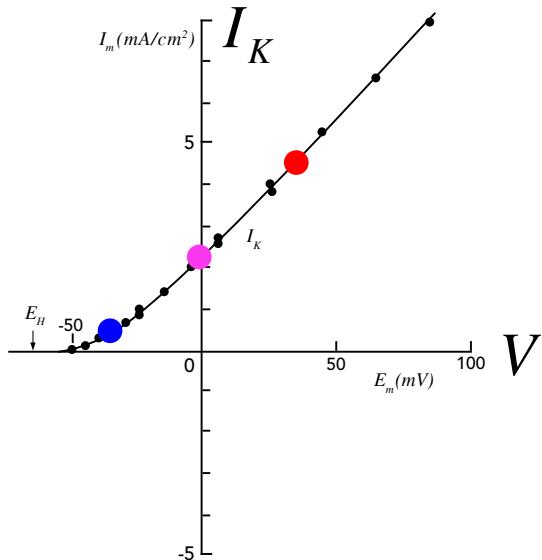
K current



Na current



# Ionic currents (voltage dependence)



We used the voltage clamp to measure **current** as a function of voltage.

But what we are really trying to extract is **conductance** as a function of voltage!

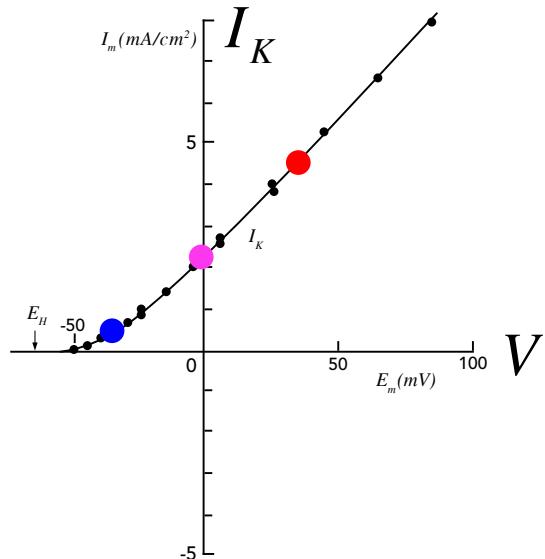
$$I_K(V) = G_K(V)(V - E_K)$$

$$I_{Na}(V) = G_{Na}(V)(V - E_{Na})$$

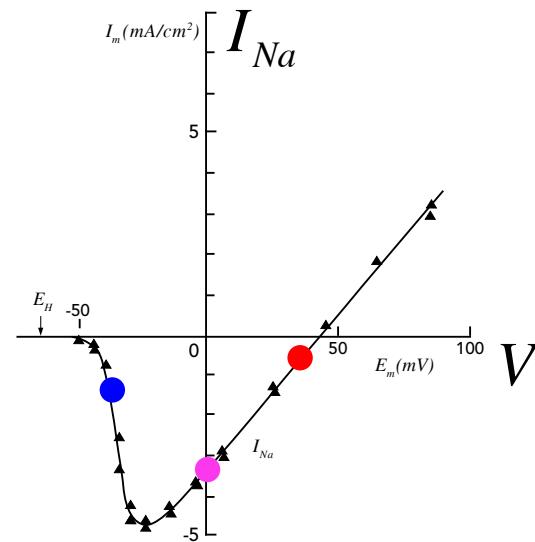
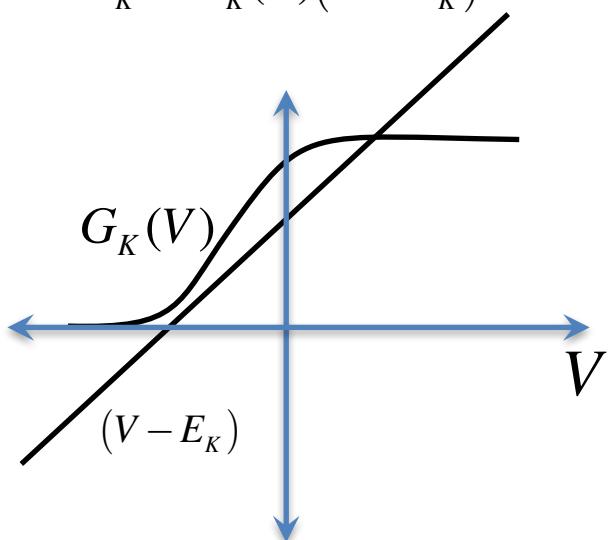
$$G_K(V) \equiv \frac{I_K(V)}{(V - E_K)}$$

$$G_{Na}(V) \equiv \frac{I_{Na}(V)}{(V - E_{Na})}$$

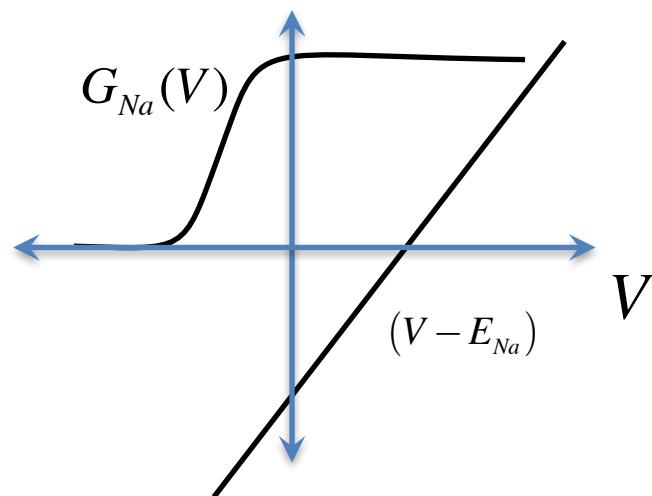
# Ionic currents (voltage dependence)



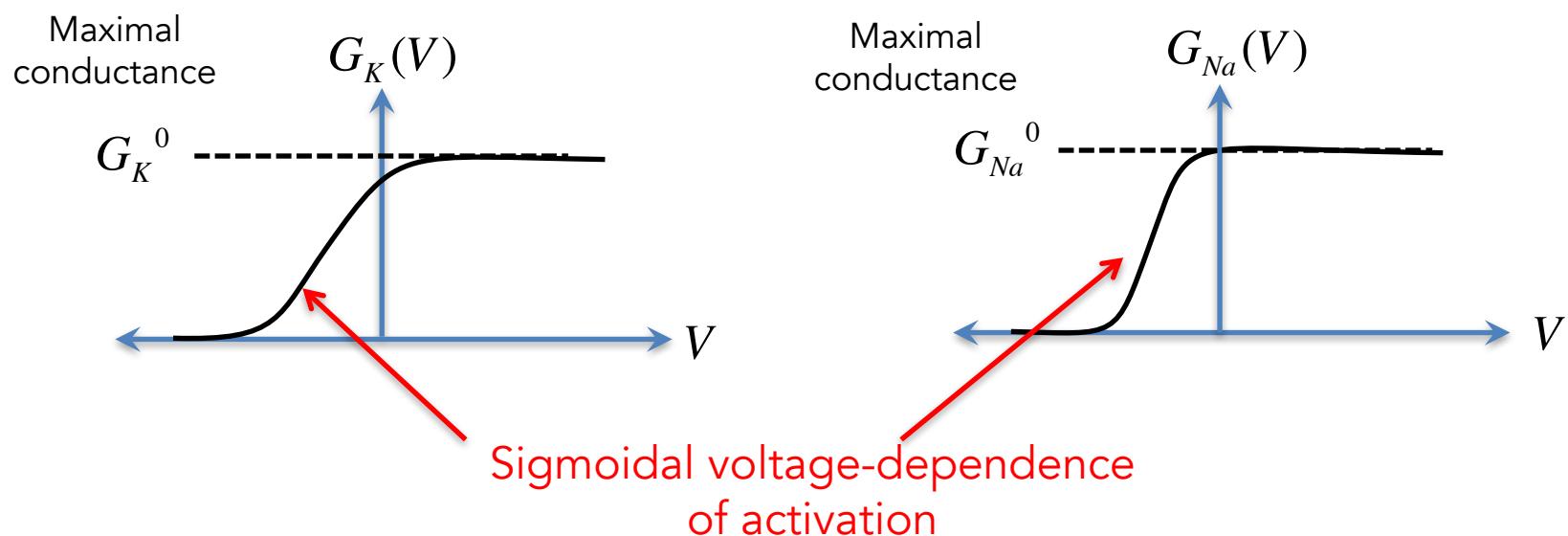
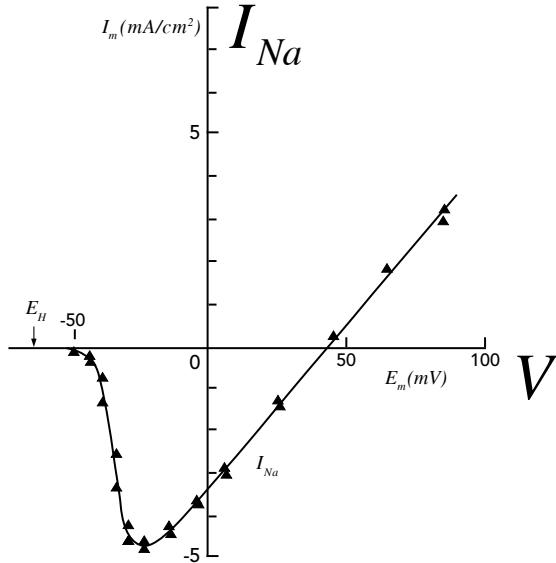
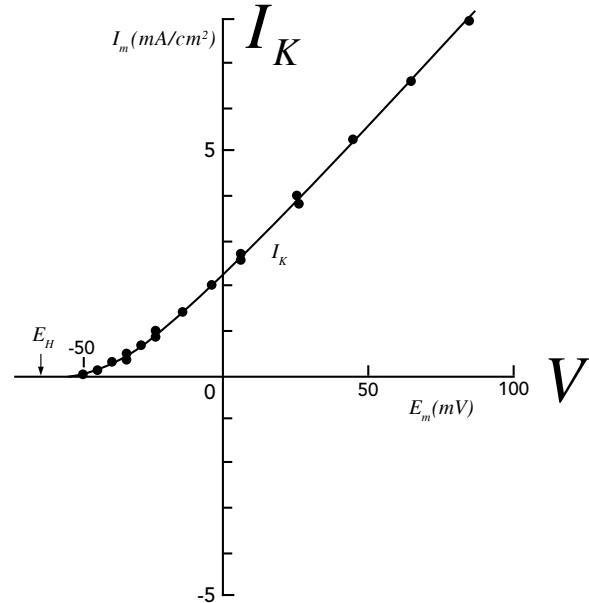
$$I_K = G_K(V)(V - E_K)$$



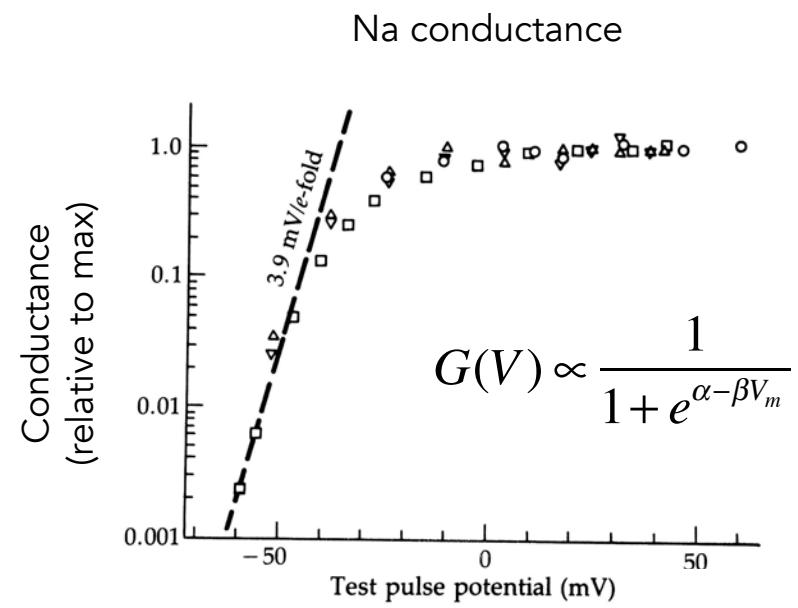
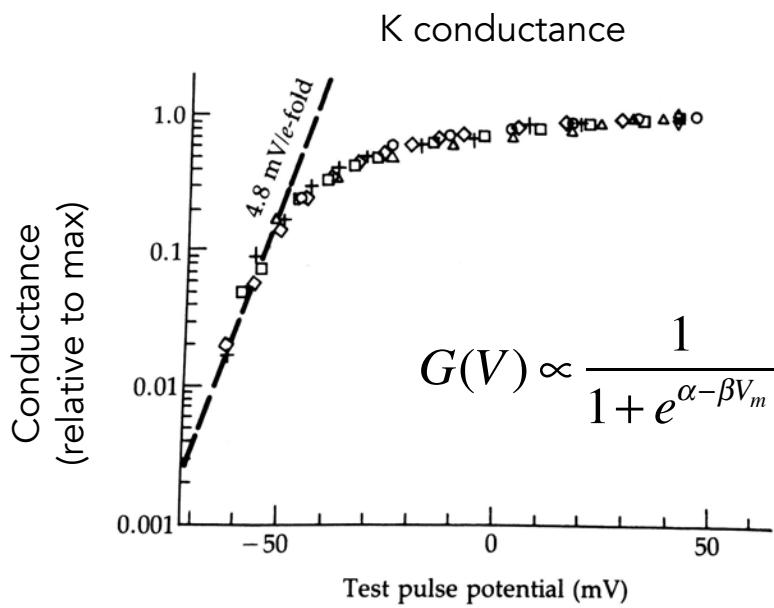
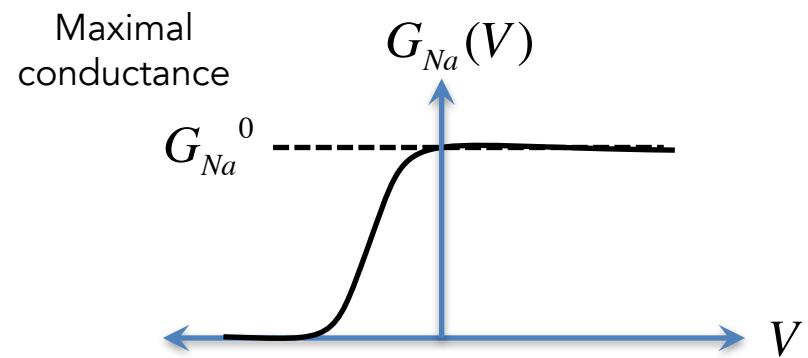
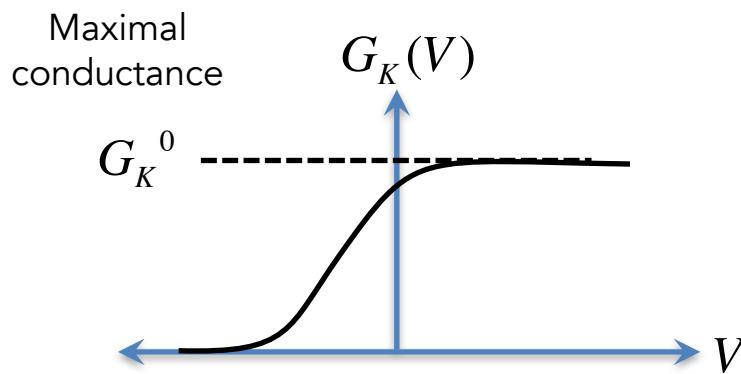
$$I_{Na} = G_{Na}(V)(V - E_{Na})$$



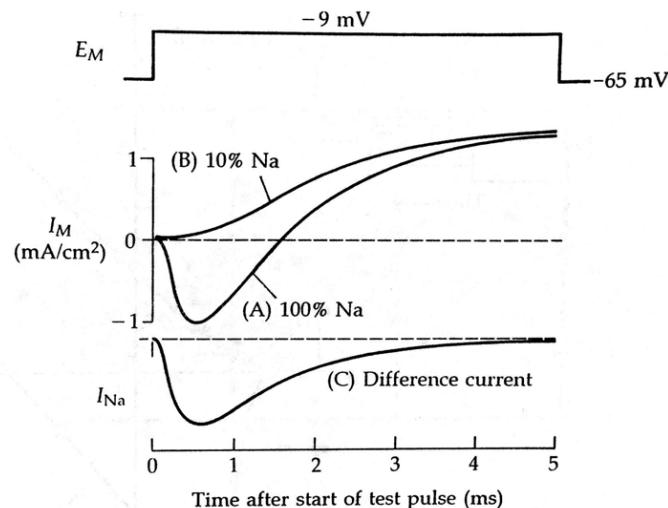
# Ionic currents (voltage dependence)



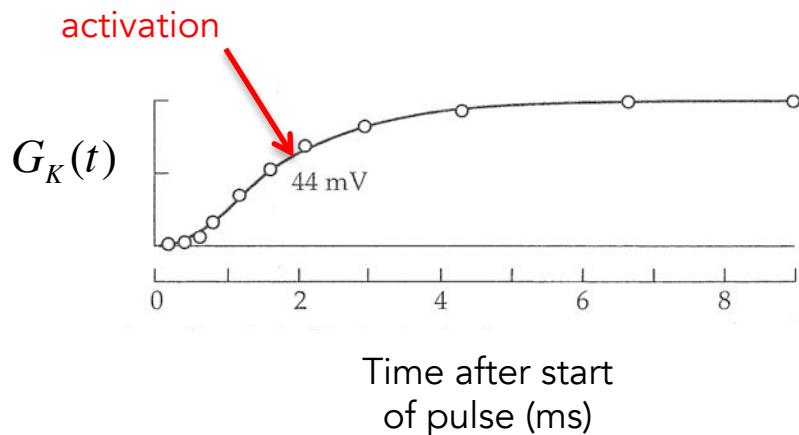
# Ionic currents (Voltage dependence)



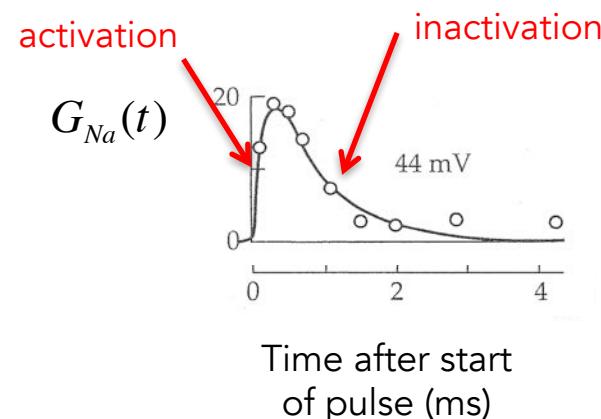
# Ionic currents (time dependence)



$$I_K(t) = G_K(t)(V - E_K)$$

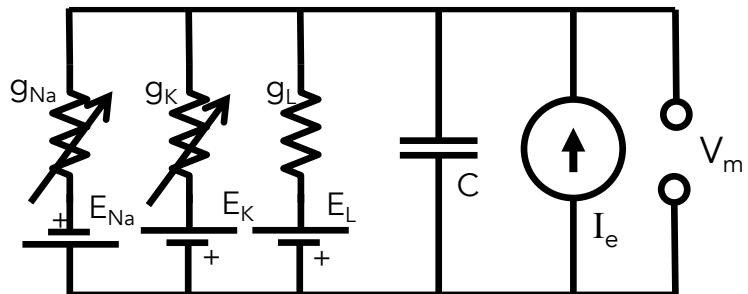
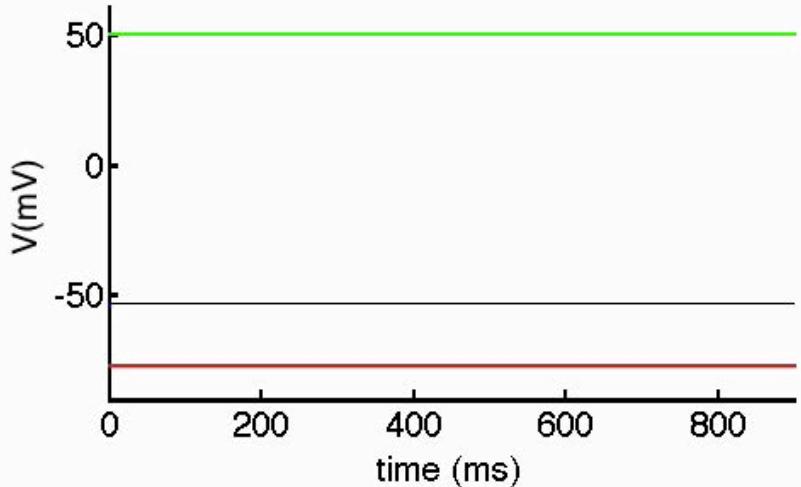


$$I_{Na}(t) = G_{Na}(t)(V - E_{Na})$$



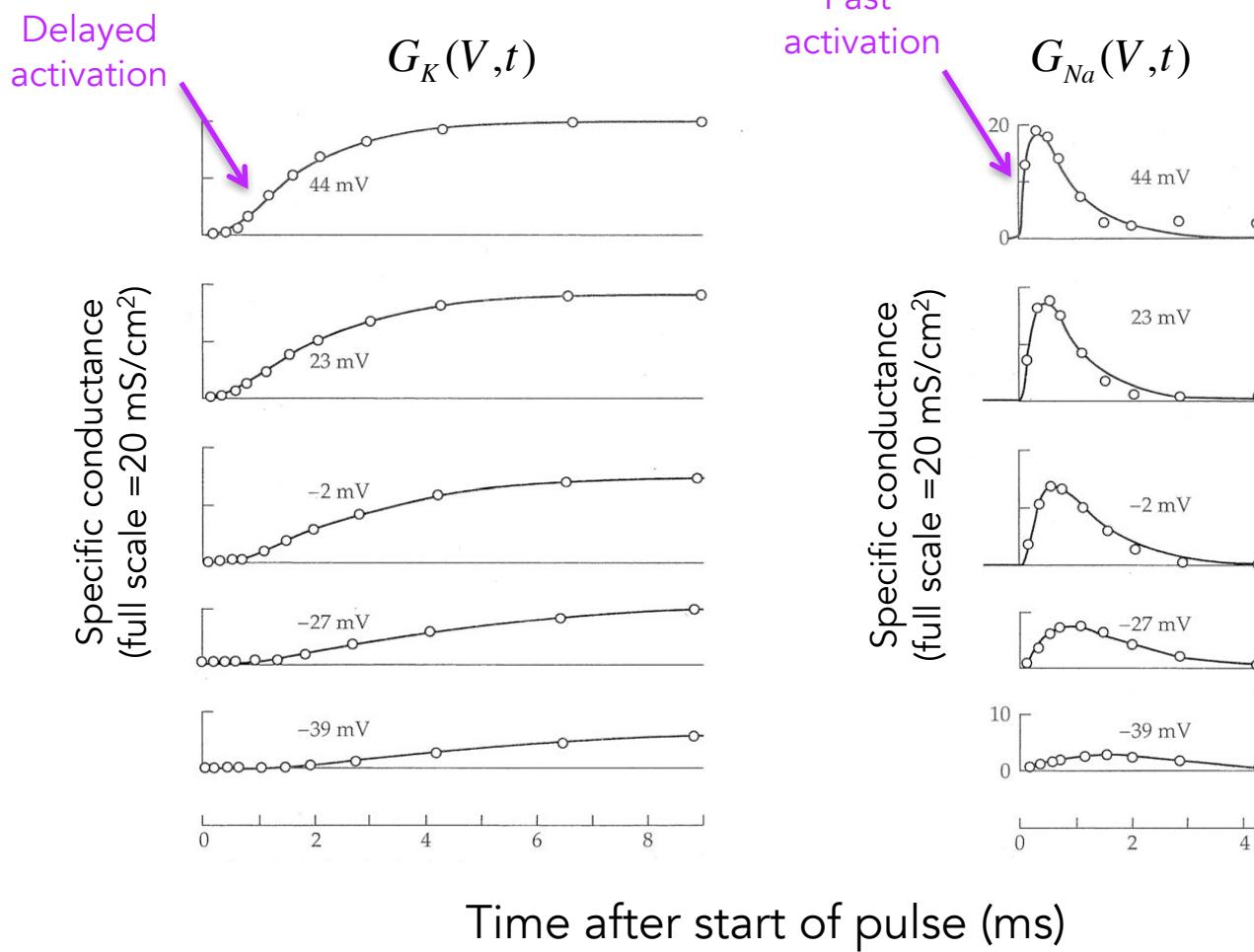
# Outline of HH model

Voltage and time-dependent ion channels are the 'knobs' that control membrane potential.



- Some ion channels push the membrane potential positive.
- Other ion channels push the membrane potential negative.
- Together these channels give the neural machinery flexible control of voltage!

# Ionic currents (time and voltage dependence)

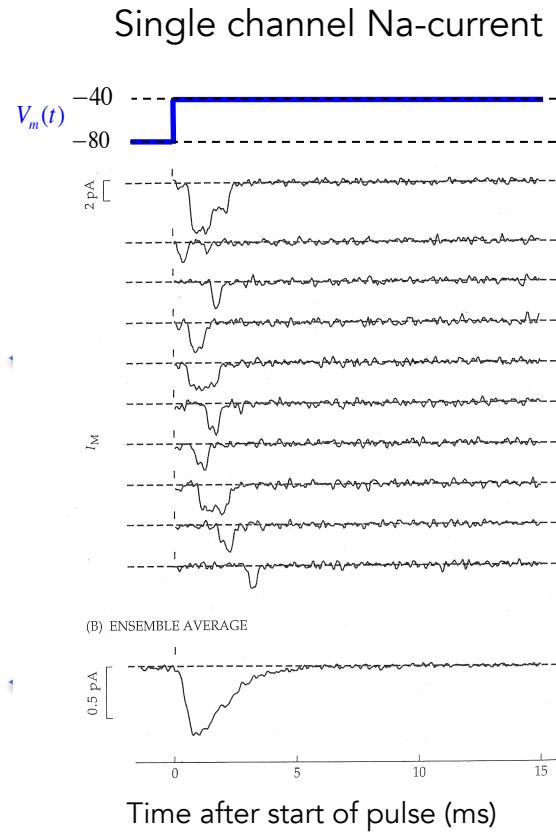
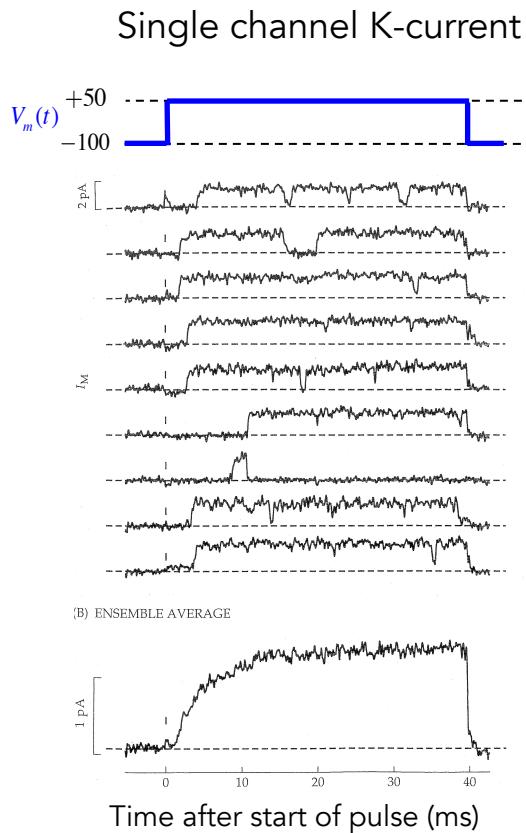


# Ionic Currents (time and voltage dependence)

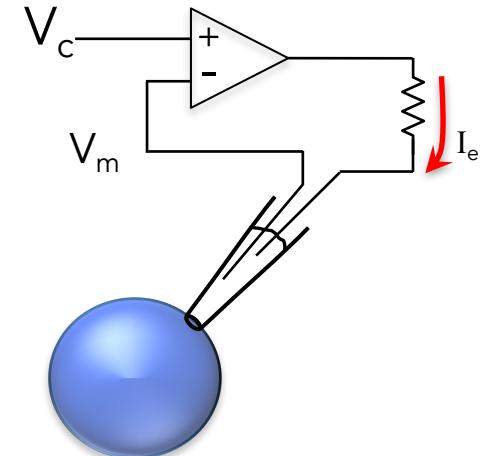
- Now we are going to develop a detailed understanding of where these voltage and time-dependencies come from!
- We are going to derive equations for the voltage dependence
- And differential equations for the time dependence.
- Once we do that, we will be able to write down a system of equations that allows us to simulate an action potential!

# Single Channels

So far, we have been discussing total currents into a neuron. However, these total currents result from ionic flow through thousands of individual ion channels.



It is possible to record from single ion channels using a 'patch clamp'.



Neher & Sakmann

# Ionic conductance in terms of single channels

Individual ion channels are either OPEN or CLOSED.

The total conductance through a membrane is given by the total number of open channels times the conductance one ion channel.

$P_K$  = probability of being 'open'

Number of 'open' ion

$N_K$  = total number of ion channels

channels =  $P_K N_K$

$\hat{g}_K$  = unitary conductance

Total K conductance:

$$G_K = P_K(V, t) N_K \hat{g}_K$$



Remember, K current is:

$$I_K = G_K(V, t) (V - E_K)$$

All of the interesting time and voltage dependence  
comes from here

# Ionic conductance in terms of single channels

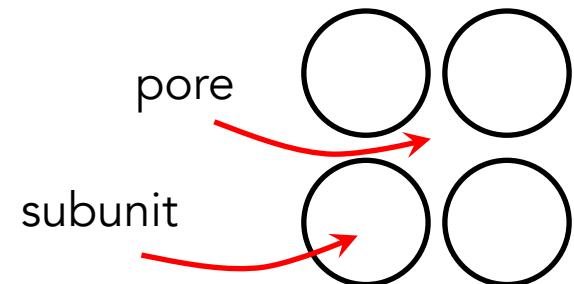
How do we describe the probability that a channel is open?

Lets start with the K channel.

The pore of a  $K^+$  channel is formed by 4 identical subunits.

Each subunit has a voltage sensor and gate to turn the channel on and off.

Each subunit has two states: 'open' and 'closed'.  
'n' is the probability that a subunit is open.

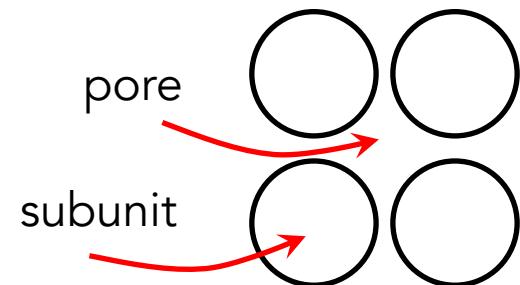


Finally, all subunits must be in the 'open' state for the channel to be permeable.

# Ionic conductance in terms of single channels

How do we describe the probability that a channel is open?

If 'n' is the probability that one subunit is open, then the probability that all four subunits is open is given by:



$$P_K = n^4 \quad \text{Assuming independence}$$

We can now write down the conductance of our K channels as:

$$G_K = \bar{G}_K n^4$$

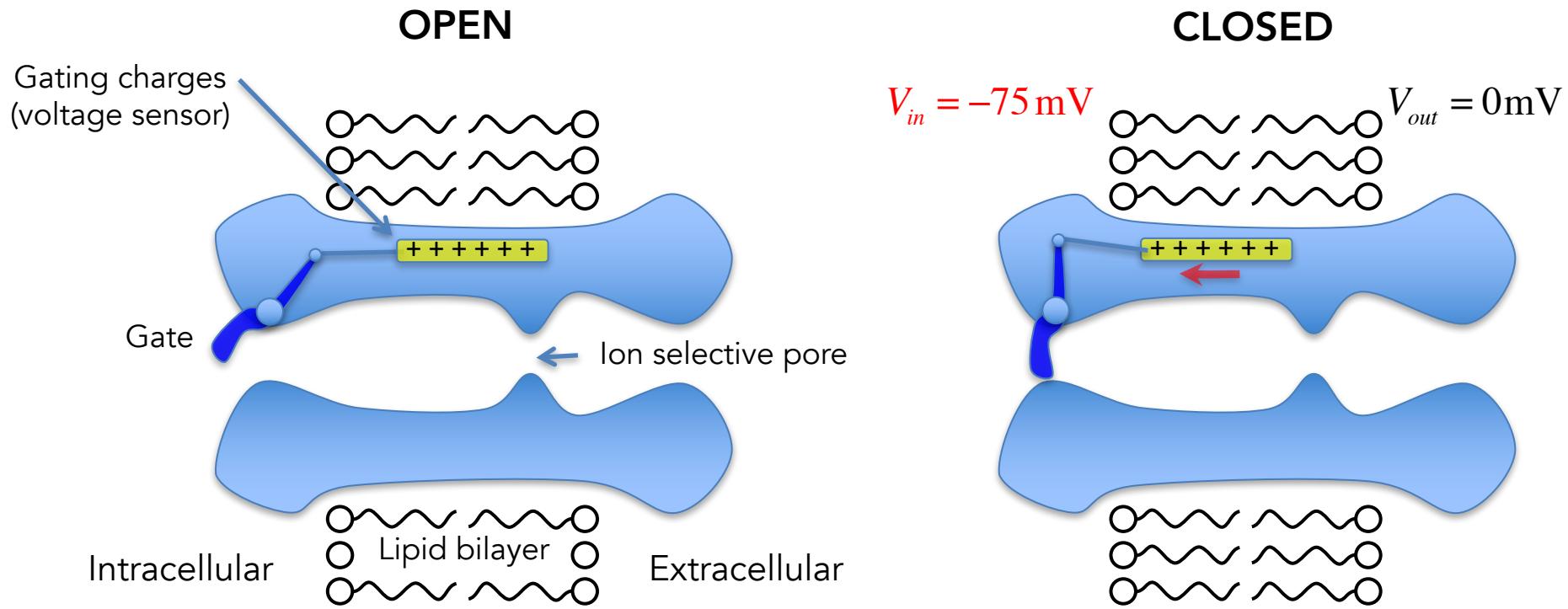
$\bar{G}_K$  is the maximal open conductance

And we can write the K current as:

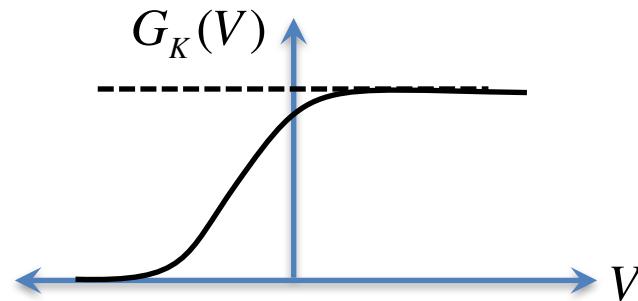
$$I_K = \bar{G}_K n^4 (V - E_K)$$

H & H called  $n$  the 'gating variable' for the potassium current

# Ionic conductance in terms of single channels



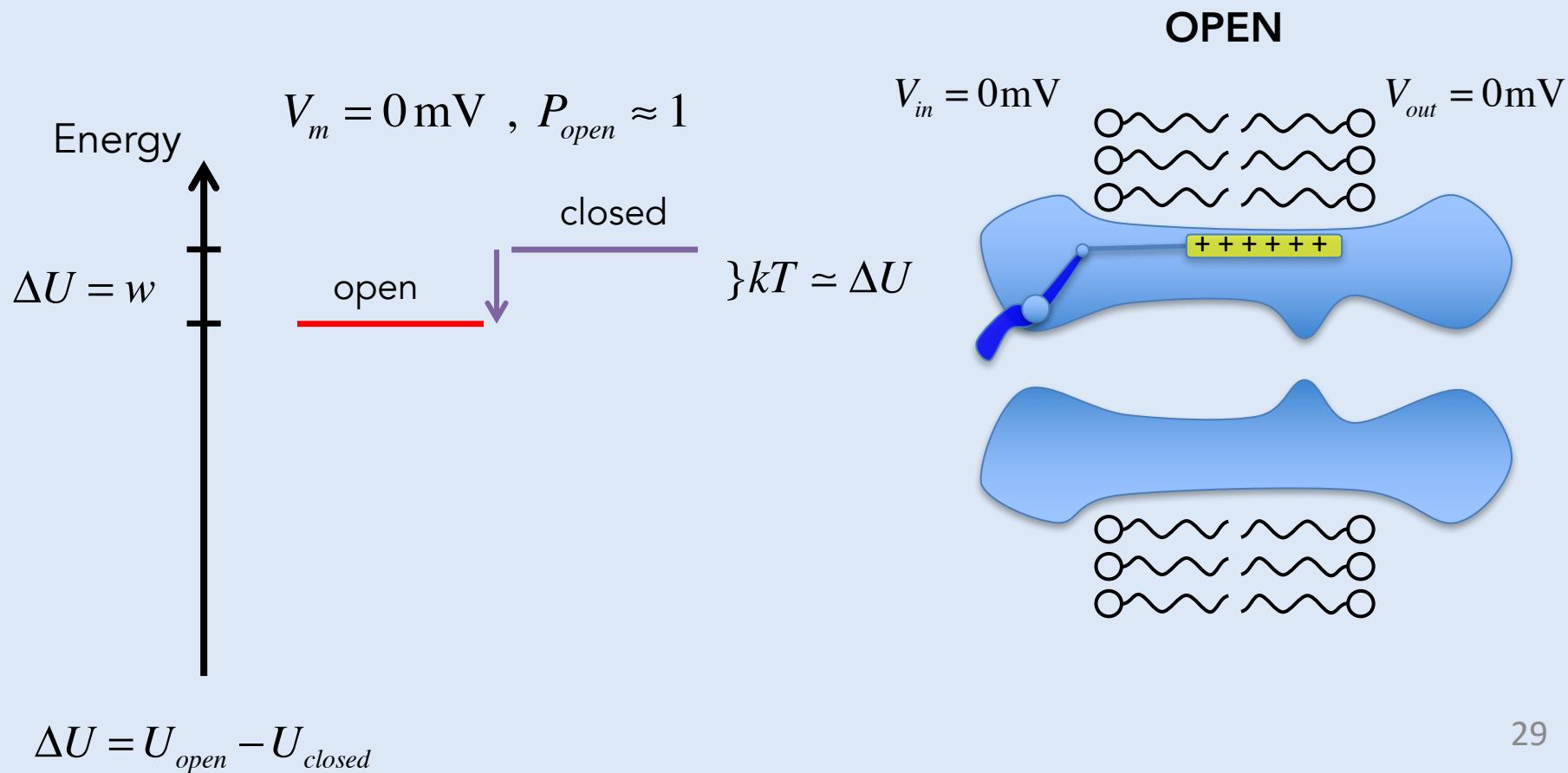
We are going see how to derive  
the voltage-dependence from first  
principles!



# Ionic conductance in terms of single channels

Ion channels are stochastic. They are either open or closed, and flicker back and forth between the open and closed states.

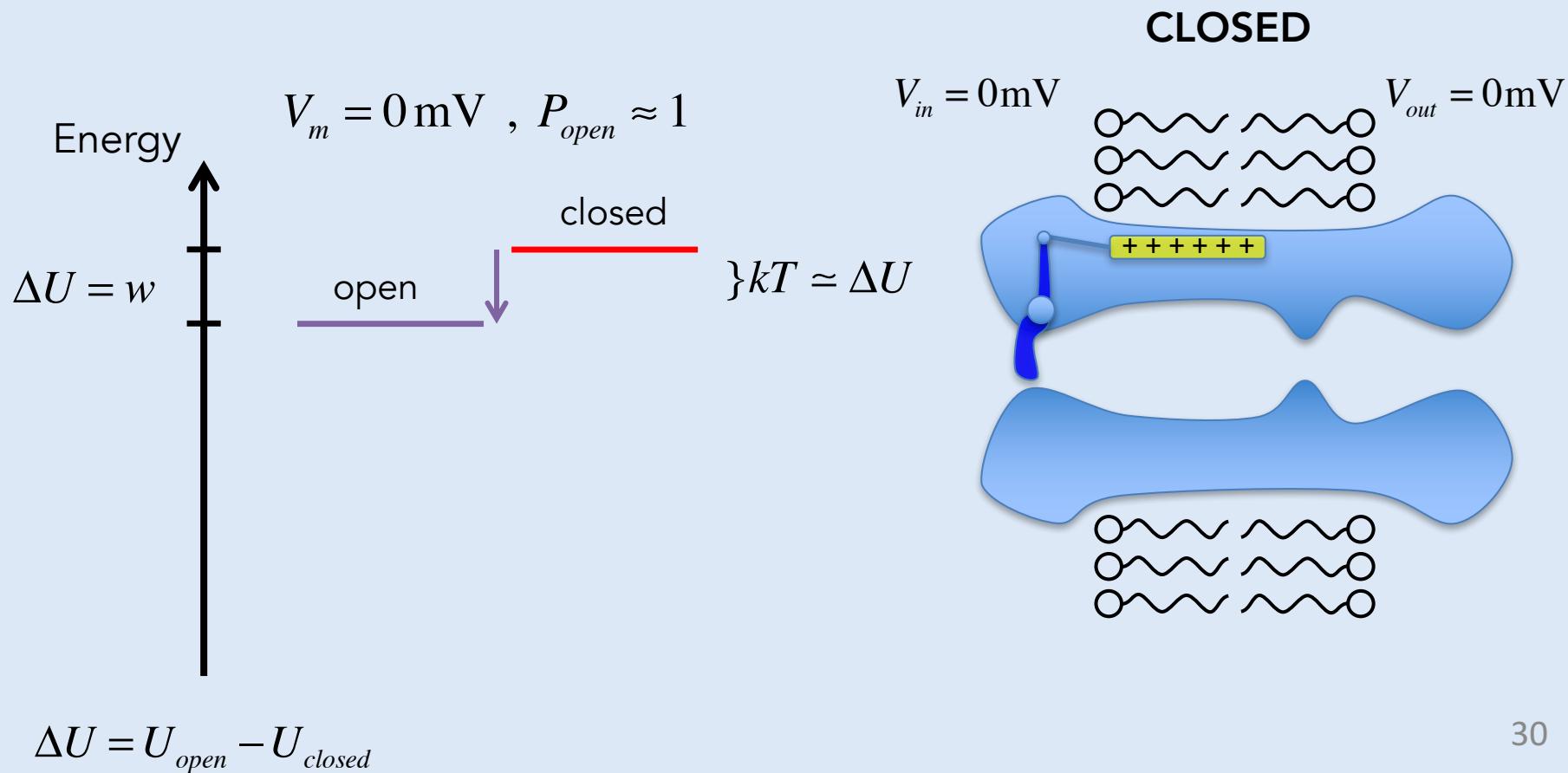
Lets see if we can predict the voltage-dependence of an ion channel!



# Ionic conductance in terms of single channels

Ion channels are stochastic. They are either open or closed, and flicker back and forth between the open and closed states.

Lets see if we can predict the voltage-dependence of an ion channel!

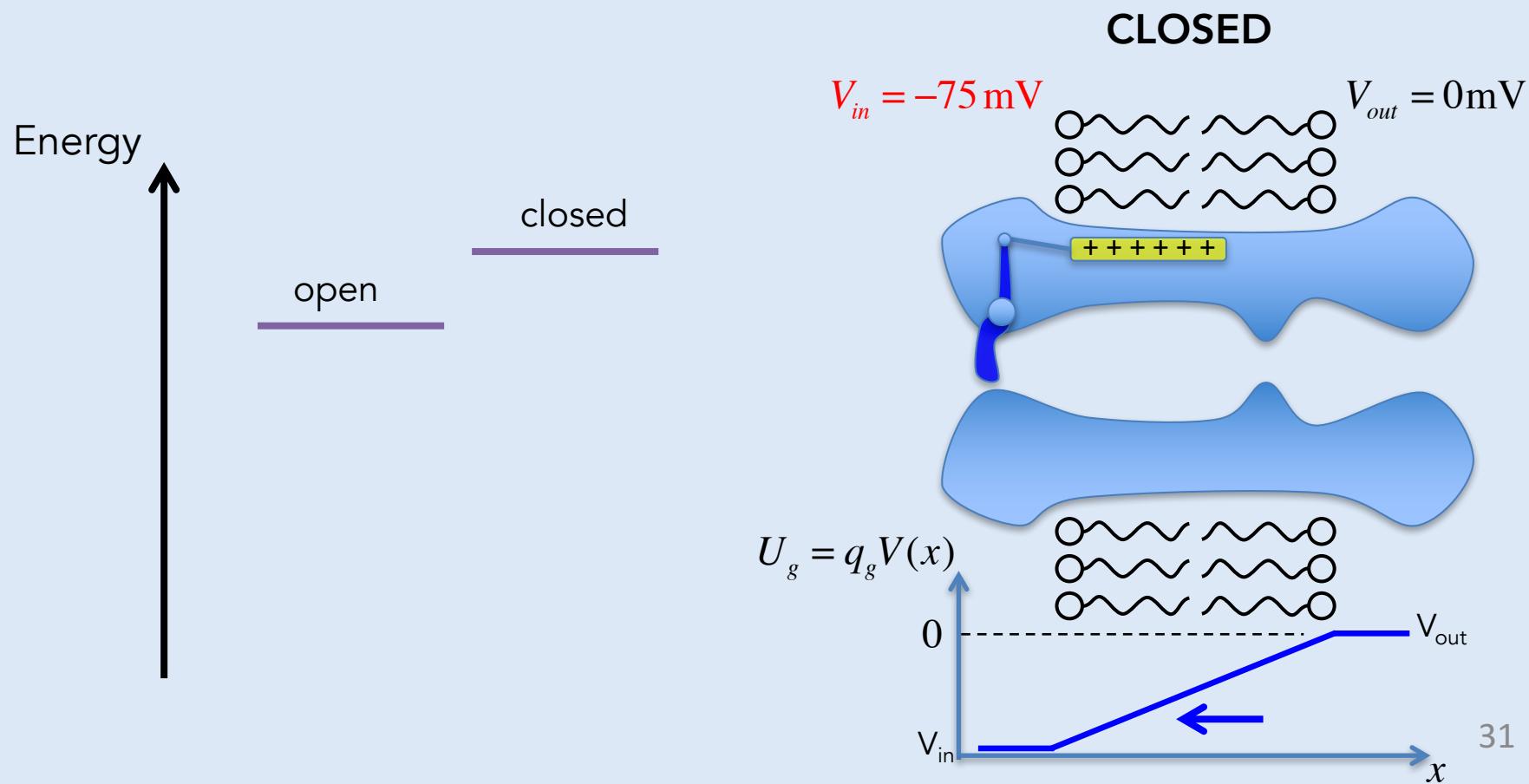


$$\Delta U = U_{open} - U_{closed}$$

# Ionic conductance in terms of single channels

Ion channels are stochastic. They are either open or closed, and flicker back and forth between the open and closed states.

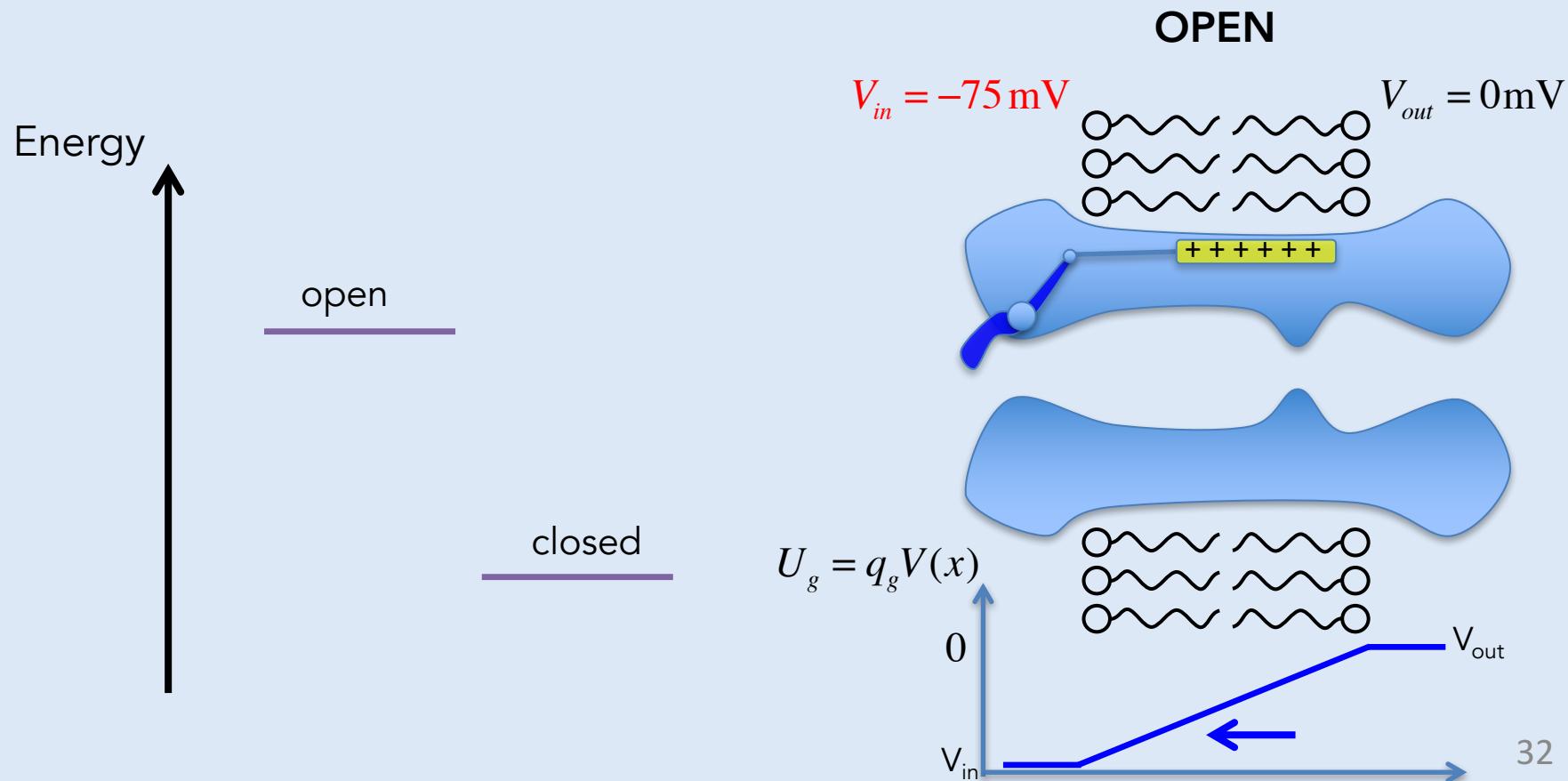
Lets see if we can predict the voltage-dependence of an ion channel!



# Ionic conductance in terms of single channels

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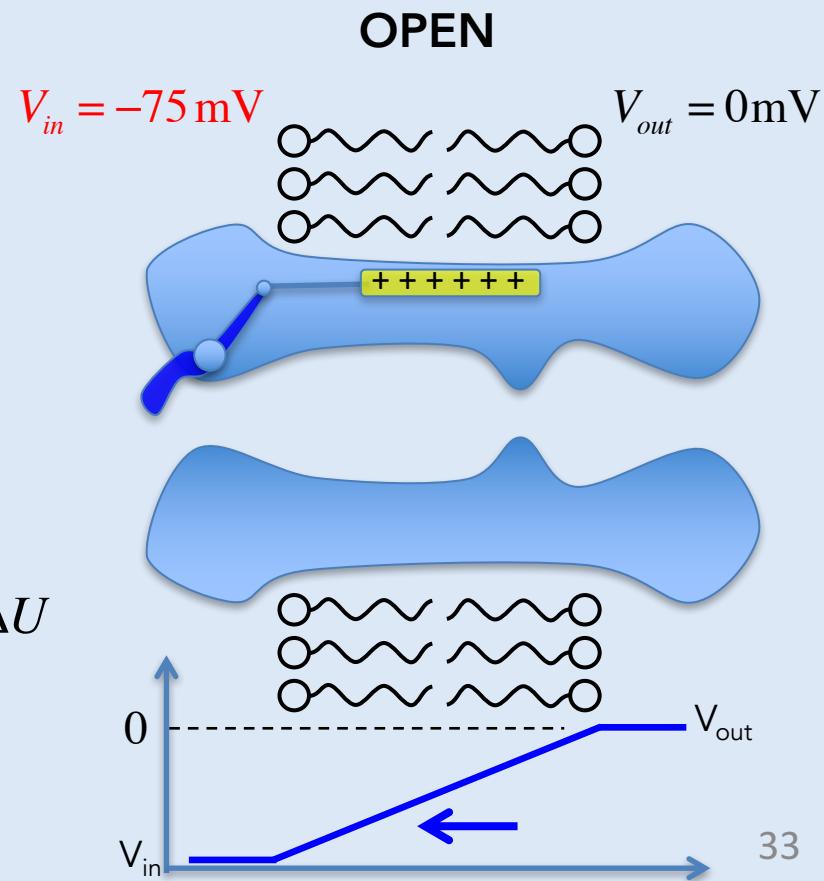
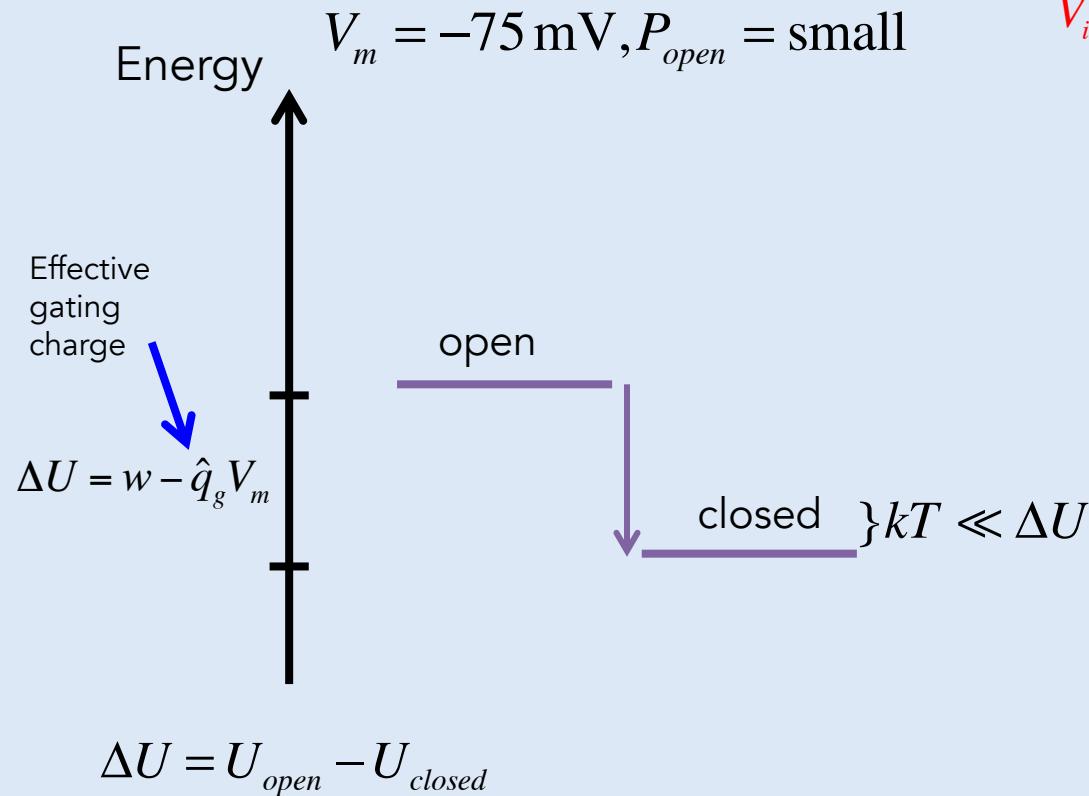
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# Ionic conductance in terms of single channels

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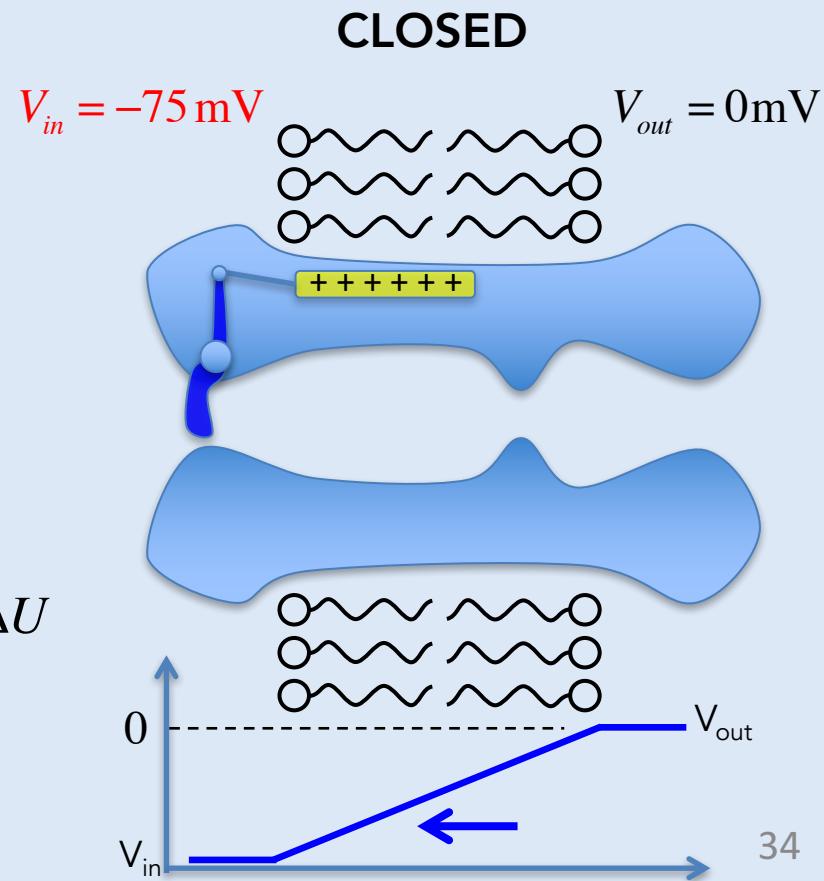
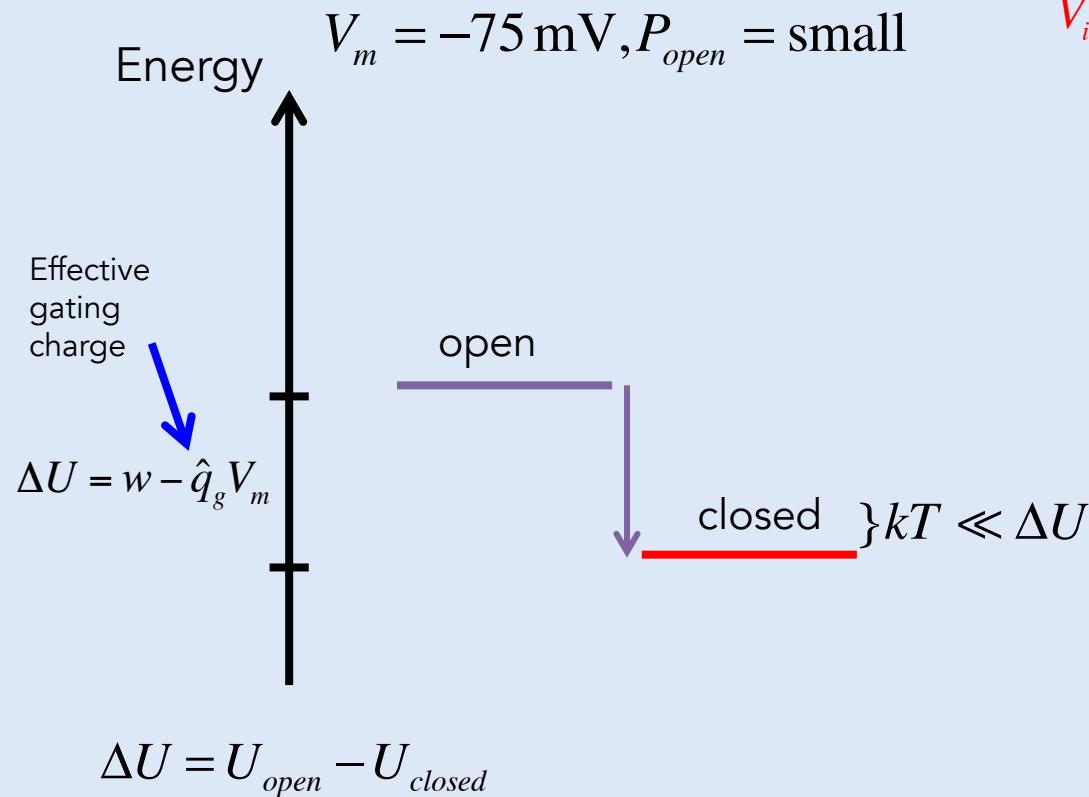
Lets see if we can predict the voltage-dependence of an ion channel!



# Ionic conductance in terms of single channels

Ion channels are stochastic. They are either open or closed, and flicker back and forth between the open and closed states.

Lets see if we can predict the voltage-dependence of an ion channel!



# To derive the voltage dependence: boltzmann equation

We can use the Boltzmann equation to describe the ratio of probabilities of being in the open or closed state:

$$\frac{P_{open}}{P_{closed}} = e^{-\left(\frac{U_{open}-U_{closed}}{kT}\right)}$$

$$\frac{P_{open}}{P_{closed}} = e^{-\left(\frac{w-q_g V_m}{kT}\right)}$$

The probability of having an open subunit is:

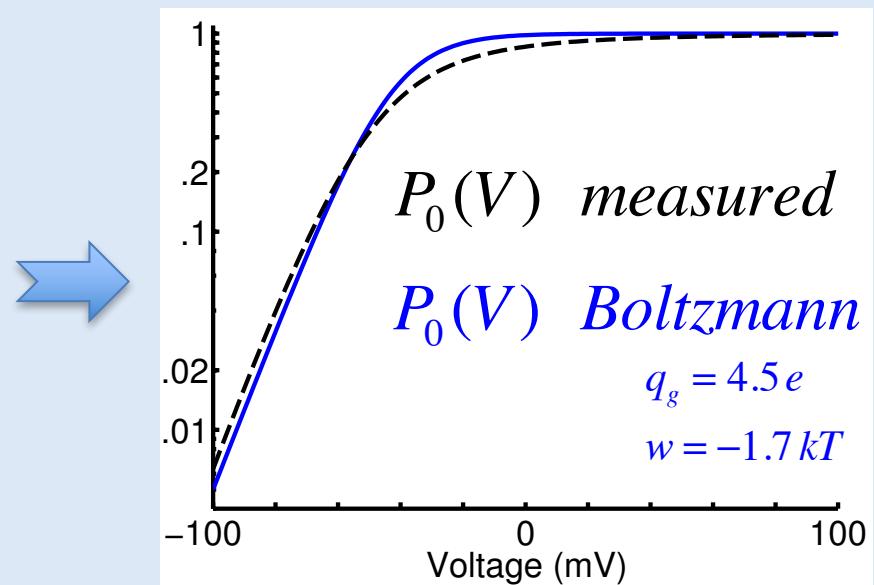
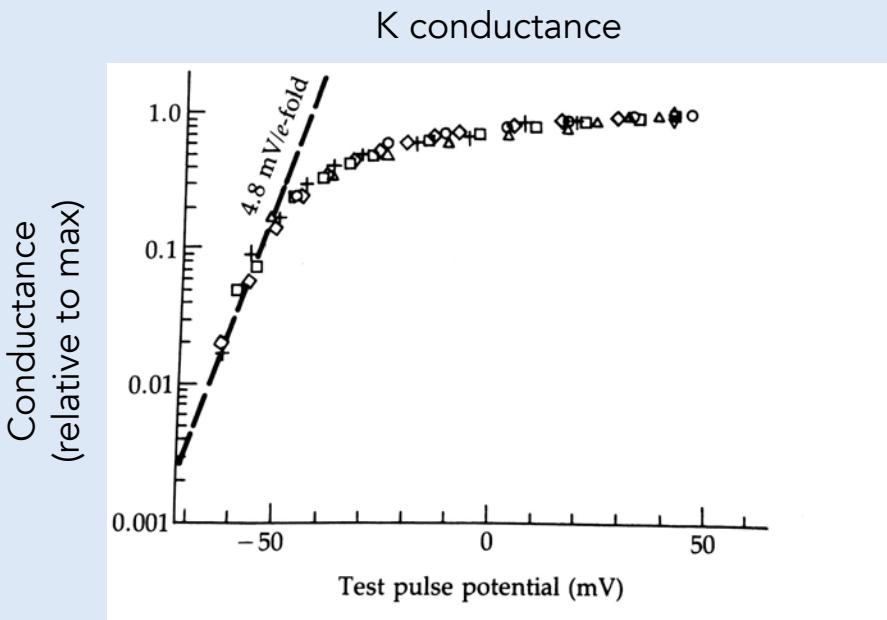
$$n = P_0 = \frac{P_o}{P_o + P_c} = \frac{1}{1 + \frac{P_c}{P_o}} = \frac{1}{1 + e^{(w-q_g V_m)/kT}}$$

# To derive the voltage dependence: boltzmann equation

$$P_o(V) = \frac{1}{1 + e^{(w - q_g V_m)/kT}}$$

Is this the observed voltage-dependence?

YES!

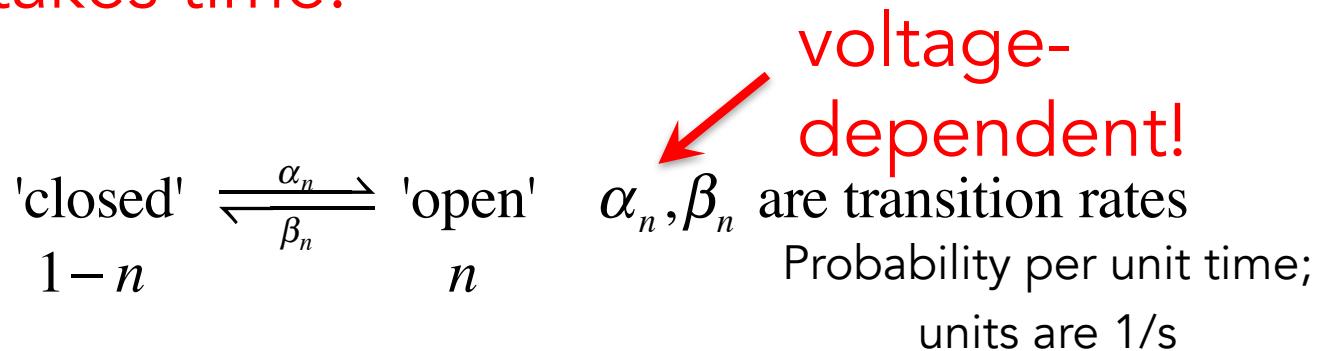


# Now get the time dependence!

Note that for any given subunit or channel, when we change the voltage, the energy levels shift (nearly) instantaneously. The probability of making a transition changes instantly, however the number of open channels does not change instantly.

We still have to wait for thermal fluctuations to kick the channel or subunit open, or wait for the subunit to make a conformational change.

This takes time!



We are going to model the transitions between open and closed states with a simple rate equation. We can do this because we have many channels to average over.

'closed'  $\xrightleftharpoons[\beta_n]{\alpha_n}$  'open'  $\alpha_n, \beta_n$  are transition rates

$1-n$                            $n$

Change in the number of open subunits = The number of closed subunits that open - The number of open subunits that close

Change in the number of open subunits per unit time = The number of closed subunits  $(1-n)$  times the probability that a closed subunit opens per unit time  $(\alpha_n)$  - The number of open subunits  $(n)$  times the probability that an open subunit closes per unit time  $(\beta_n)$

$$\frac{dn}{dt} = \alpha_n(1-n) - \beta_n n$$

$$\frac{dn}{dt} = \alpha_n(1-n) - \beta_n n$$

Let's rewrite this equation as follows:

$$= \alpha_n - \alpha_n n - \beta_n n$$

$$= \alpha_n - (\alpha_n + \beta_n)n$$

$$\underbrace{\frac{1}{(\alpha_n + \beta_n)} \frac{dn}{dt}}_{\text{red bracket}} = \underbrace{\frac{\alpha_n}{(\alpha_n + \beta_n)}}_{\text{red bracket}} - n$$

Thus, we can rewrite this equation in terms of the steady state open probability and a time constant:

$$\tau_n \frac{dn}{dt} = n_\infty - n$$

Remember:  $n$  is the probability that a subunit is open.

The steady state solution!

$$n_\infty = \frac{\alpha_n}{(\alpha_n + \beta_n)}$$

A time constant!

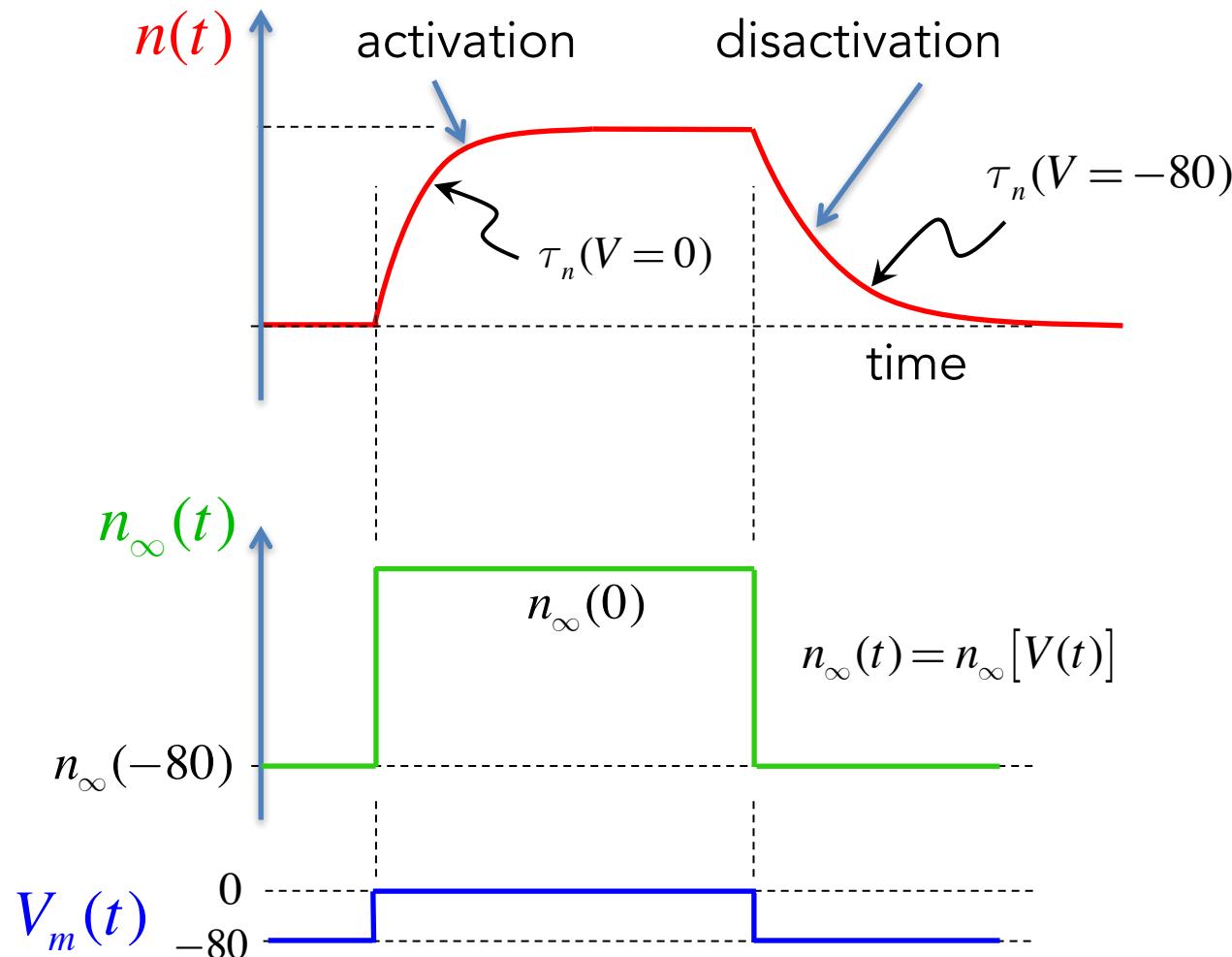
$$\tau_n = \frac{1}{(\alpha_n + \beta_n)}$$

But we just derived  $n_\infty(V)$  !

Remember...  $\alpha_n, \beta_n, n_\infty, \tau_n$  are all voltage dependent

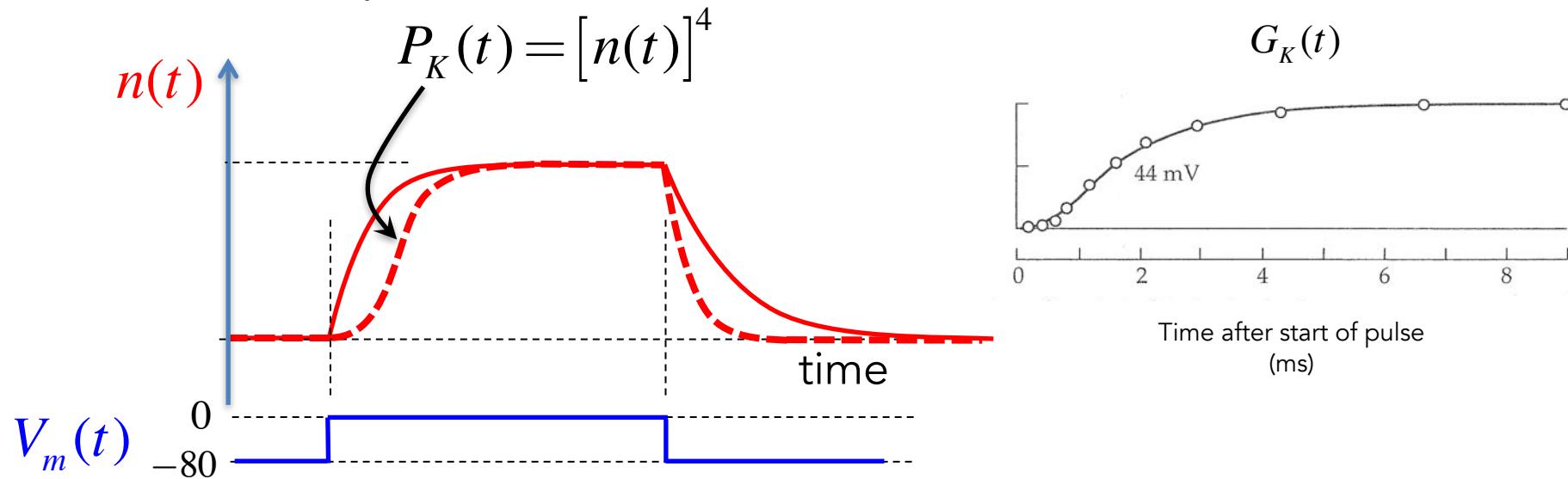
# Response to voltage change

How does the gating variable 'n' change as we step the membrane potential?



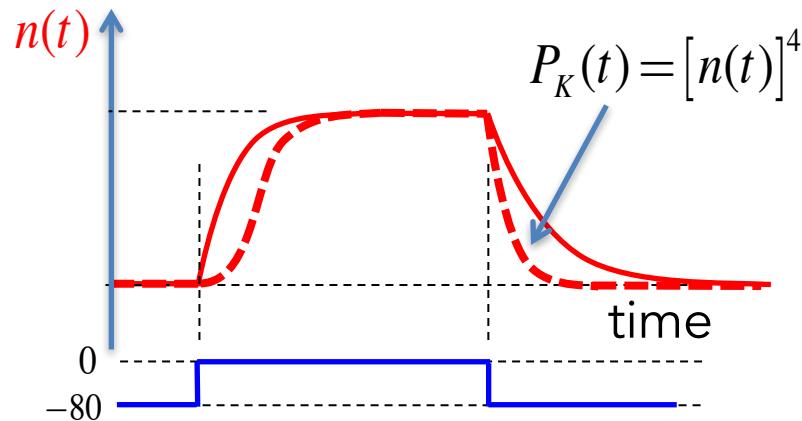
# Response to voltage change

How does the 'open' probability change as we step the membrane potential?

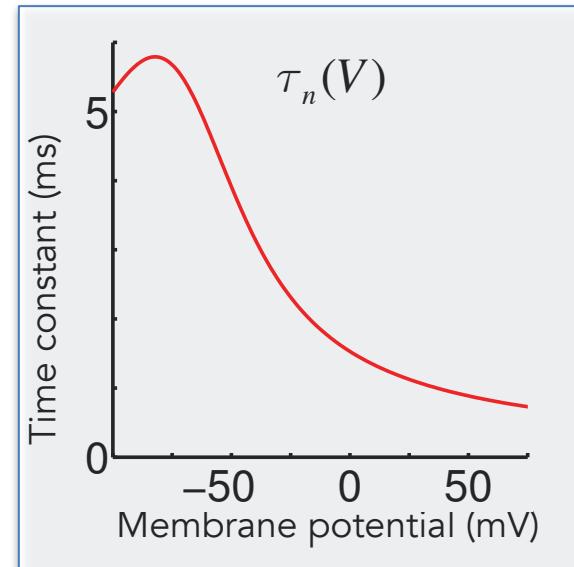
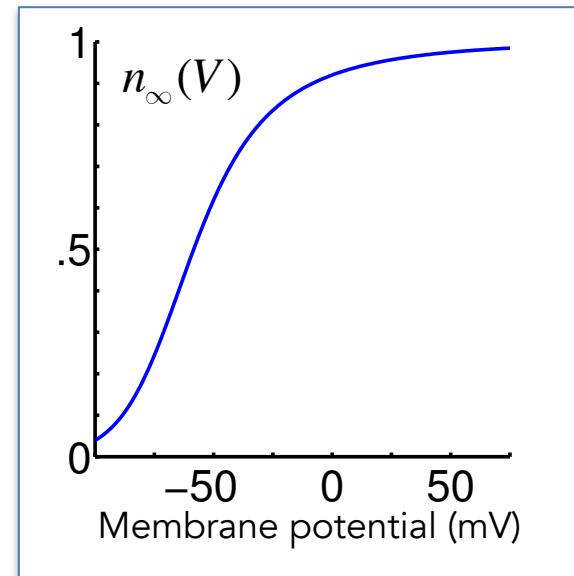


- The shape of the K conductance was well fit by a rising exponential raised to the fourth power.
- H & H inferred from this that the K-current was governed by four independent first-order processes! (They didn't know about the structure of K-channels at the time!)

# Measuring the parameters



- By measuring the persistent conductance at different voltages, they were able to measure  $n_\infty$  as a function of voltage.
- By measuring the time course of the conductance at onset and offset of the voltage steps, they were able to measure  $\tau_n$  as a function of voltage.



# Measuring the parameters

Hodgkin and Huxley summarized their data using algebraic expressions for the rate functions  $\alpha_n(V)$  ,  $\beta_n(V)$

$$\alpha_n(V) = \frac{1}{1 - \exp(-0.1(V + 55))}$$

$$\beta_n(V) = 0.125 \exp(-0.0125(V + 65))$$

$V$  is in mV

$\alpha_n, \beta_n$  are in  $\text{ms}^{-1}$

Remember that...

$$n_\infty = \frac{\alpha_n}{(\alpha_n + \beta_n)}$$

$$\tau_n = \frac{1}{(\alpha_n + \beta_n)}$$

# Measuring the parameters

Why did we do all of this?

Because once we have expressions for  $n_\infty(V)$  and  $\tau_n(V)$ , we can integrate the differential equation for  $n$  :

$$\tau_n \frac{dn}{dt} = n_\infty - n$$

to get the potassium conductance:

$$G_K = \bar{G}_K n^4$$

and the potassium current:

$$I_K = \bar{G}_K n^4 (V - E_K)$$

# We are going to write down an algorithm for how a neuron spikes!

For now here the parts related to the potassium current...

Start with  $V_m$  at time step t

→ Compute  $n_\infty(V)$  and  $\tau_n(V)$

Integrate  $\frac{dn}{dt}$  one time step to get  $n(t)$

Compute K current:  $I_K = \bar{G}_K n^4 (V - E_K)$

Compute total membrane current  $I_m = I_K + I_{Na} + I_L$

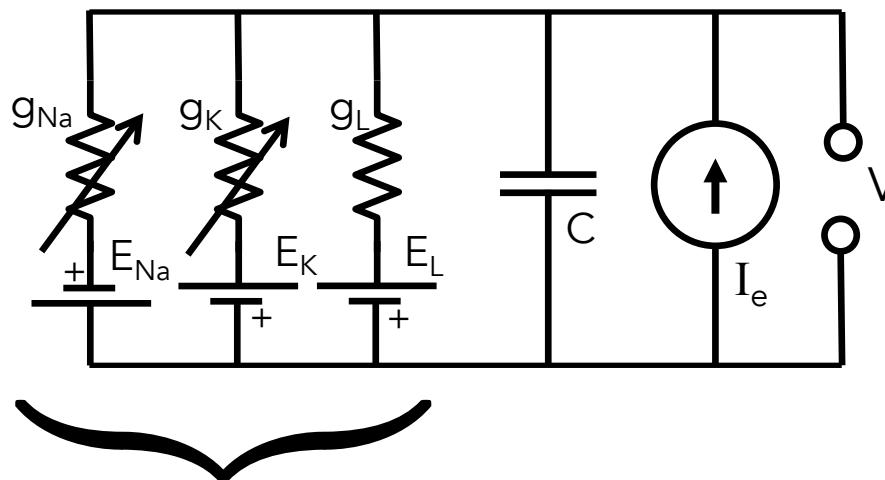
Compute  $V_\infty$

Integrate  $\frac{dV_m}{dt}$  to get  $V_m$  at next time step

# Learning objectives for Lecture 2

- To be able to draw the circuit diagram of the HH model
- Understand what a voltage clamp is and how it works
- Be able to plot the voltage and time dependence of the potassium current and conductance
- Be able to explain the time and voltage dependence of the potassium conductance in terms of Hodgkin-Huxley gating variables

# Outline of HH model



$$I_m = I_{Na} + I_K + I_L$$

$$I_m(t) + C \frac{dV(t)}{dt} = I_e(t)$$

$$I_{Na} = G_{Na}(V,t)(V - E_{Na})$$

The sodium conductance is time-dependent and voltage-dependent

$$E_{Na} = +55\text{mV}$$

$$I_K = G_K(V,t)(V - E_K)$$

The potassium conductance is time-dependent and voltage-dependent

$$E_K = -75\text{mV}$$

$$I_L = G_L(V - E_L)$$

The leak conductance is neither time-dependent nor voltage-dependent

$$E_L = -50\text{mV}$$

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