

Simulation and Inference in Neuroscience

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Lecture 1

What is Hertie AI?

- Institute at the Medical School of the University of Tübingen
- Research at the interface between AI and clinical neuroscience
- Funded by the Charitable Hertie Foundation
- Two departments:
 - Data Science
 - Machine Learning



Models in science



Making models is part
of the scientific method



Models capture some
aspect of reality

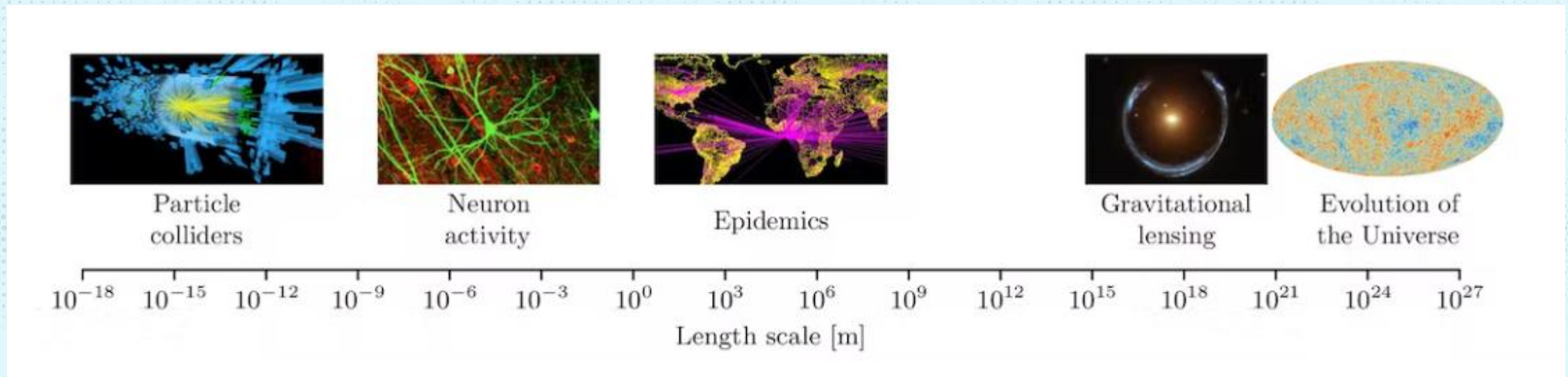


Models are testable and
make predictions

- Didactic models
- Idealized models
- Phenomenological models
- Computational models
- Living models

<https://plato.stanford.edu/entries/models-science/>

Models in science

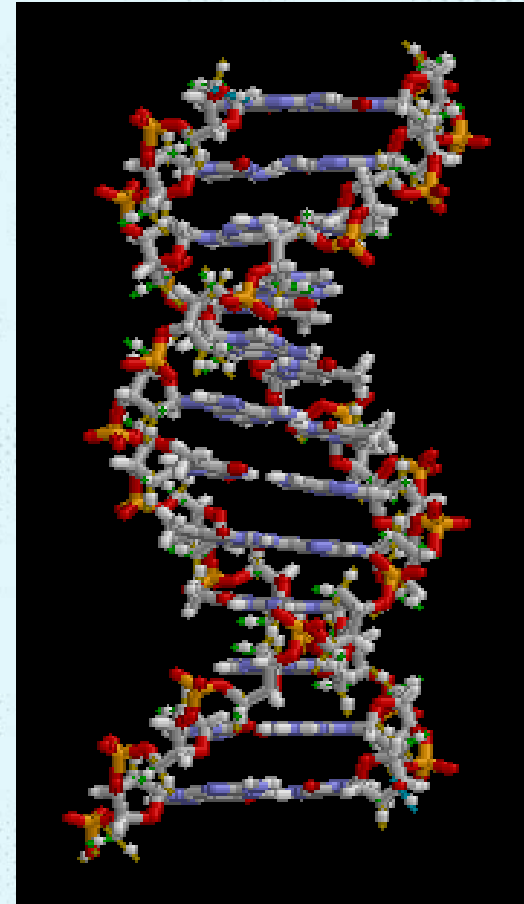


DNA double helix

Physical model

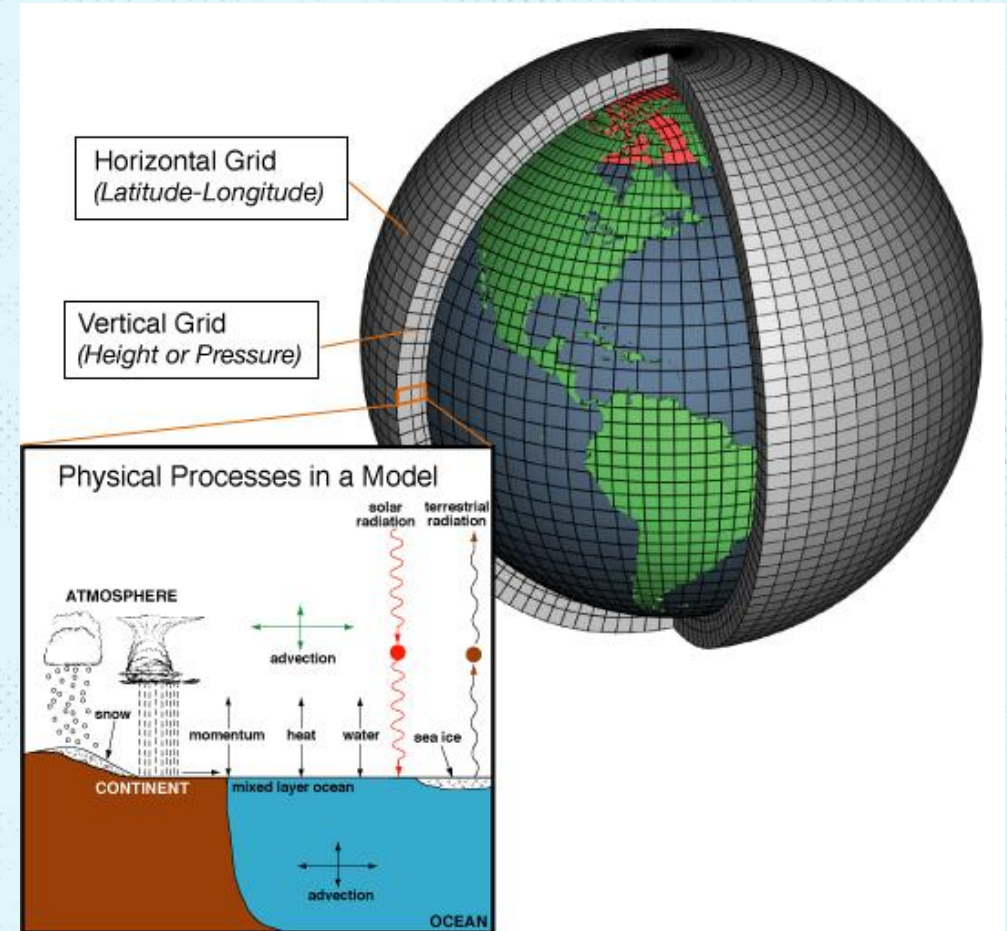
Explains previous data, e.g. X-ray diffraction images

Makes predictions about function of enzymes, consequences of mutations and ...



Climate model

- Describes physical processes: material in each cell and how energy moves around
- Based on our knowledge of physics and chemistry
- Consists of equations
- Can be simulated, i.e. computed in a computer



Simulation in science

Equations define explicit model

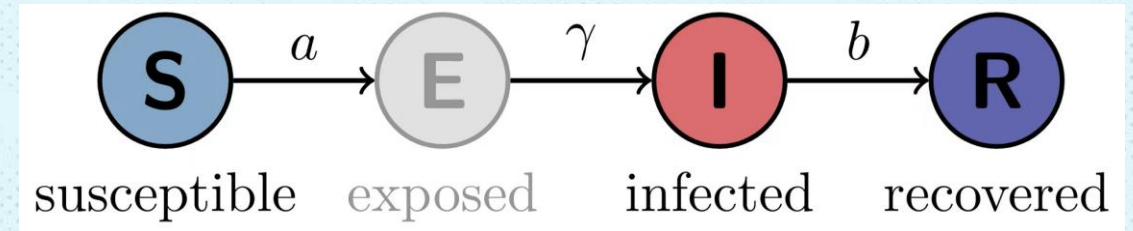
Ordinary differential equations

Partial differential equations

Stochastic differential equations

...

If there is no analytic solution, use numerical integration to solve the equation

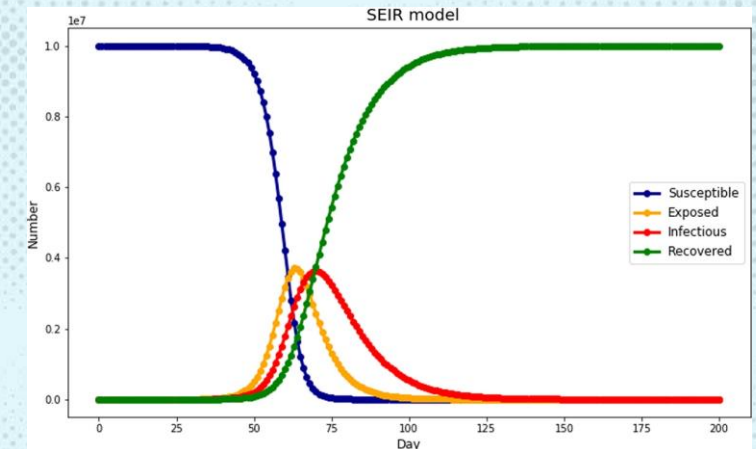


$$\frac{dS}{dt} = -aSI$$

$$\frac{dE}{dt} = aSI - \gamma E$$

$$\frac{dI}{dt} = \gamma E - bI$$

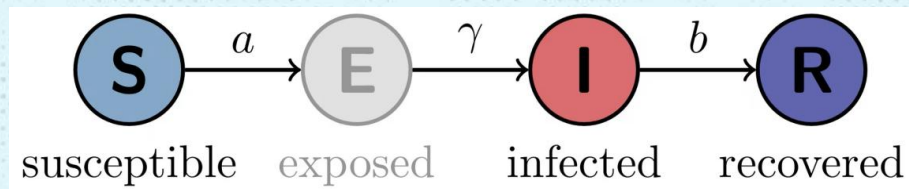
$$\frac{dR}{dt} = bI$$



Simulation vs. statistical models

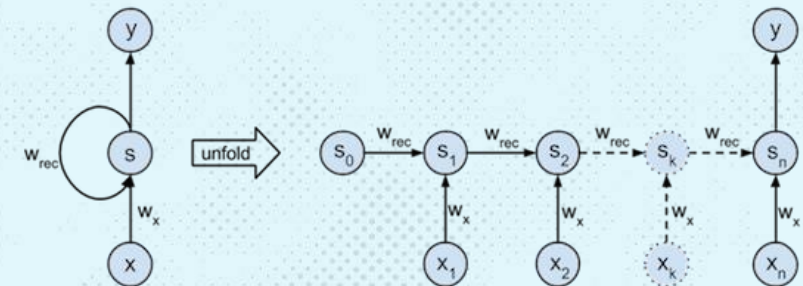
ODE-based SIR model

- ODE, limited modeling repertoire
- Use rate constant for modeling kinetics
- Few interpretable parameters
- Expresses knowledge about world



Recurrent Neural Network model

- RNN, flexible approximator
- Specific architecture to fit temporal correlations in data
- Many opaque parameters
- Model hard to interpret
- Data hungry
- Unclear extrapolation



Simulation and inference?

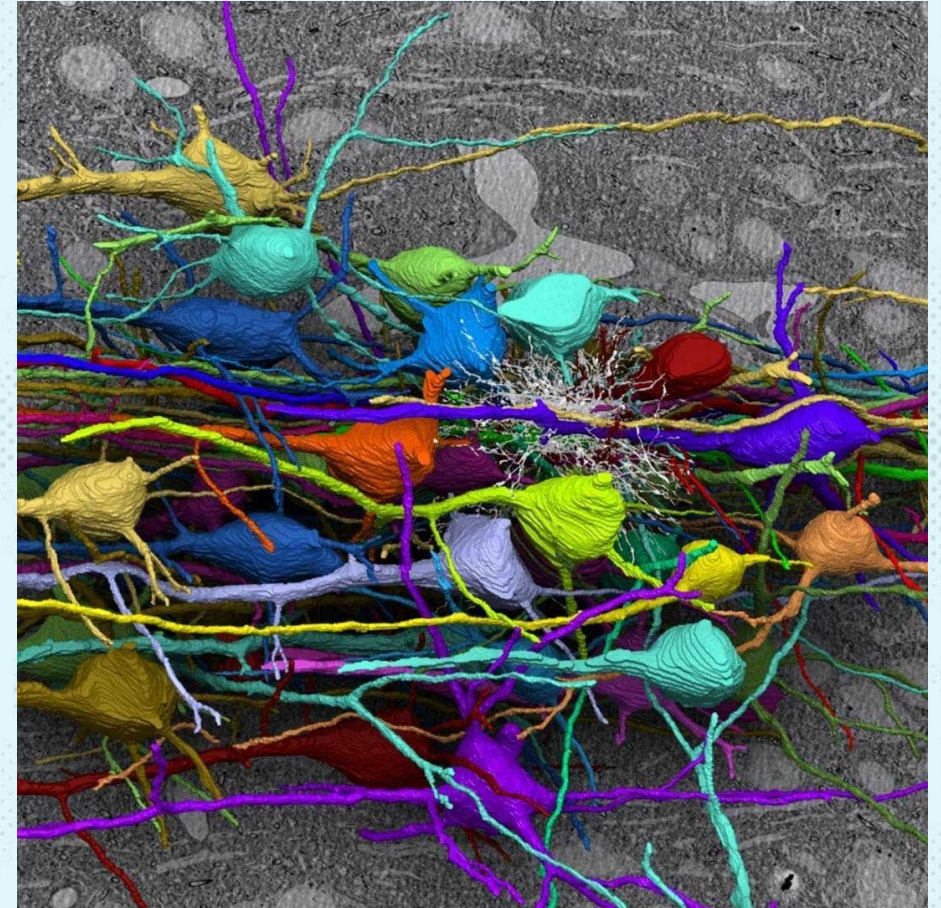
- From parameters to measurements:
simulation - *forward problem*
- From measurements to parameters:
parameter estimation - *inverse problem*
 - Classical modeling (hand-tuning, experiments)
 - Point estimates via optimization
 - Parameter estimates with uncertainty
- Modern ML techniques for this will be the content of week 2 and 3!

Overview of the course

- Week 1:
 - Neuroscience basics
 - Mechanistic models in neuroscience
 - Implementing and simulating models
- Week 2:
 - Bayesian parameter inference
 - Simulation-based inference (SBI) basics
 - Implementing examples to get experience with SBI
- Week 3:
 - Advanced SBI techniques
 - Mini-project

Why neuroscience?

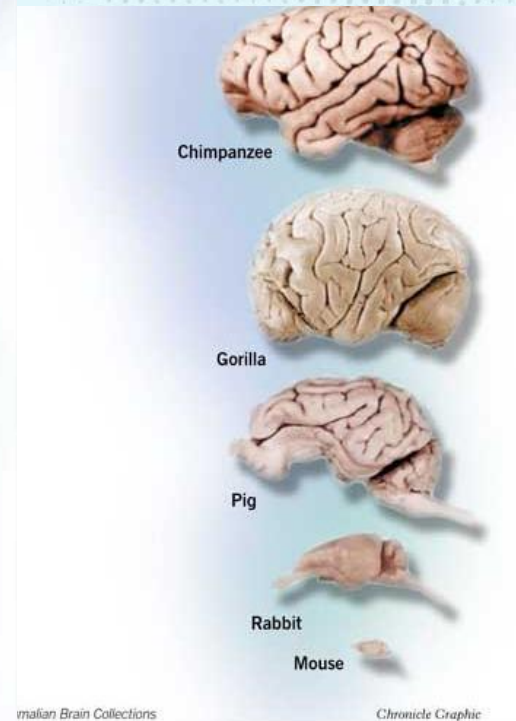
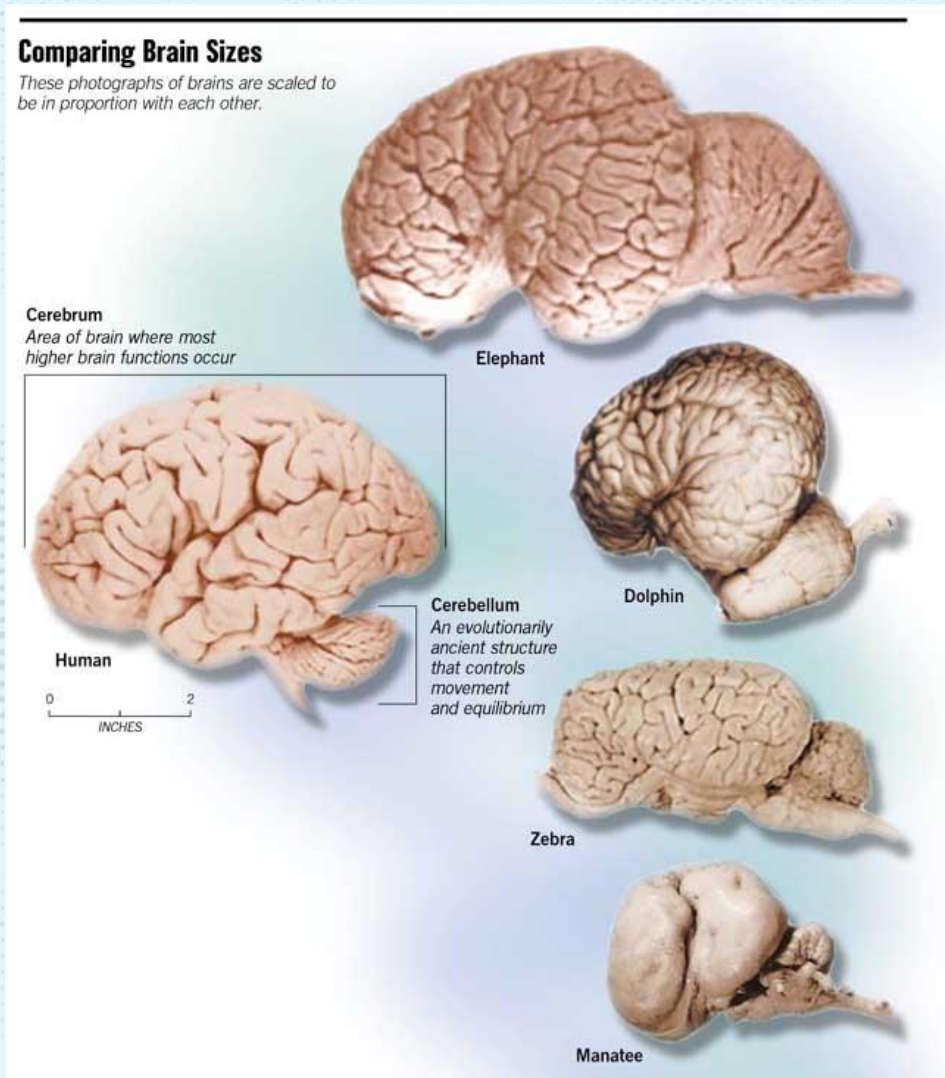
- Understanding to organ that allows us to do science is fun
- Neurological diseases are a major burden on society (dementia, multiple sclerosis, ...)
- There is amazing data!



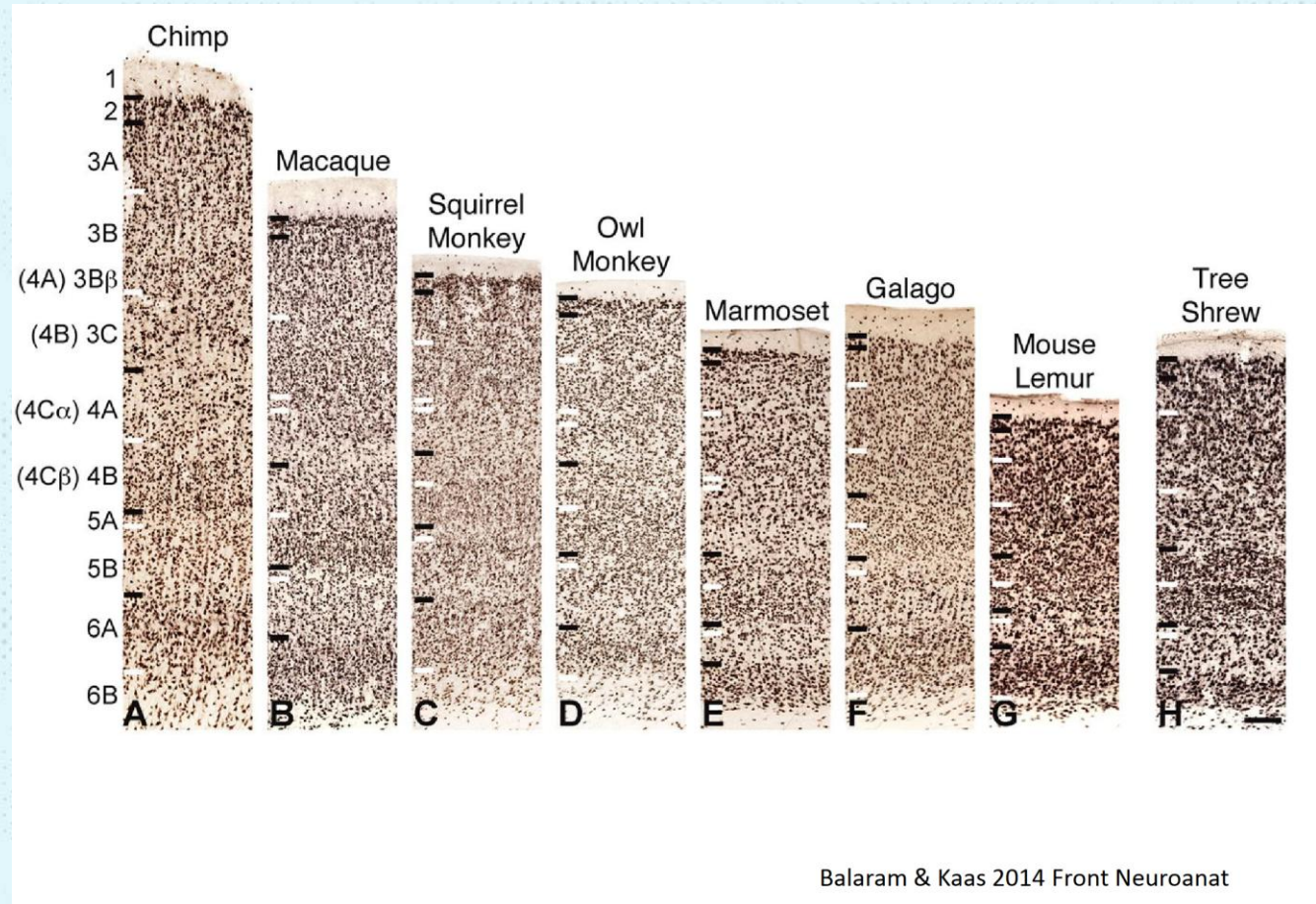
What does neuroscience do?

- Understand brain function on different levels – from psychological to cellular or even subcellular
- Figure out mechanisms of how the healthy brain works and how it changes in disease
- Test a hypothesis
 - Measure neuronal activity
 - Make mathematical models of neuronal activity

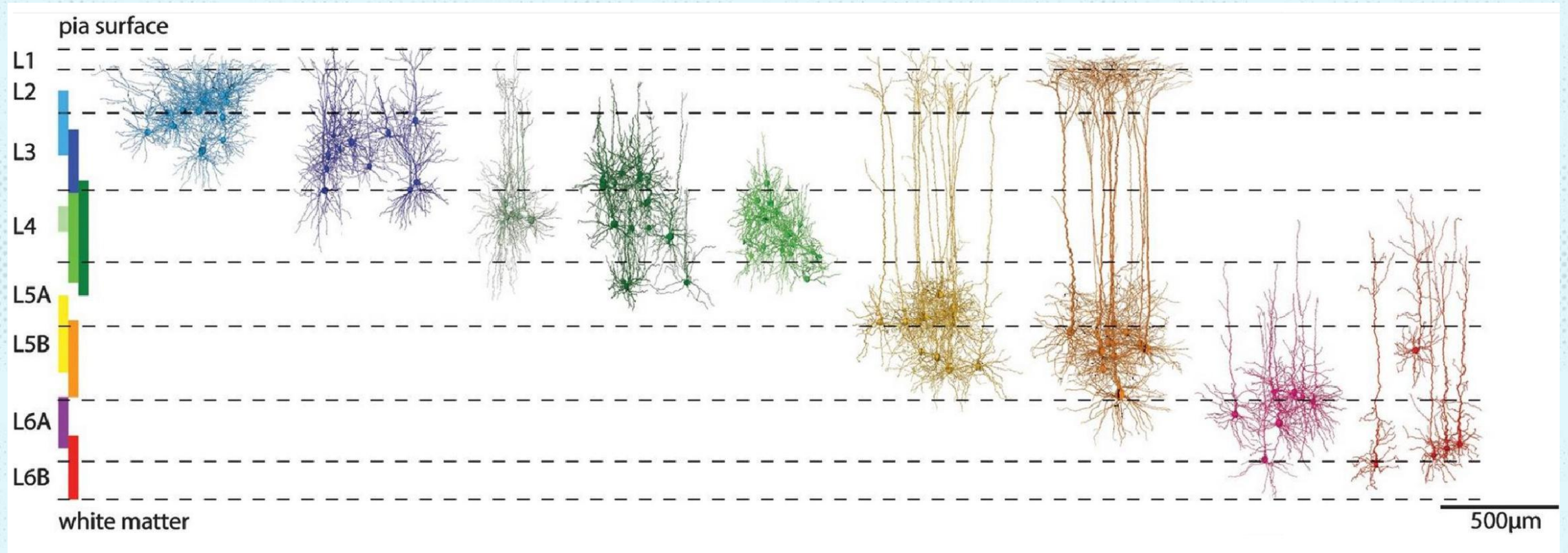
Mammalian brains



Layers of neurons

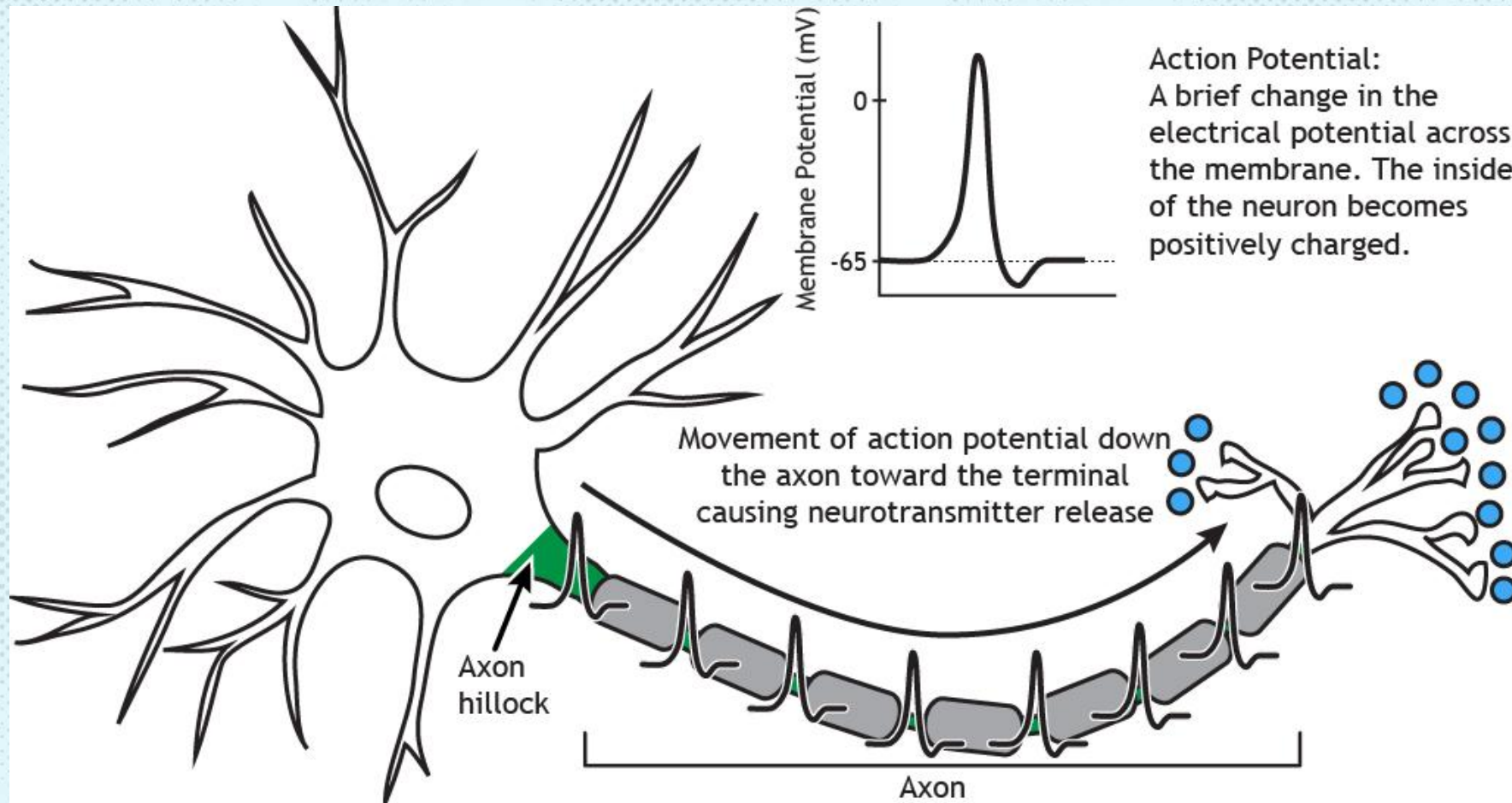


Different layers, different cells



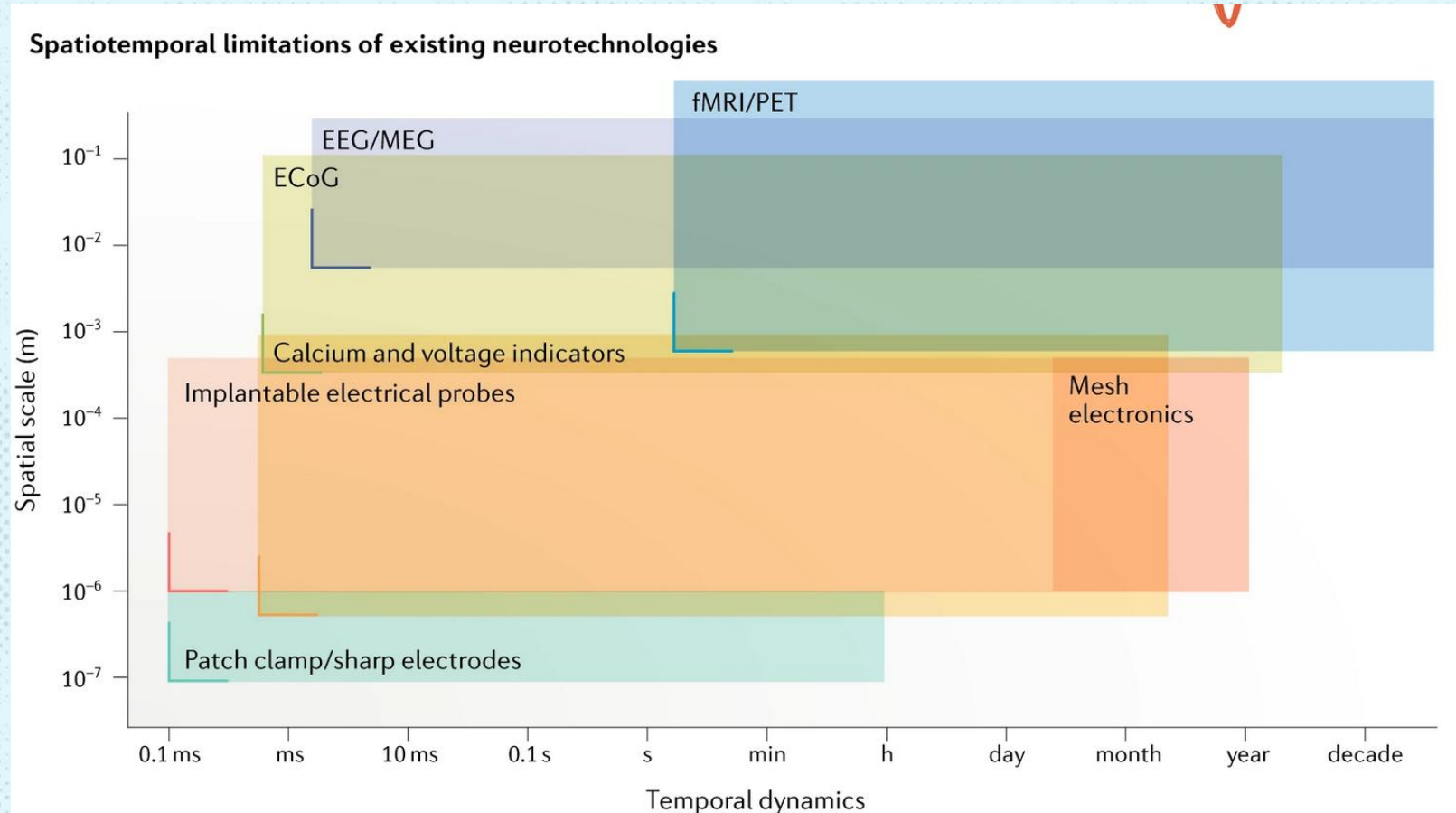
Oberlaender et al. 2012

A prototypical neuron

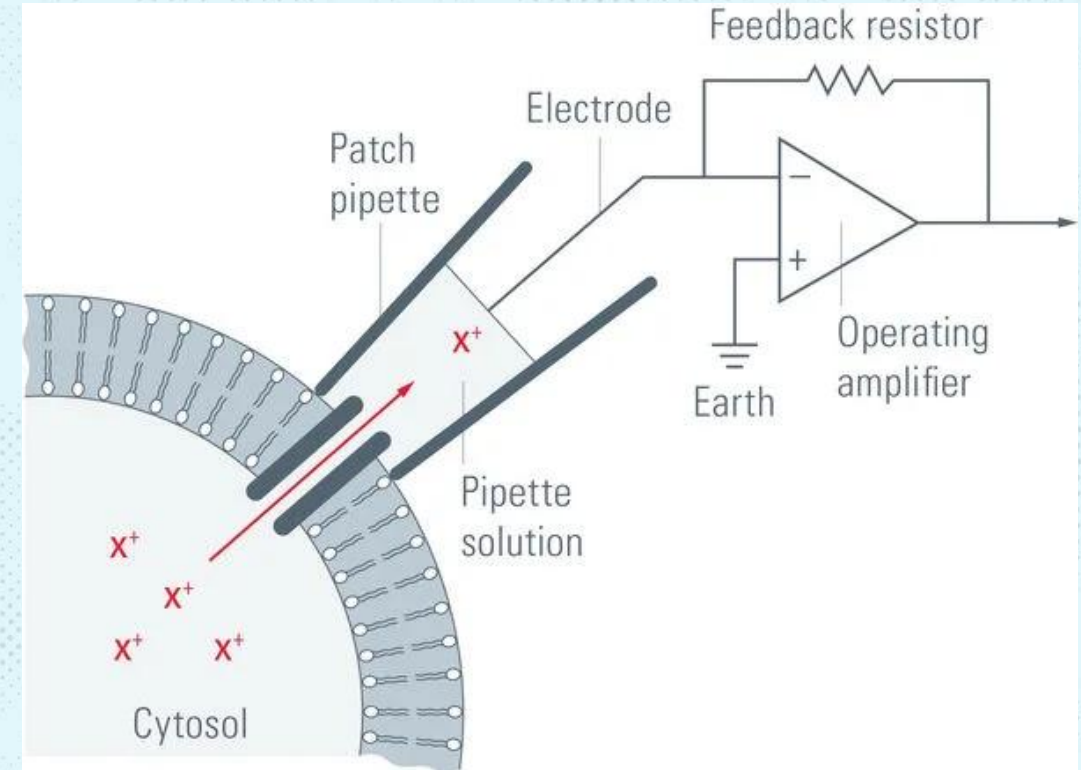
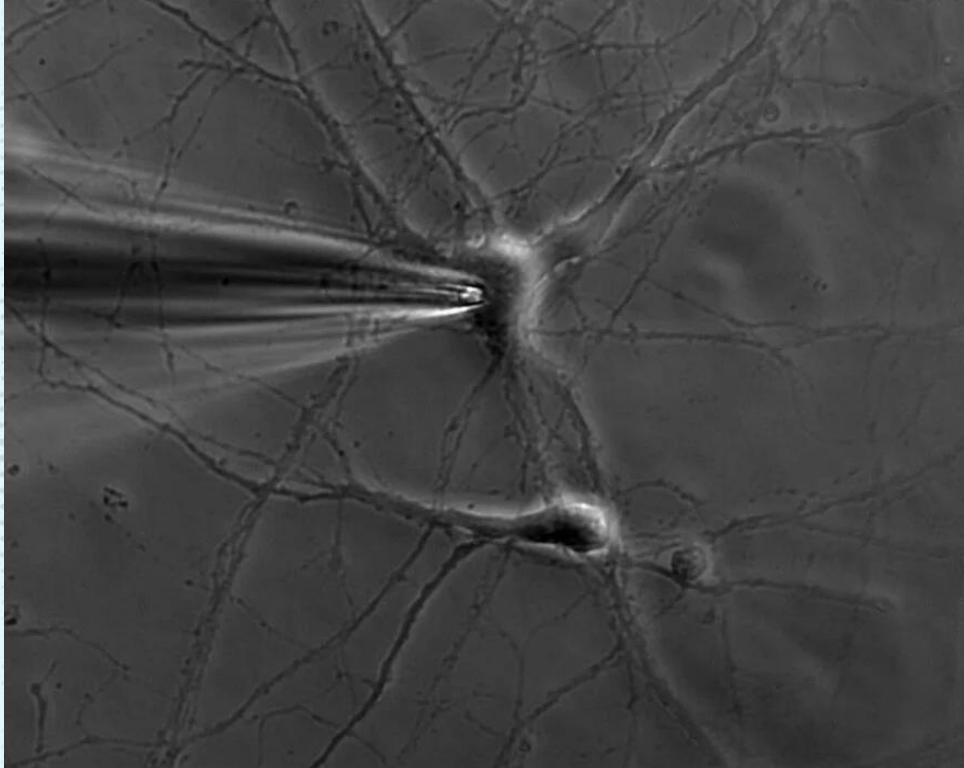


Casey Henley, CC-by-NC-SA

Neural recording techniques



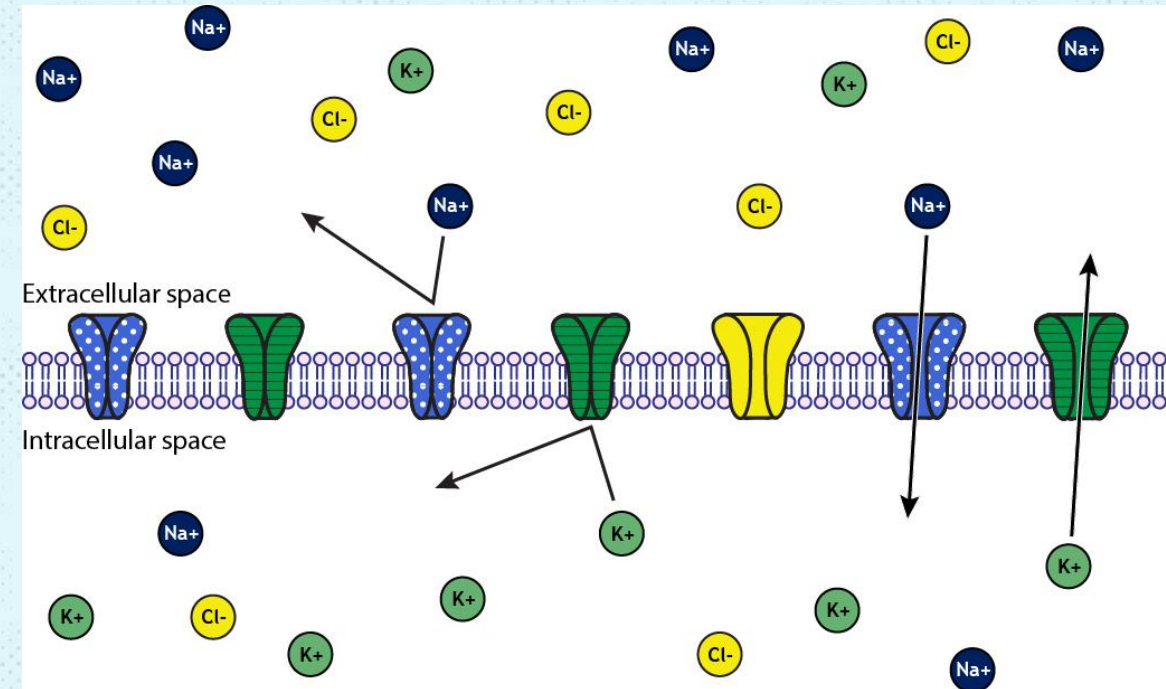
Patch-clamp electrophysiology



Leica

Electrical properties of neurons

- Lipid bilayer acts as capacitor separating charges
- Ion channels with pores allow certain ions to pass often depending on voltage or ligands
- Ion pumps maintain ion gradient
- Membrane potential V_T outside is defined to be 0
- V_T is between -90 and +50 mV



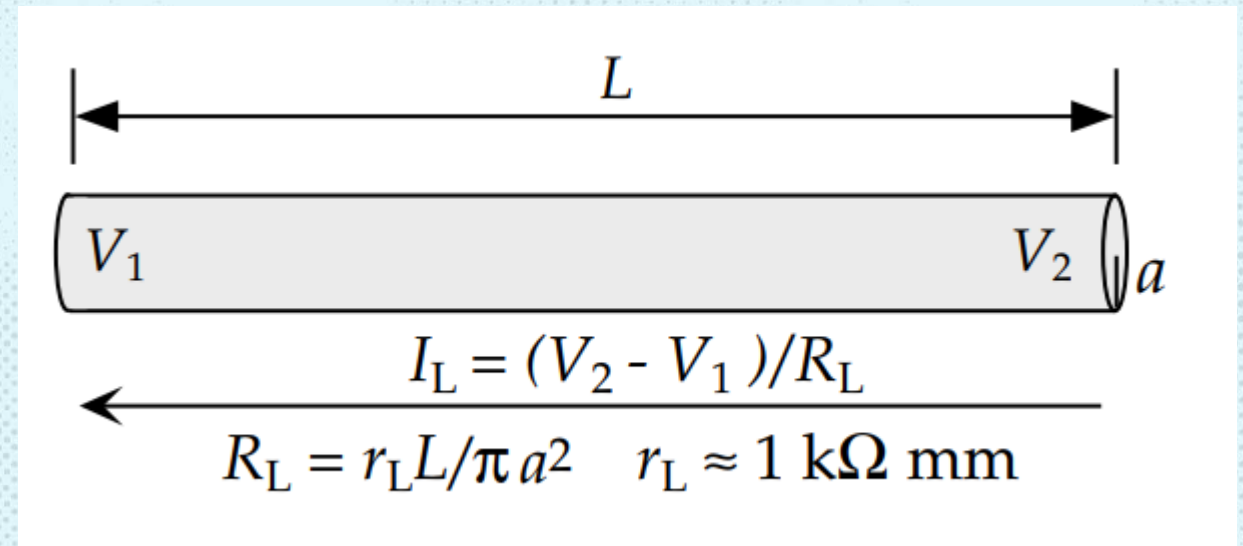
Casey Henly, CC-by-NC-SA

Intracellular resistance

- Membrane potential can vary across dendrite / axons

→ ions flow inside the cell
against resistance of intracellular
medium

- Resistance: $\sim L$
 $\sim 1/a^2$

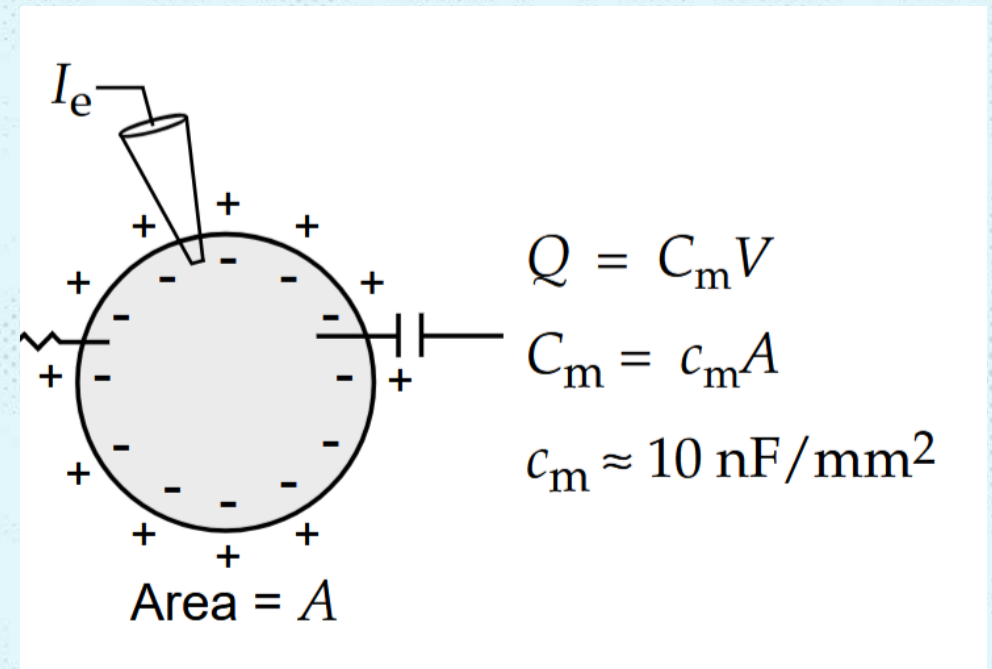


Dayan & Abbott, 2001

Membrane capacitance & resistance

- Electrotonically compact neuron:
Single compartment
- How much current is required to
change the membrane potential?

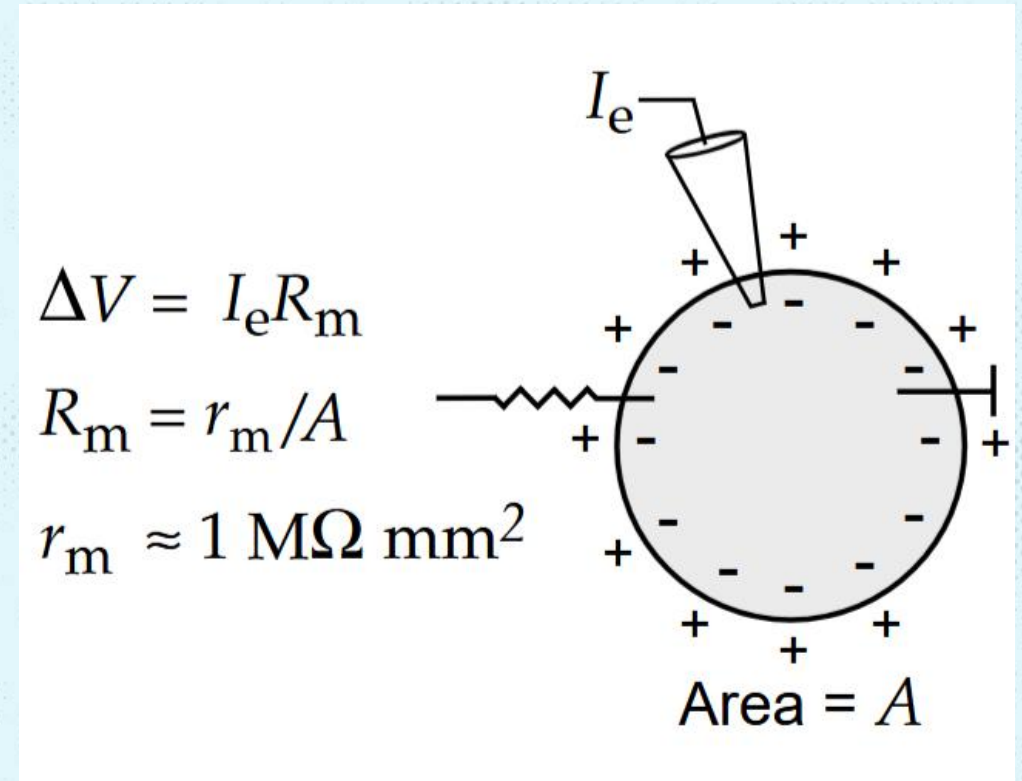
$$C_m \frac{dV}{dt} = \frac{dQ}{dt}$$



Dayan & Abbott, 2001

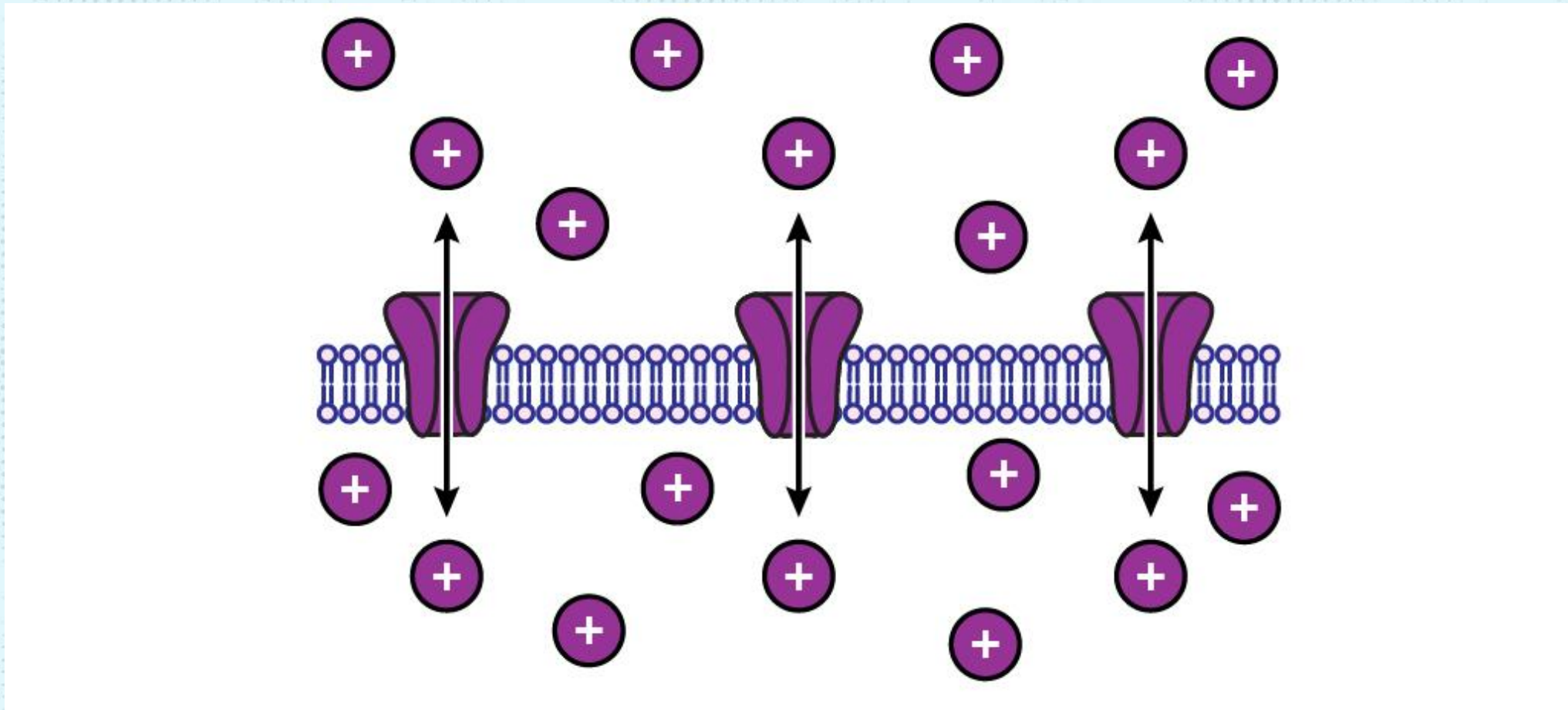
Membrane capacitance & resistance

- How much current is required to keep the membrane potential at a level different than its resting value?
- Determined by membrane resistance
- Membrane time constant:
 $\tau_m = r_m C_m$

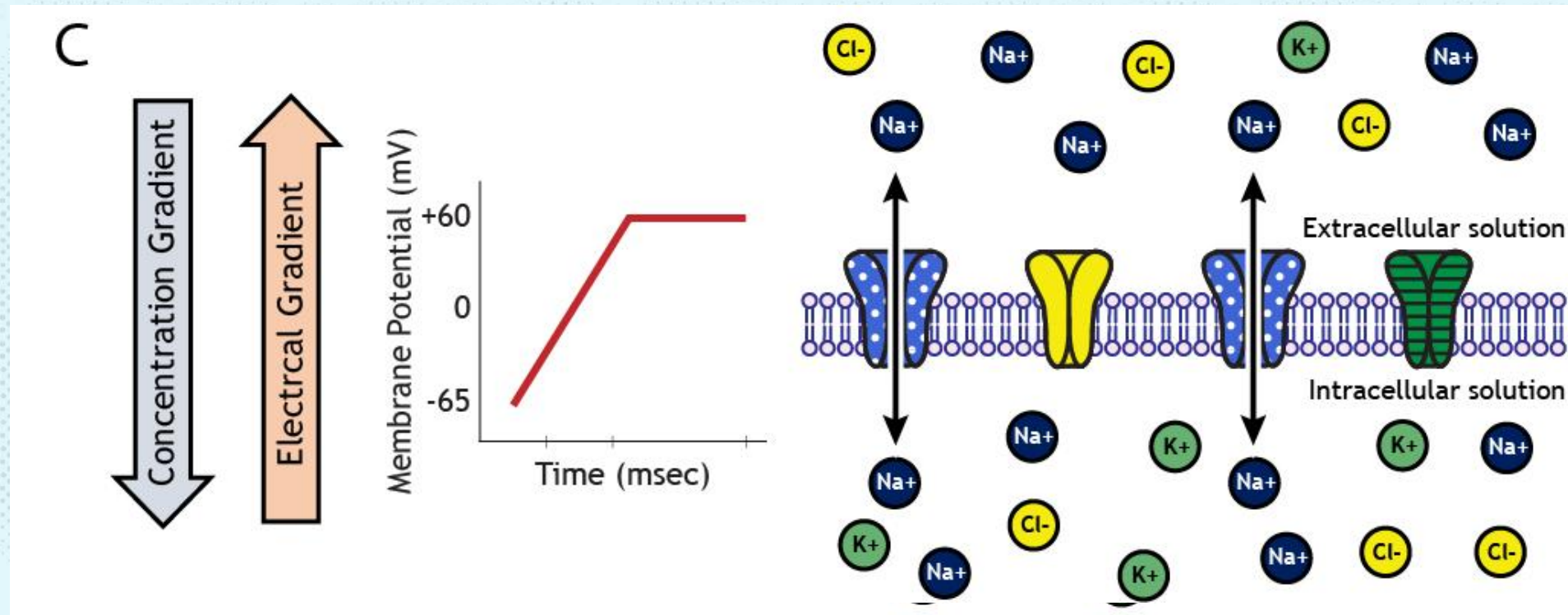


Dayan & Abbott, 2001

Equilibrium potential



Equilibrium potential of Sodium



Casey Henley, CC-by-NC-SA

Equilibrium potentials

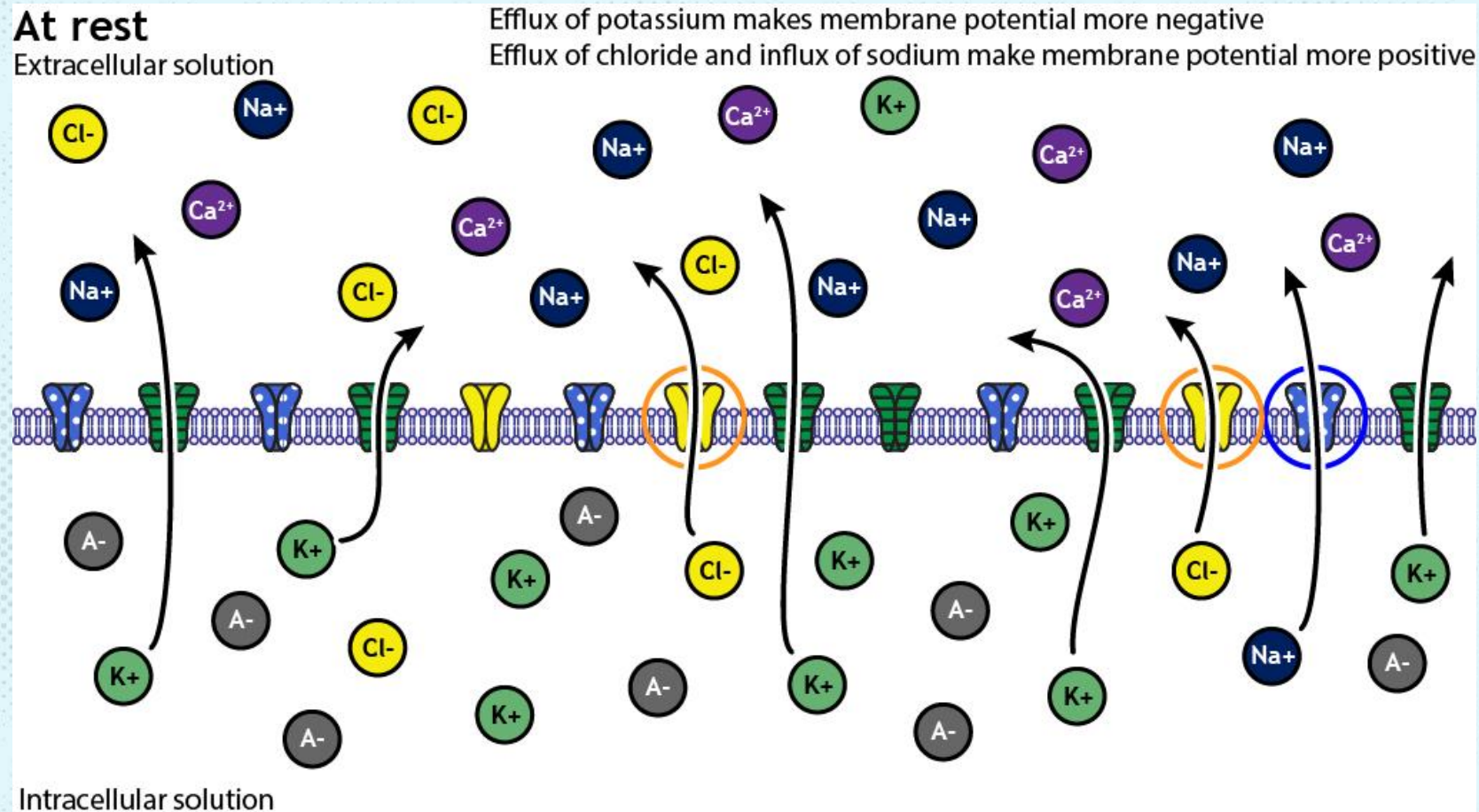
- Nernst equation:

$$E = \frac{V_T}{z} \ln \left(\frac{[\text{outside}]}{[\text{inside}]} \right)$$

- $E_{Na} = +50 \text{ mV}$
- $E_K = -70 \text{ mV}$

- $V > E$: positive current will flow outward
- $V < E$: positive current will flow inward
- What is the effect of Na-channels?
- What of K-channels?

Resting potential

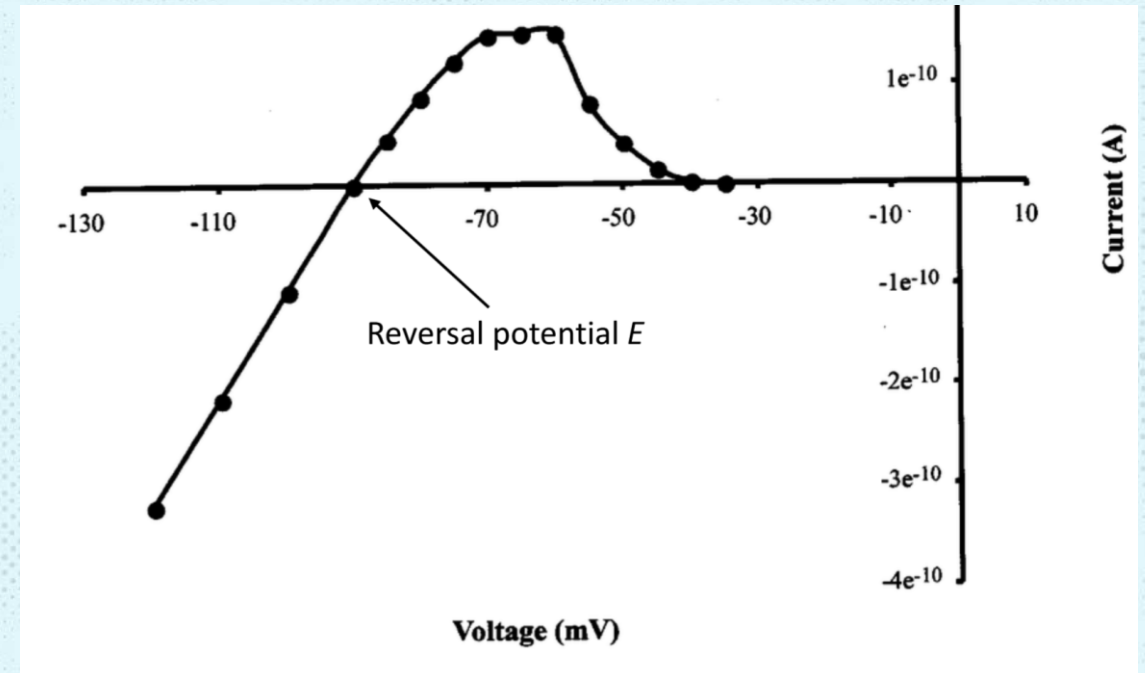


Membrane current

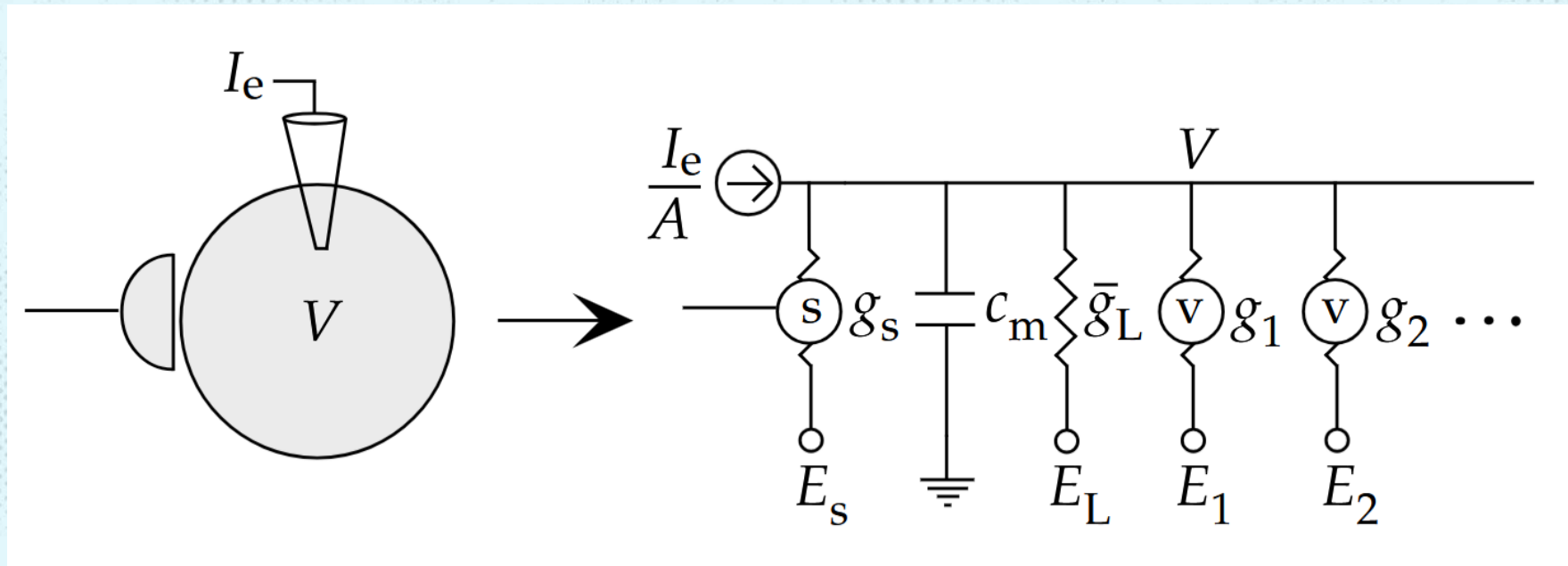
- Total flow of ions across the membrane

$$i_m = \sum_i g_i (V - E_i)$$

- g_i conductance, may change over time!
- g_l “leak”, summarizes ion-pumps, passive channels....



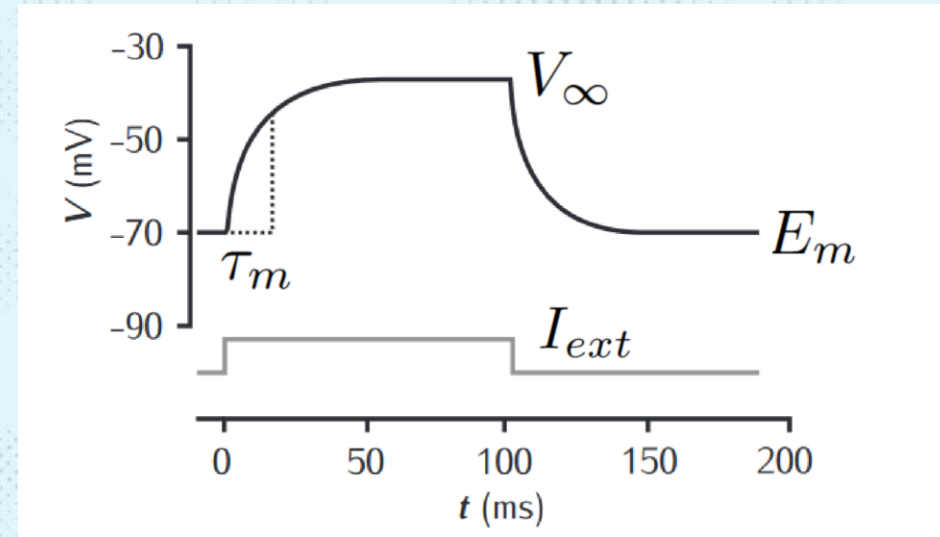
Equivalent circuit model of a single compartment neuron



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

Passive neuron model

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

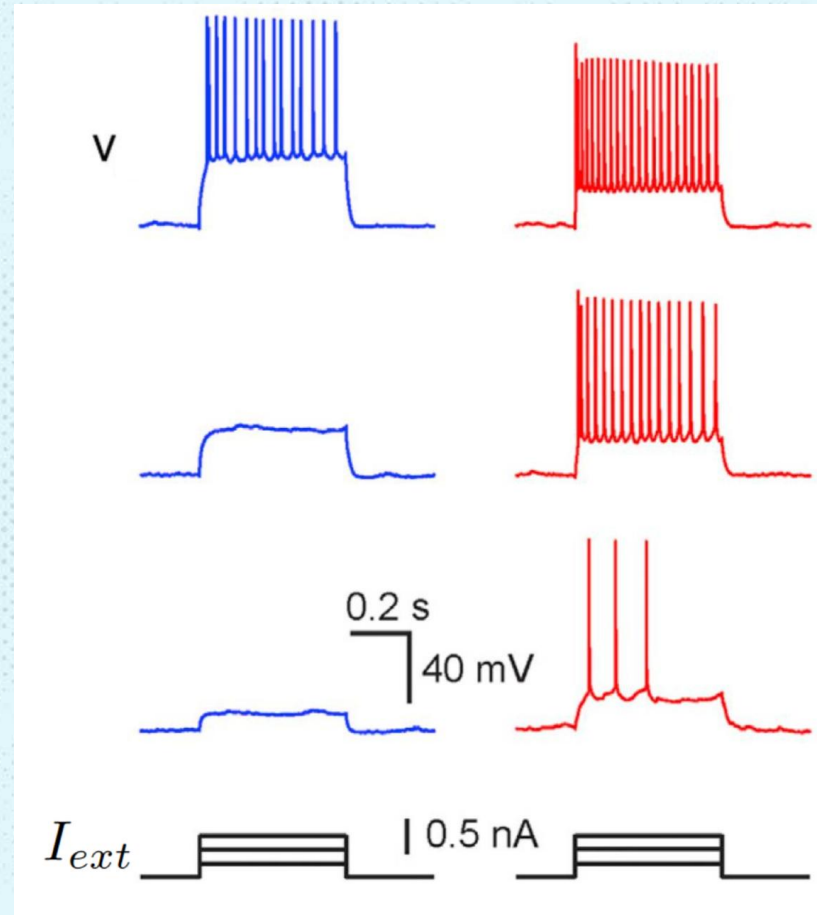


Angus Chadwick

$$V(t) - E_m = \boxed{e^{-t/\tau_m} (V(0) - E_m)} + \boxed{\frac{1}{g_m \tau_m} \int_0^t e^{-(t-t')/\tau_m} I_{ext}(t') dt'}$$

Decay of initial membrane potential
towards resting potential

Low-pass filter of external current input
(also called a “leaky integrator”)



Angus Chadwick