



# 2 Biological neurons and their functional engineering models

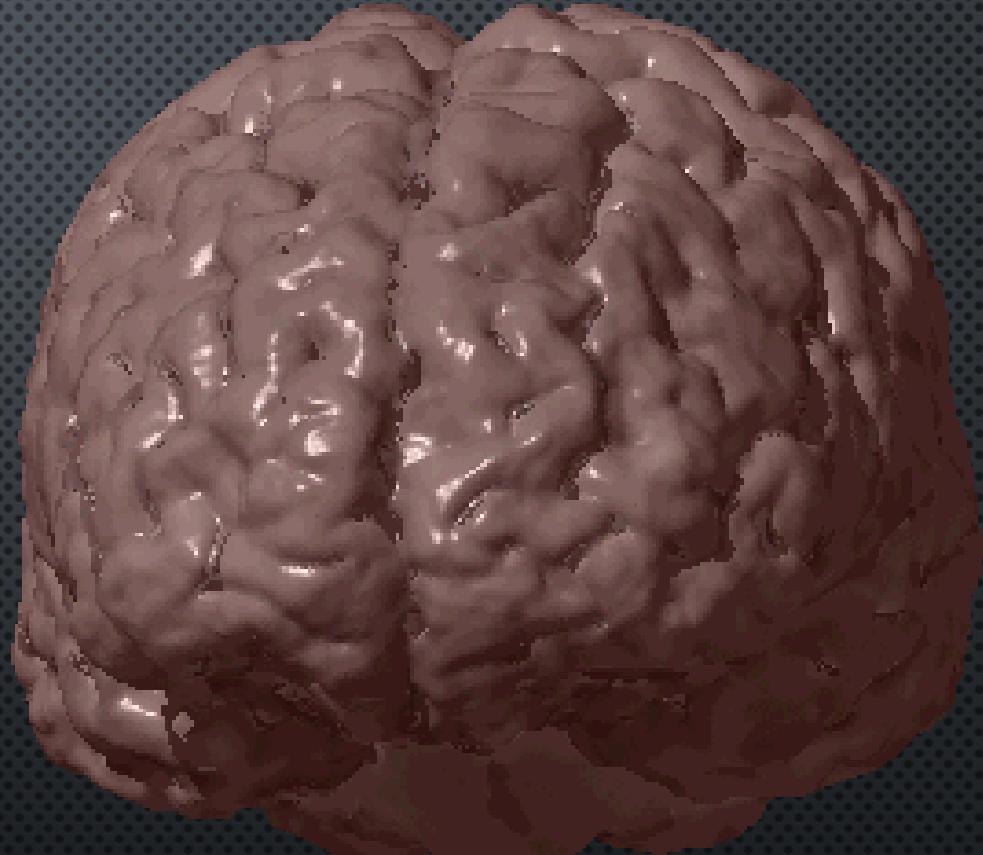


Neurobiology ahead  
(get used to it)

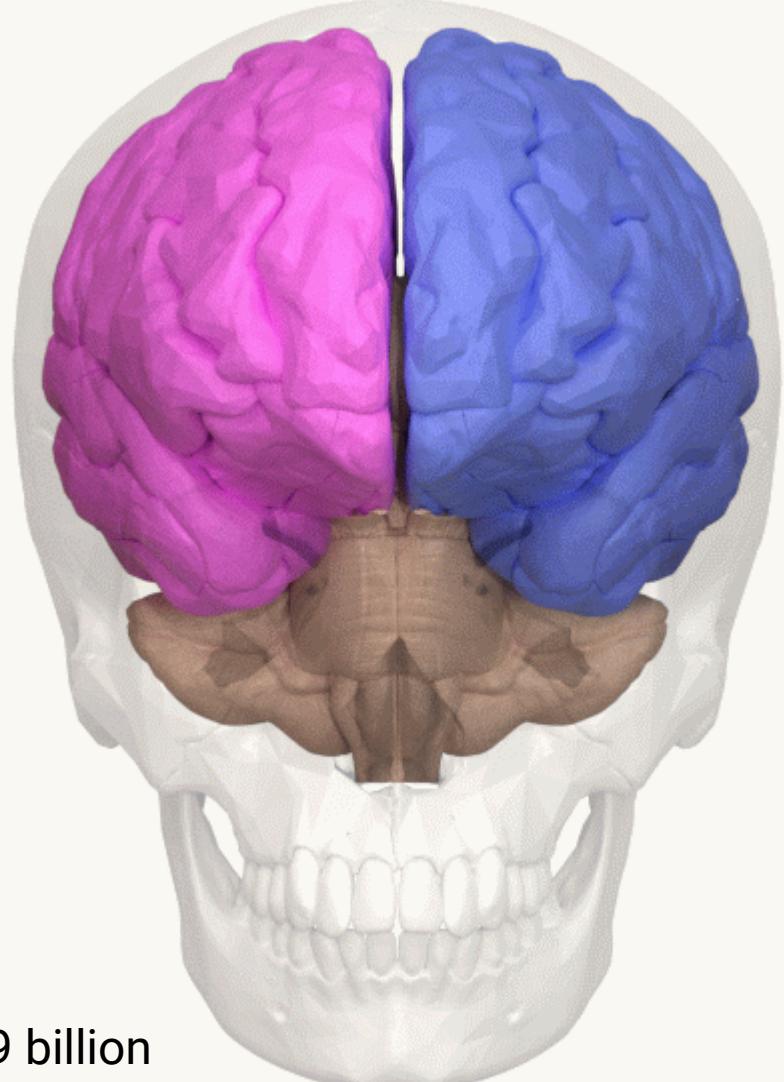


# The human brain

- Represents only **2% of the body weight** (1.2 - 1.4 kg) but receives 15% of the cardiac output, 20% of total body oxygen consumption, and 25% of total body glucose utilization
- Composed of neurons, glial cells, neural stem cells, and blood vessels
- Contains about  **$86 \pm 8$  billion neurons** and  **$85 \pm 10$  billion non-neuronal cells**

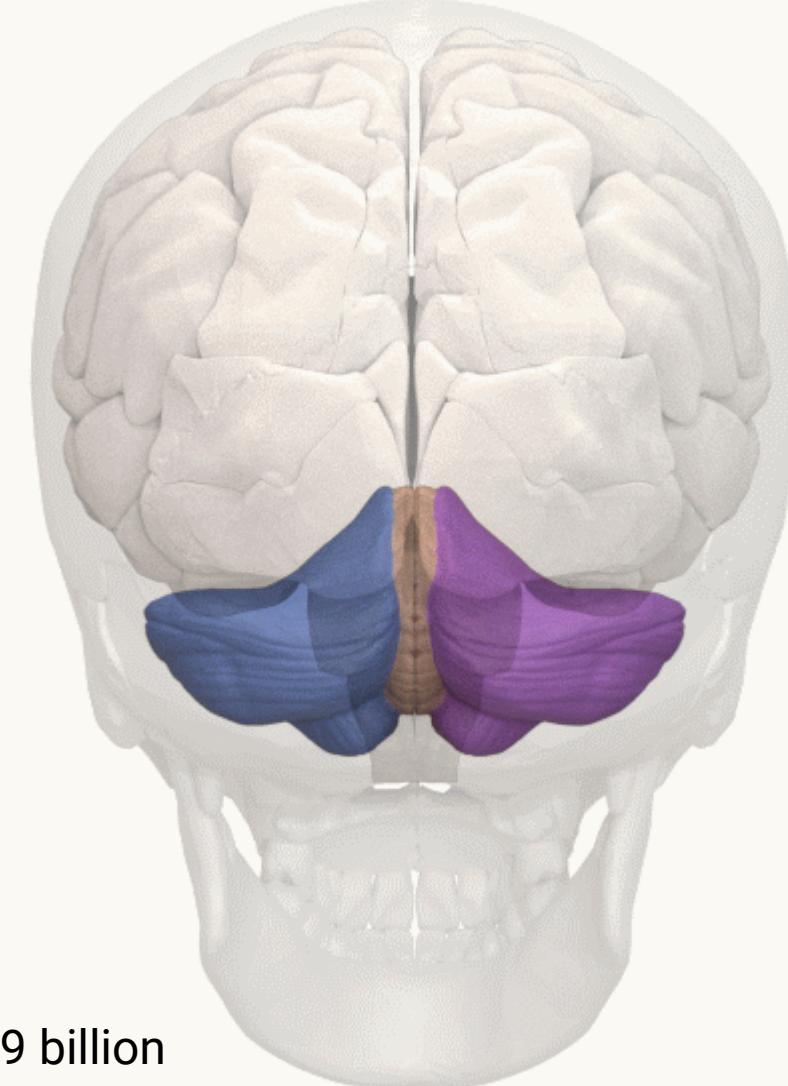


# What makes the brain so special?



About 19 billion  
(~19%) neurons

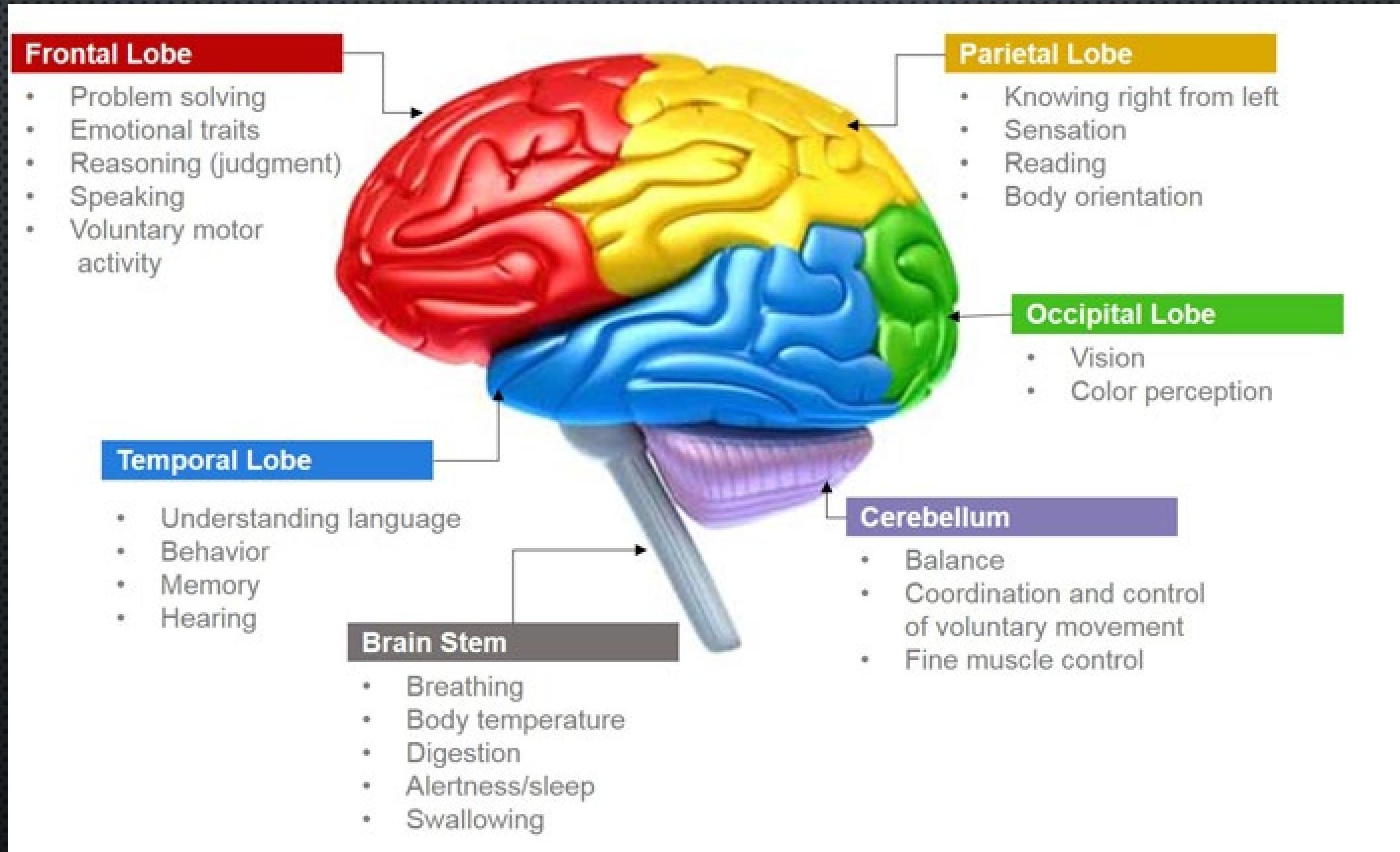
Cerebrum



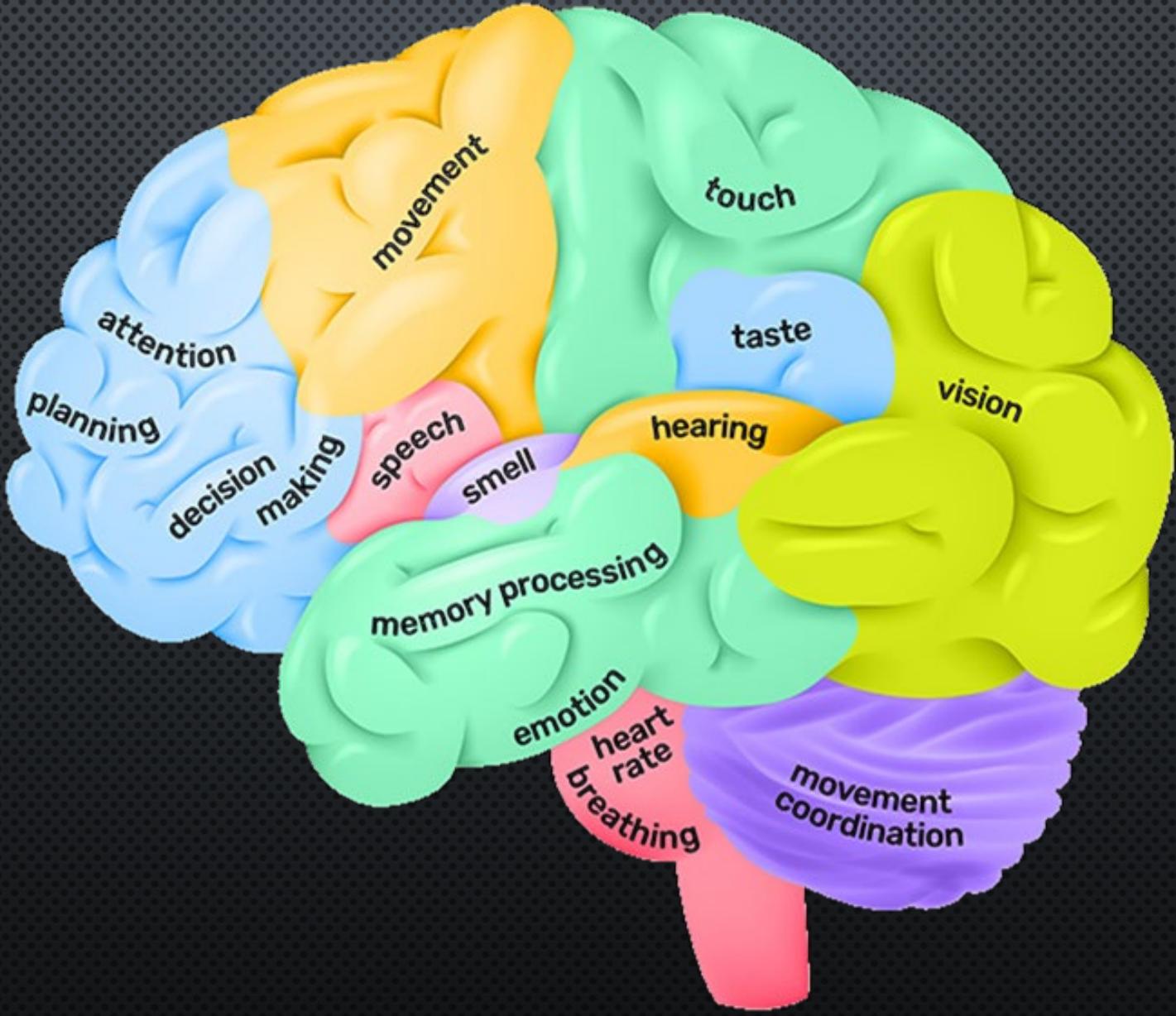
About 69 billion  
(~80%) neurons

Cerebellum

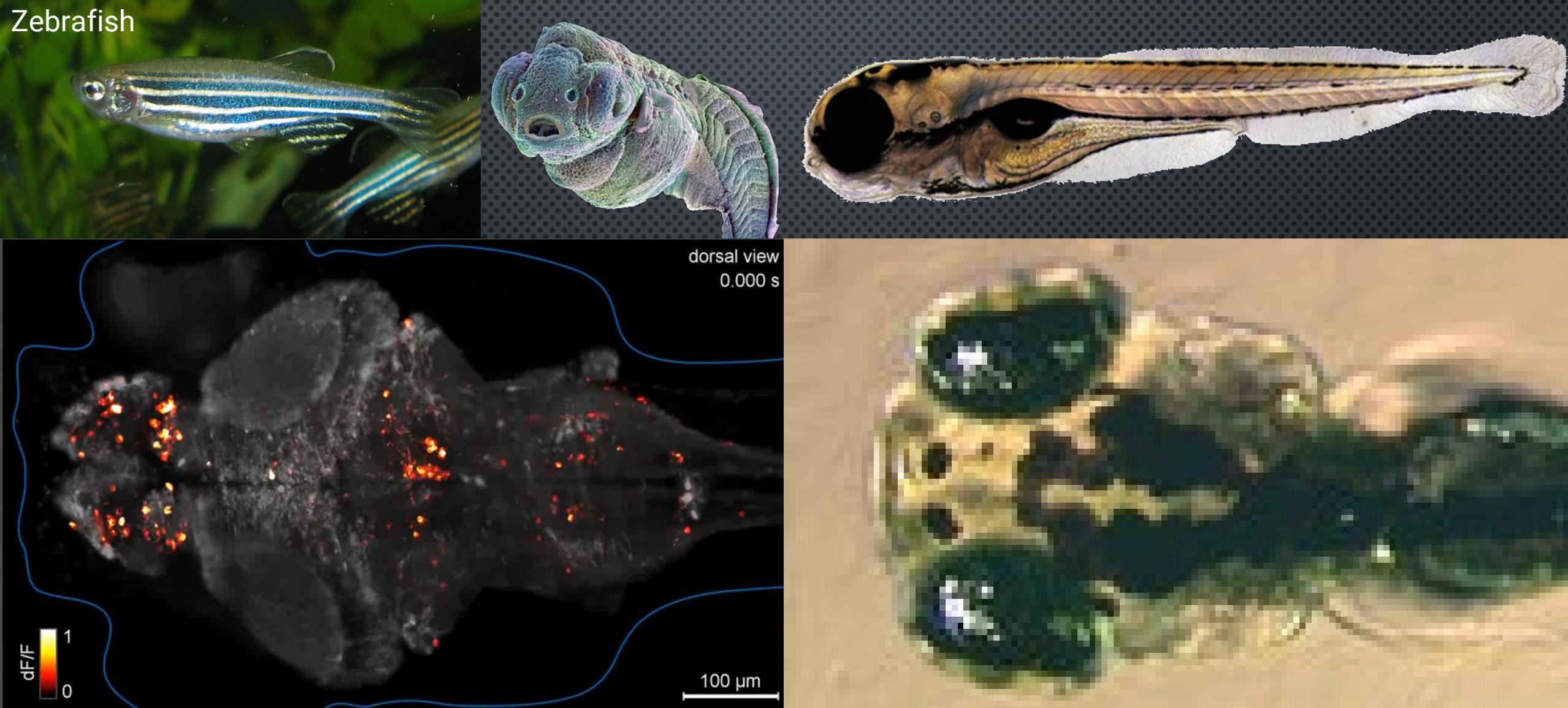
# Brain anatomy and function



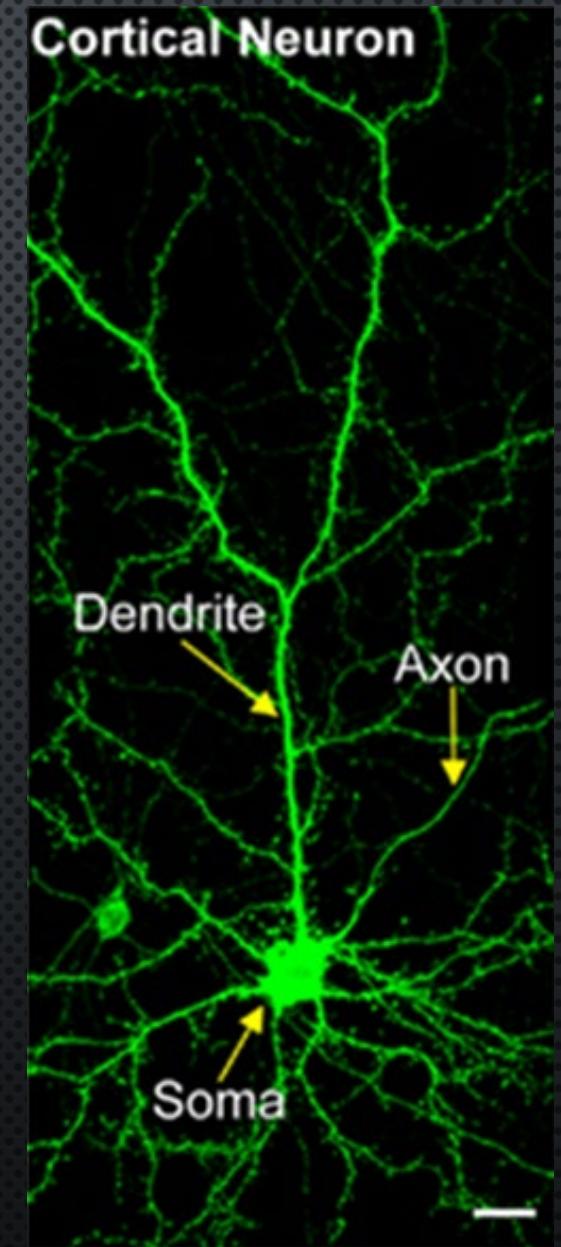
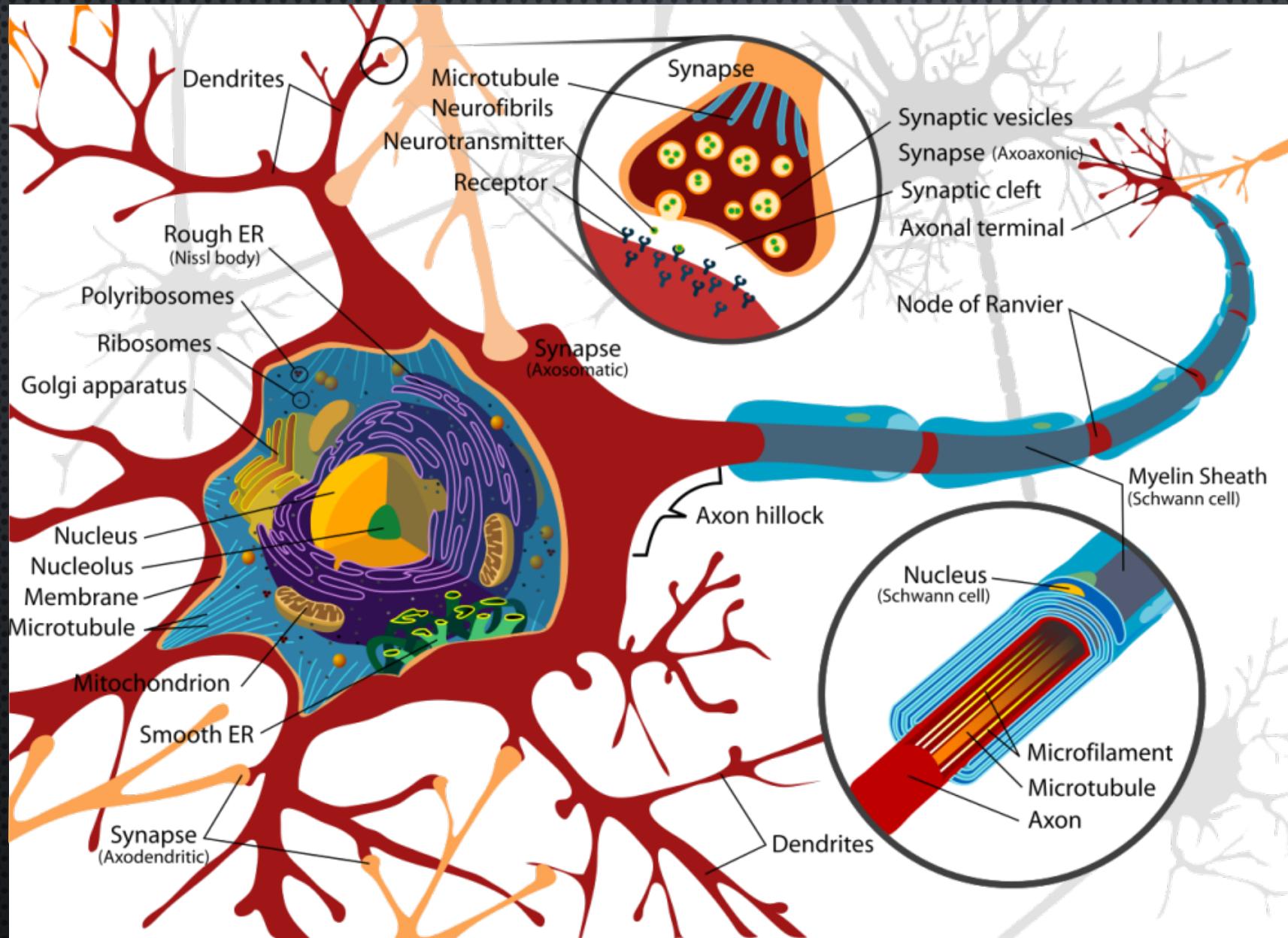
# Implications for robotics



# Visualising the brain



# What is a neuron?



# Basic types of neurons



## Unipolar

Transmit signals from outside of body to brain (found in insect brains)



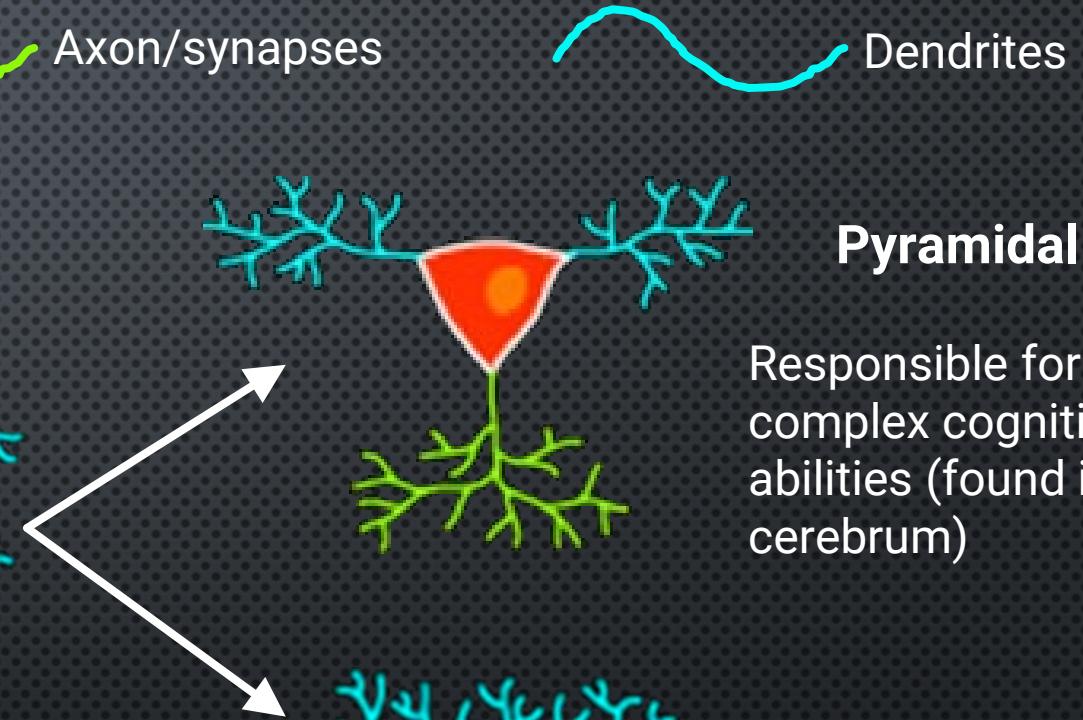
## Bipolar

Transmit signals from sensory appendages (eyes, ears etc.) to brain



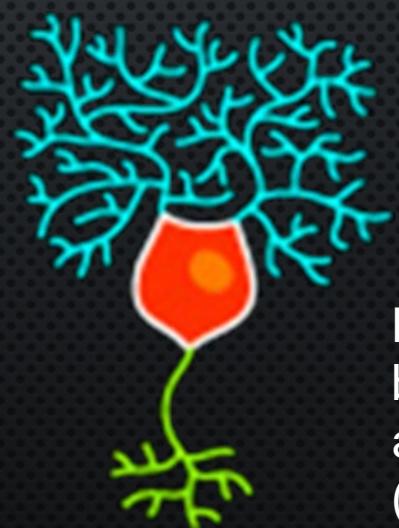
## Multipolar

Transmit signals within the brain as well as from brain to muscles



## Pyramidal

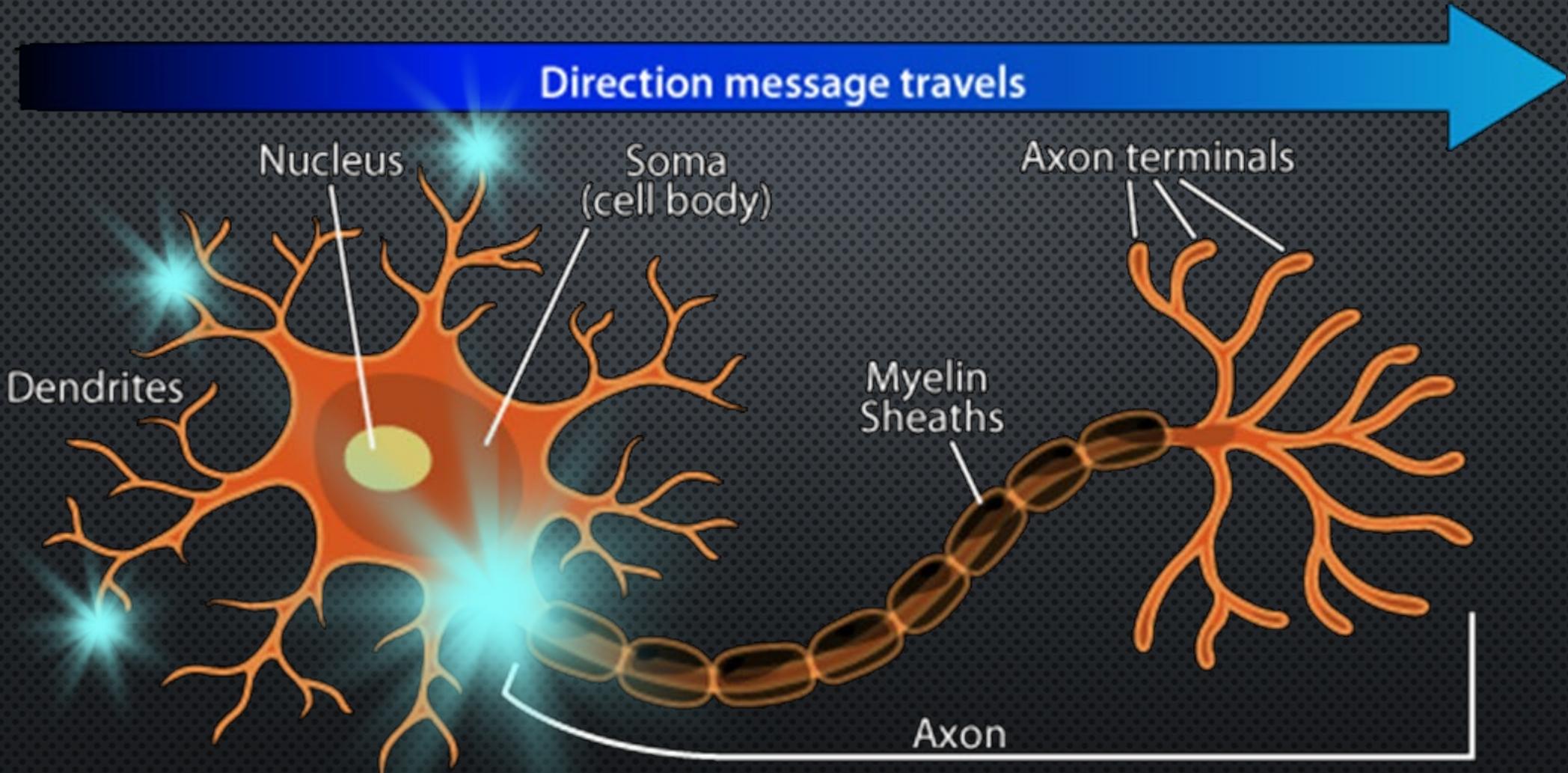
Responsible for complex cognitive abilities (found in cerebrum)



## Purkinje

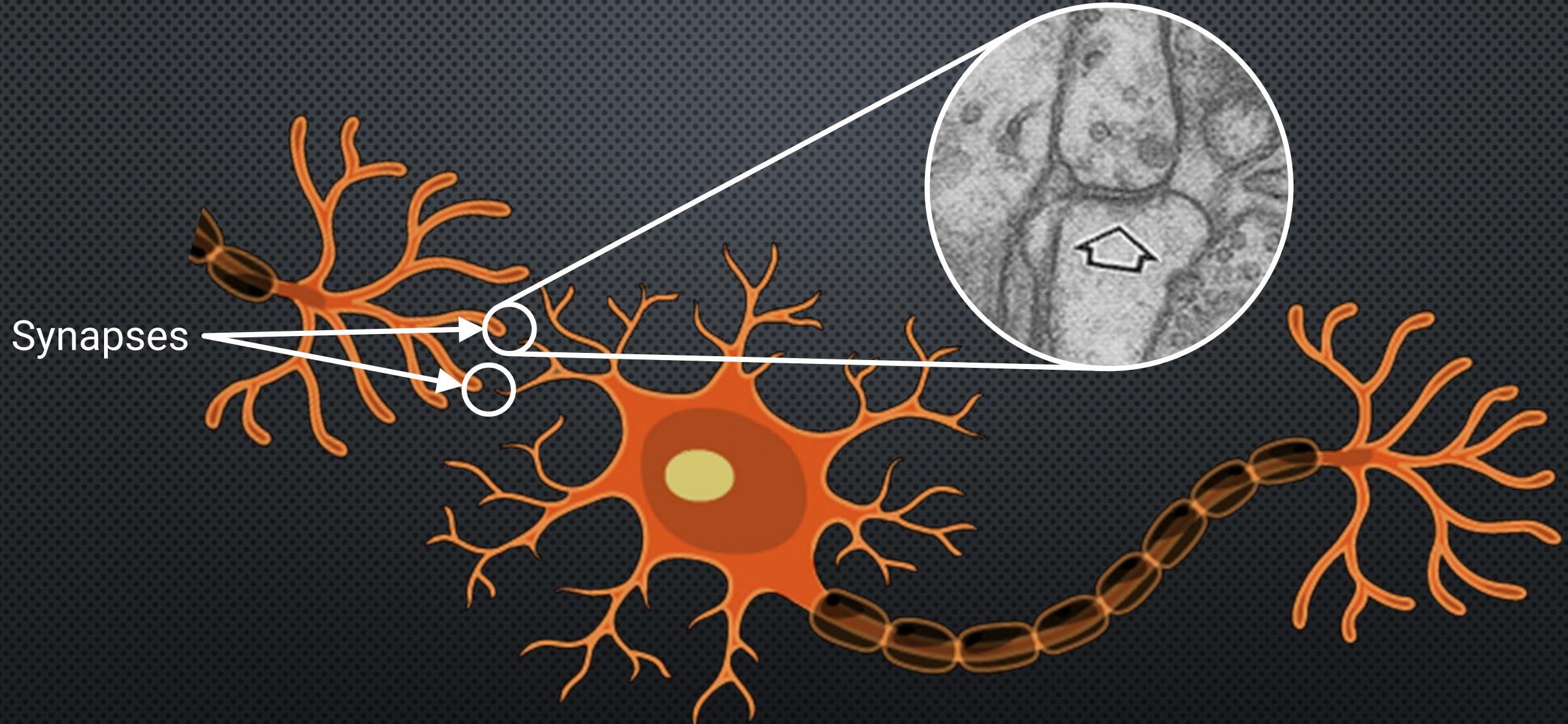
Responsible for balance, coordination and timing of actions (found in cerebellum)

# Neurons integrate and transmit information



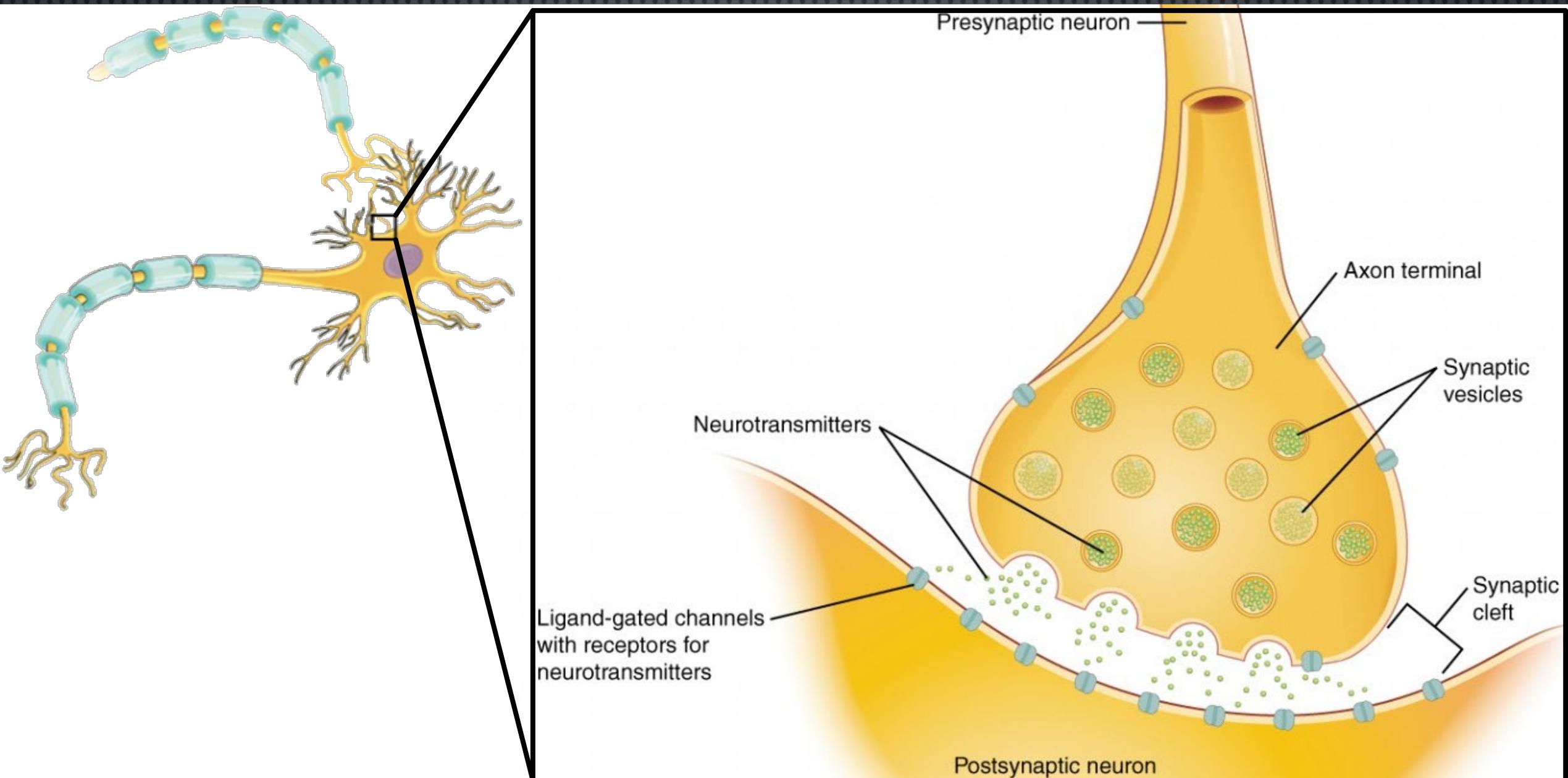
Neurons collect electrical signals (called **action potentials**) from other neurons via **synapses** and transmit them to other neurons via the axon

# Synapses bridge neurons together

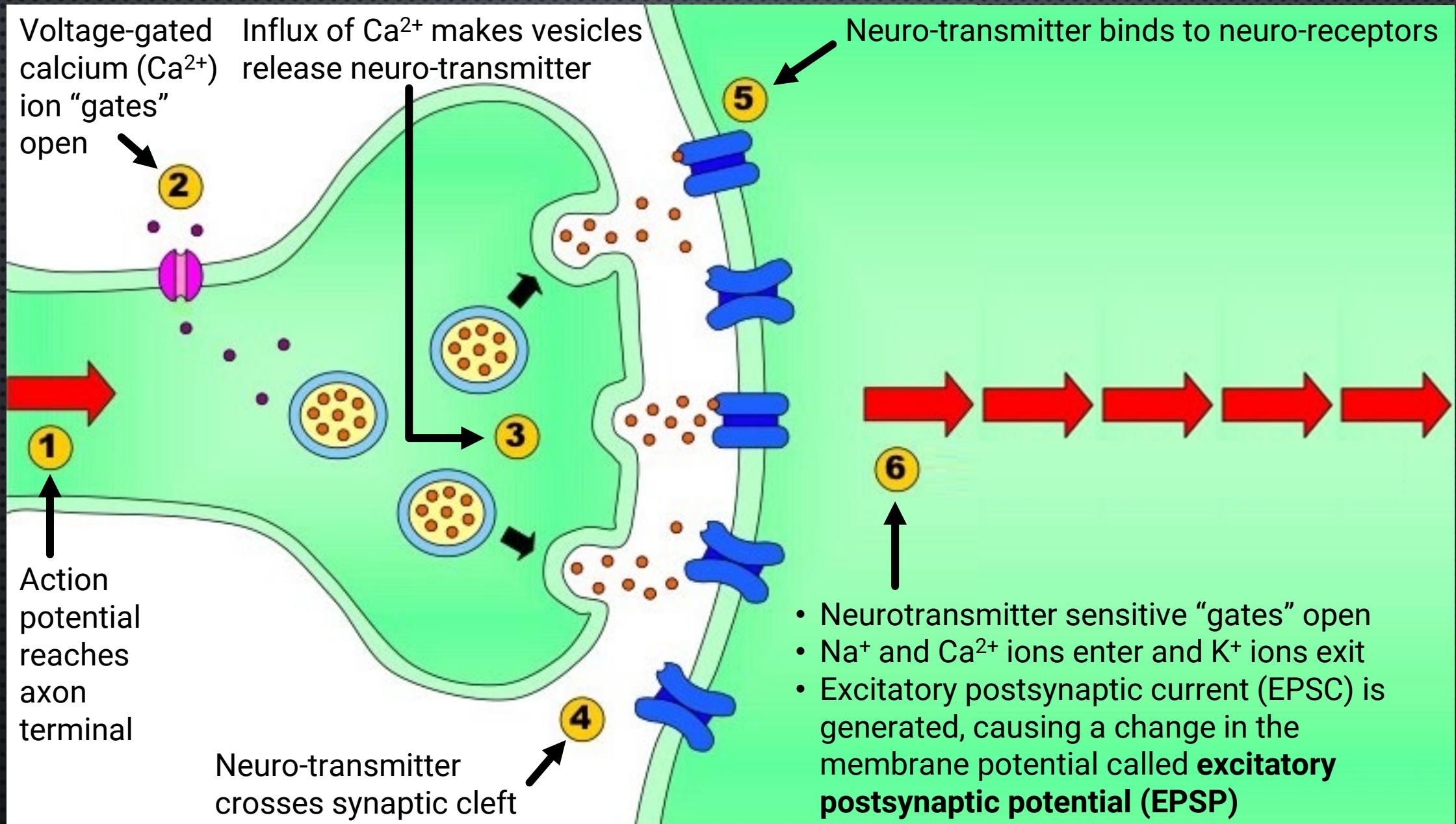


A **synapse** is a structure that permits a neuron to pass an **electrical or chemical signal** to another neuron.

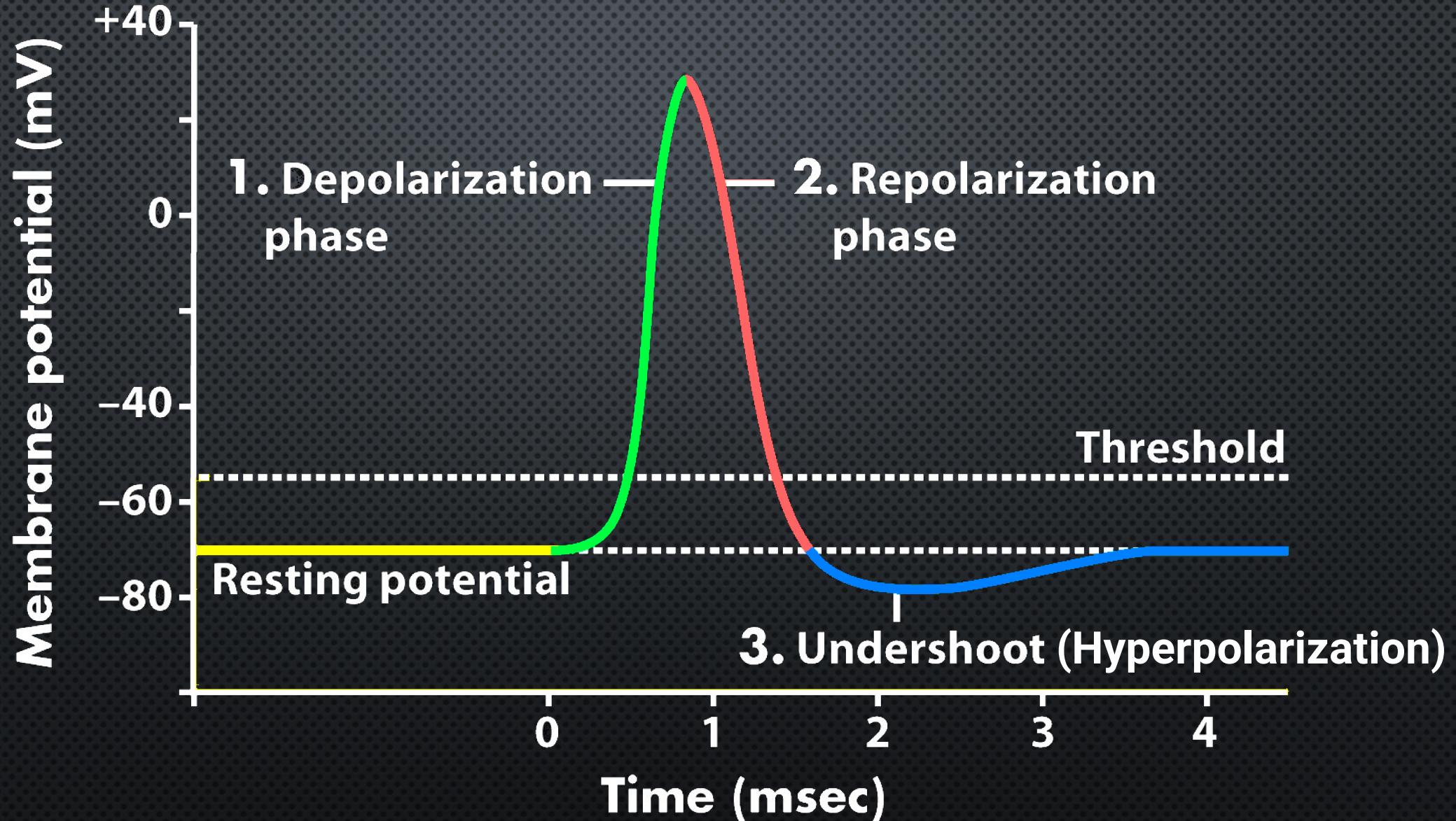
# Structure of a chemical synapse



# Signal transmission in synapses

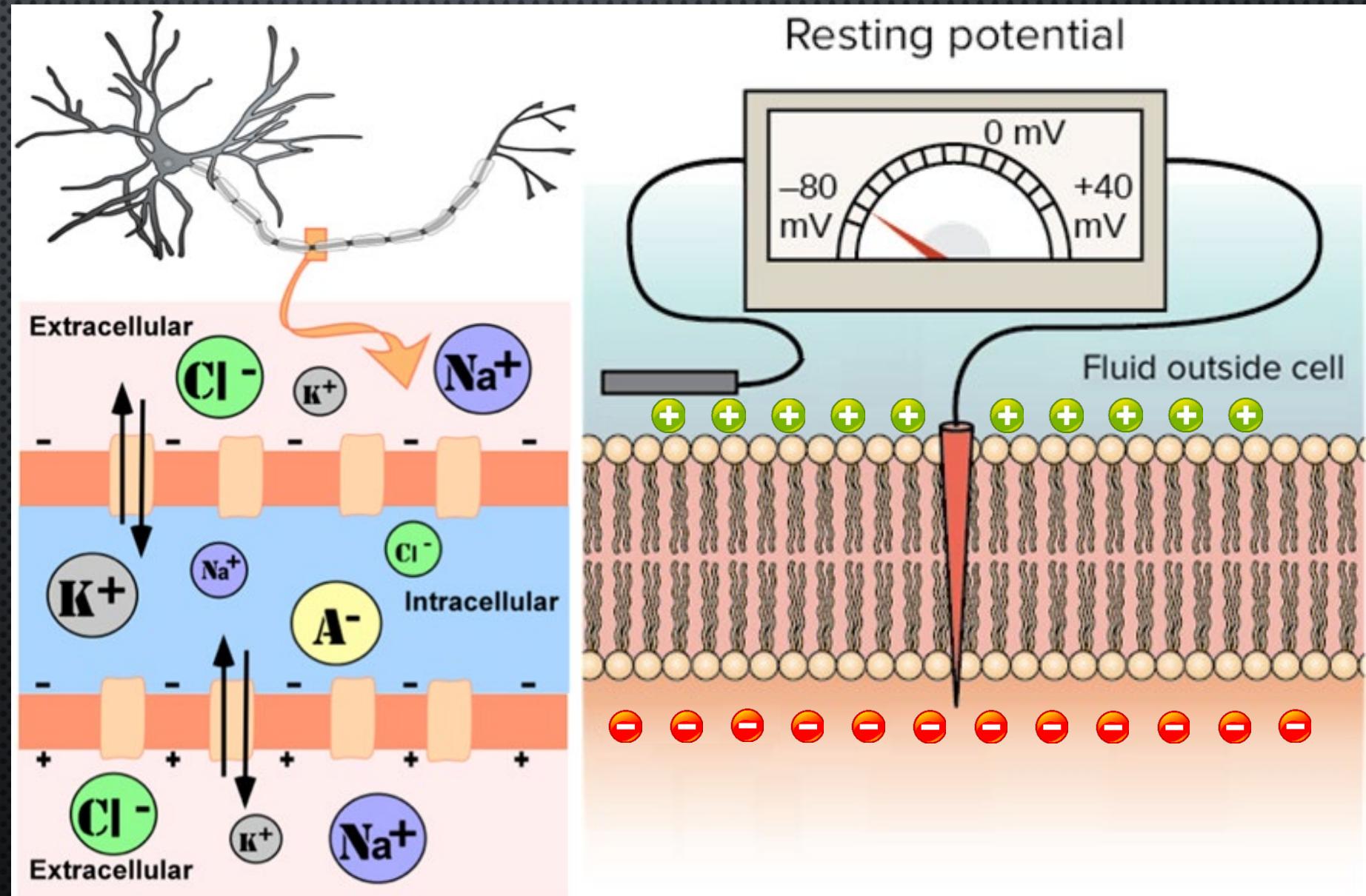


# Action potential

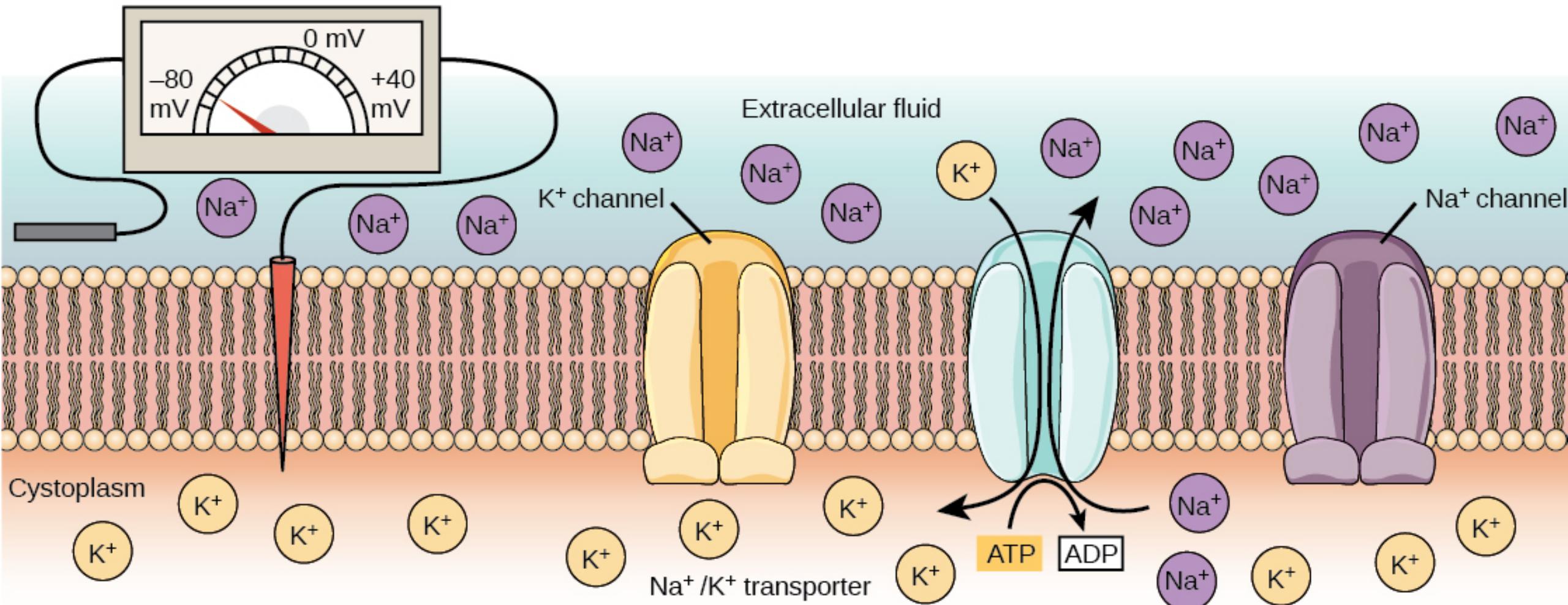


# Resting membrane potential: default state of a neuron

- Difference between electric potential inside and outside a neuron, due to different ion concentration
- Potassium ( $K^+$ ) ions have higher concentration inside
- Sodium ( $Na^+$ ) ions have higher concentration outside
- Leakage of  $K^+$  to outside leaves excess negative charge inside and excess positive charge outside
- Ion leakage sets resting membrane potential to around  $-65\text{ mV}$  to  $-70\text{ mV}$

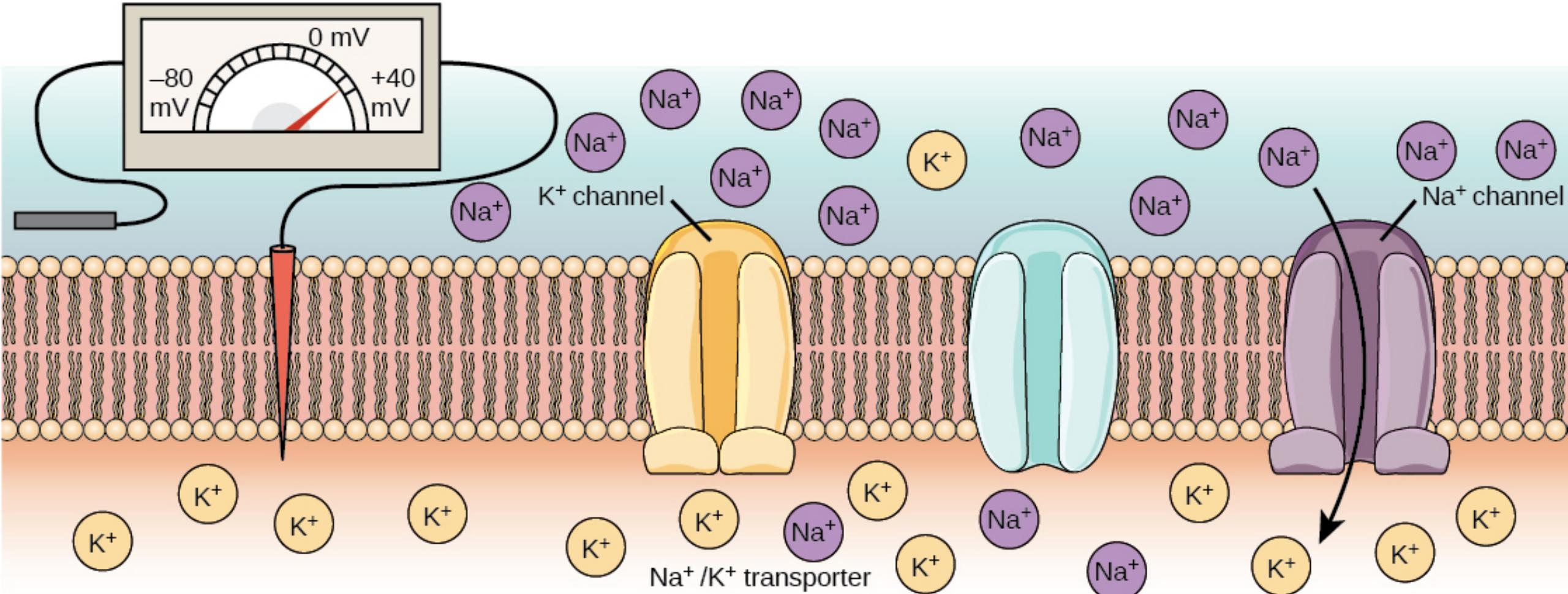


# Resting membrane potential: default state of a neuron



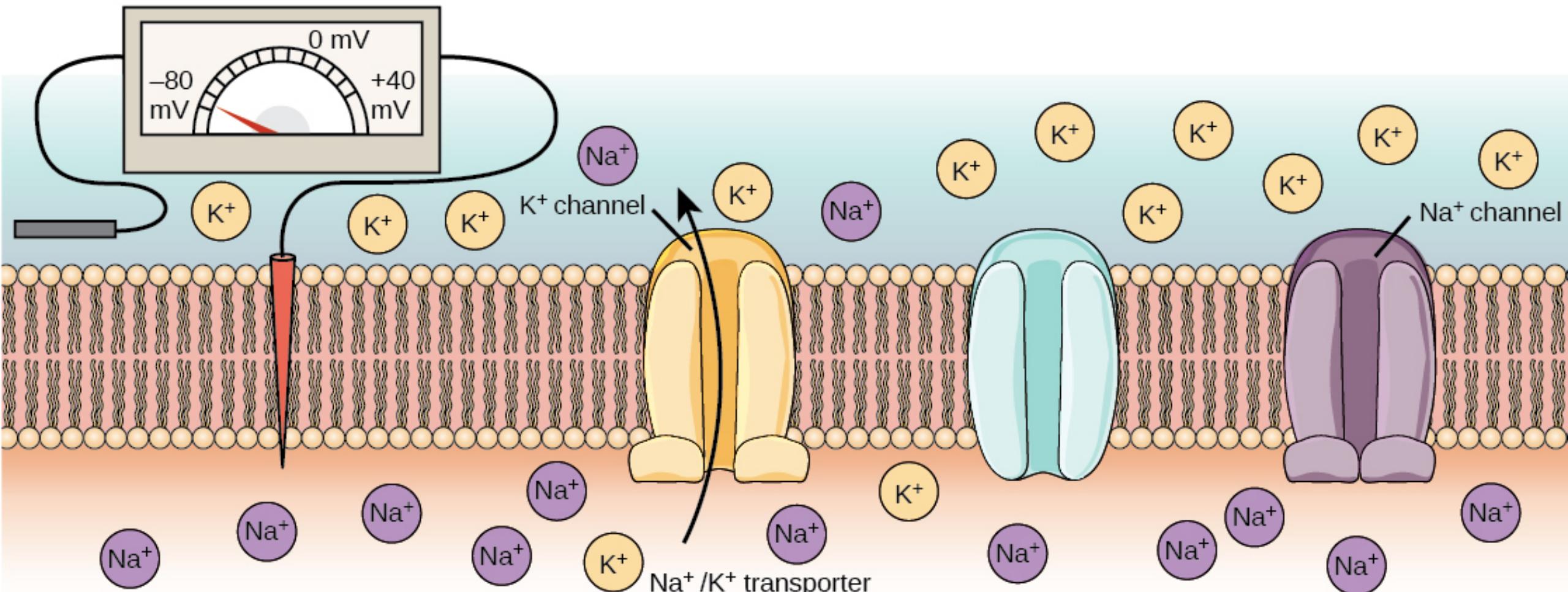
At the resting potential, all voltage-gated  $\text{Na}^+$  channels and most voltage-gated  $\text{K}^+$  channels are closed. The  $\text{Na}^+/\text{K}^+$  transporter pumps  $\text{K}^+$  ions into the cell and  $\text{Na}^+$  ions out.

# De-polarisation



In response to a depolarization, some  $Na^+$  channels open, allowing  $Na^+$  ions to enter the cell. The membrane starts to depolarize (the charge across the membrane lessens). If the threshold of excitation is reached, all the  $Na^+$  channels open.

# Hyper-polarisation

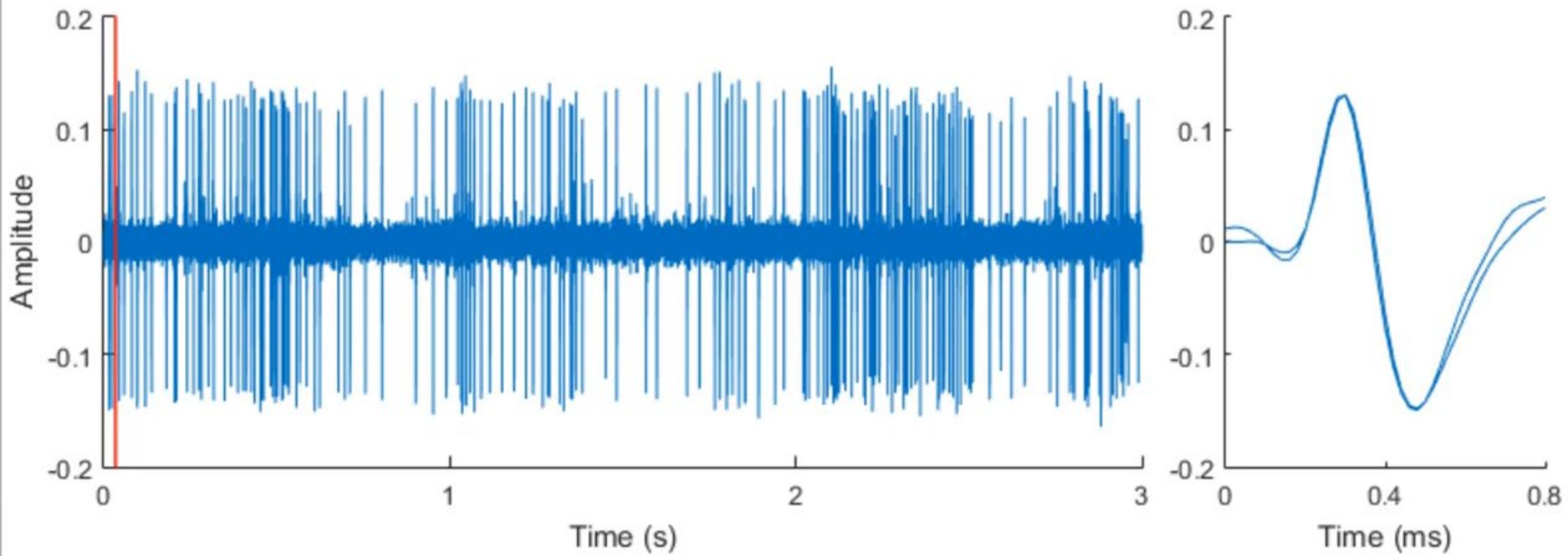


At the peak action potential,  $\text{Na}^+$  channels close while  $\text{K}^+$  channels open.  $\text{K}^+$  leaves the cell, and the membrane eventually becomes hyperpolarized.

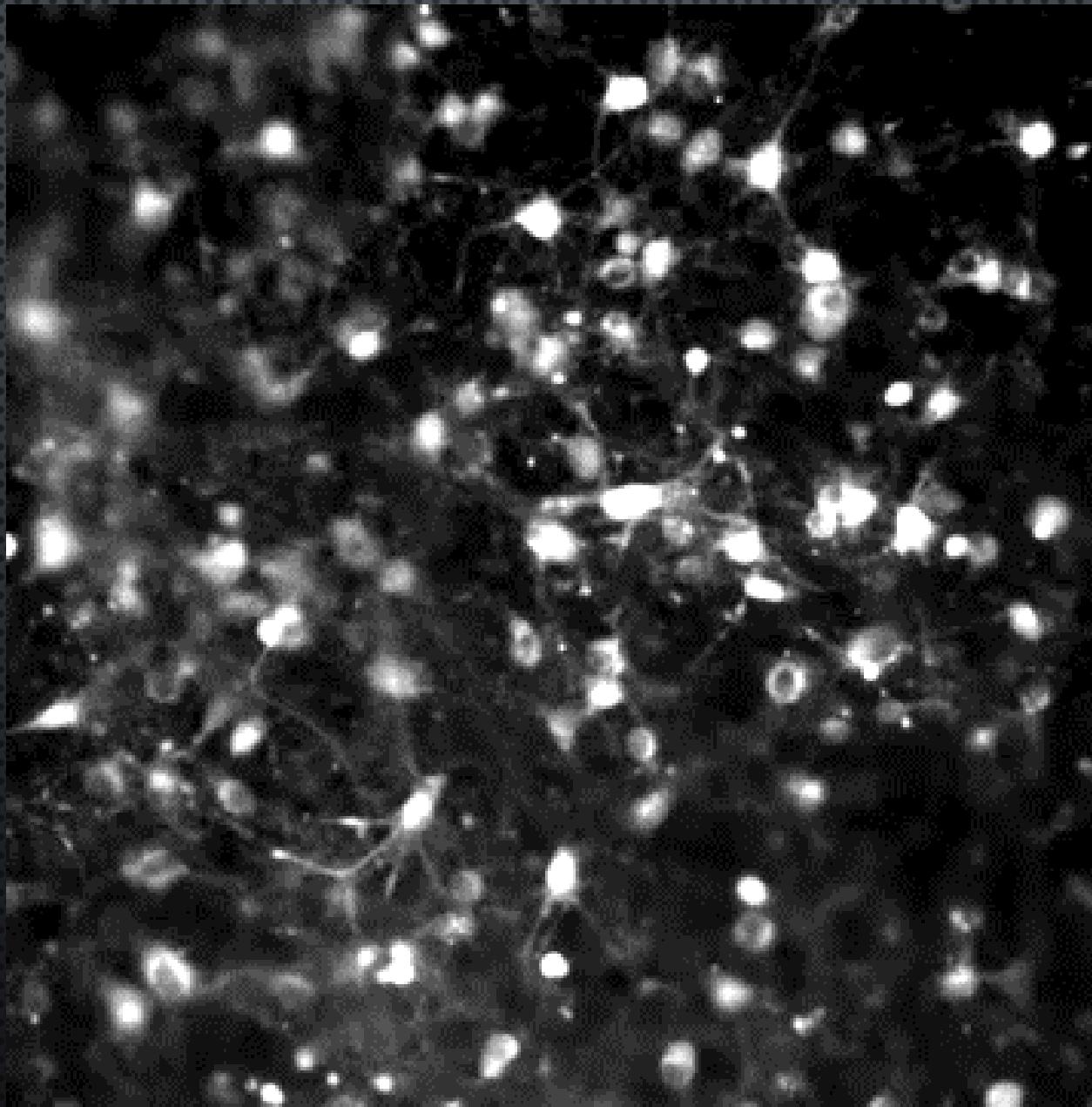
# How is action potential actually measured?



# Sights and sounds of neuronal recordings

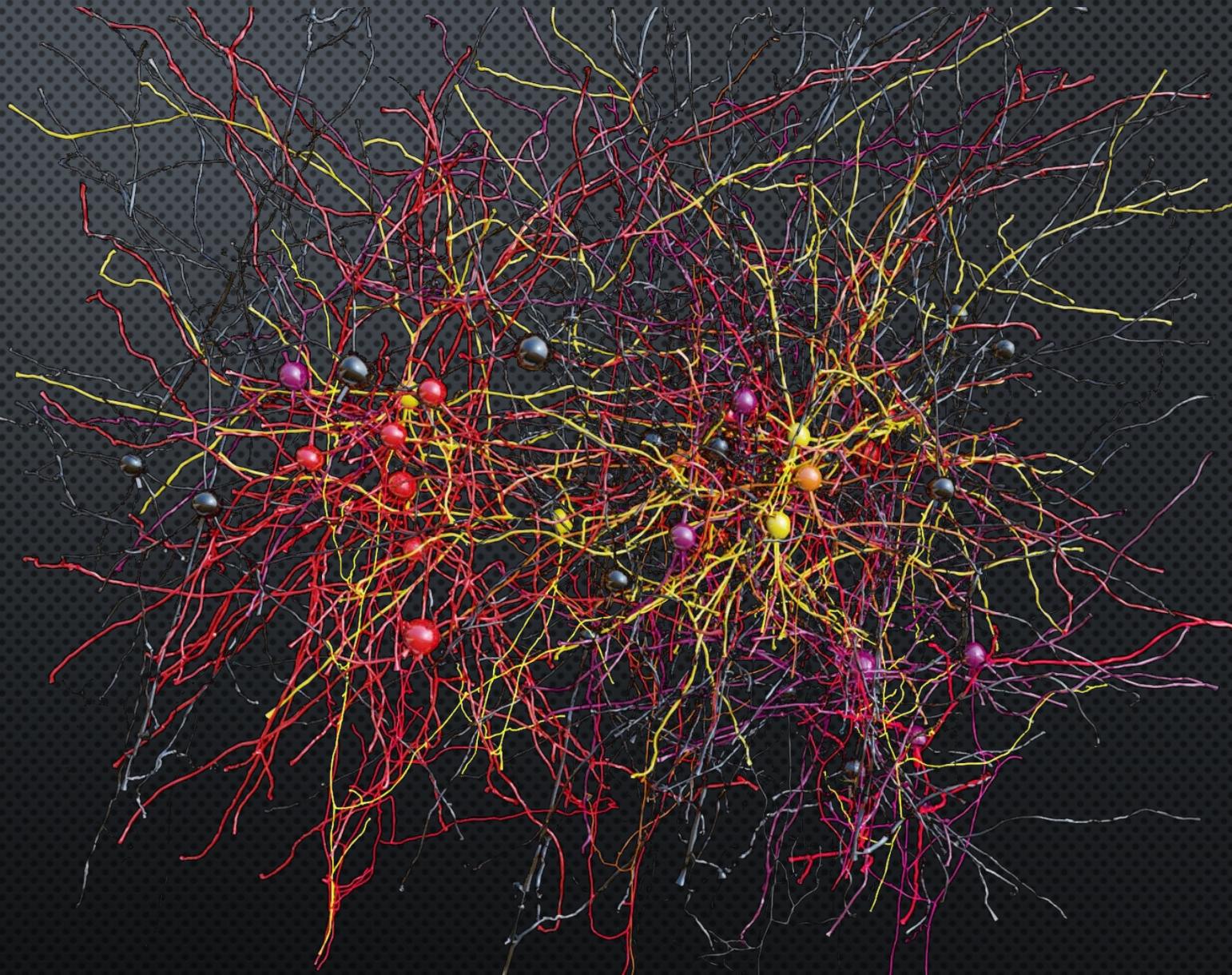


# Neuronal firing

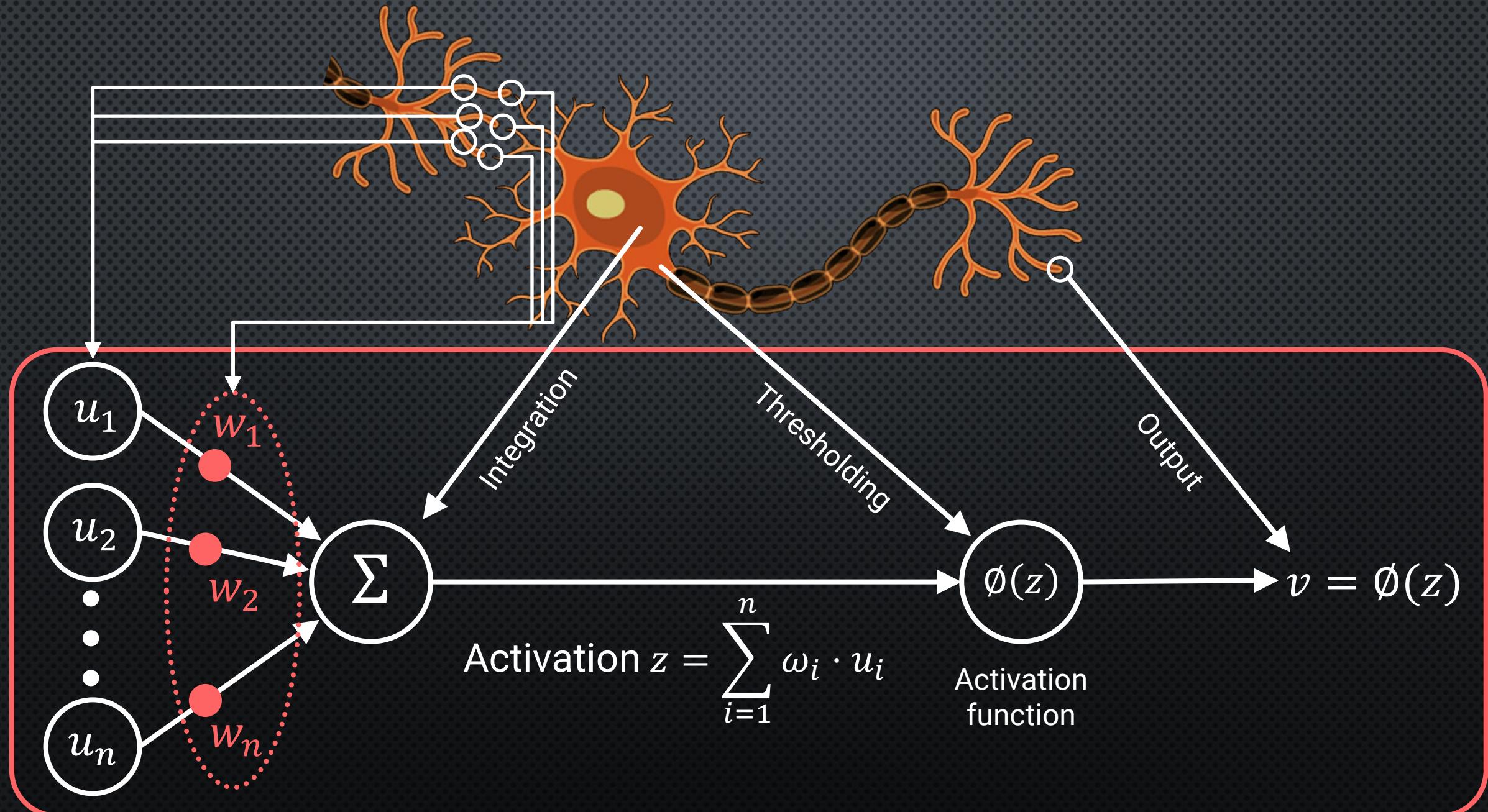


# Neural connectivity is very complex!

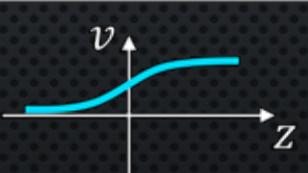
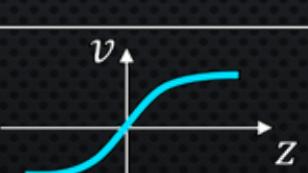
- A typical neuron may receive inputs from on the order of 10,000 neurons
- Pyramidal cells in the hippocampus can receive on the order of 50,000 inputs
- There are an estimated **1000 trillion** synapses in the human brain
- Although neural connectivity is highly dense, **the brain is not a fully connected network**



# Perceptrons: from real to artificial neurons

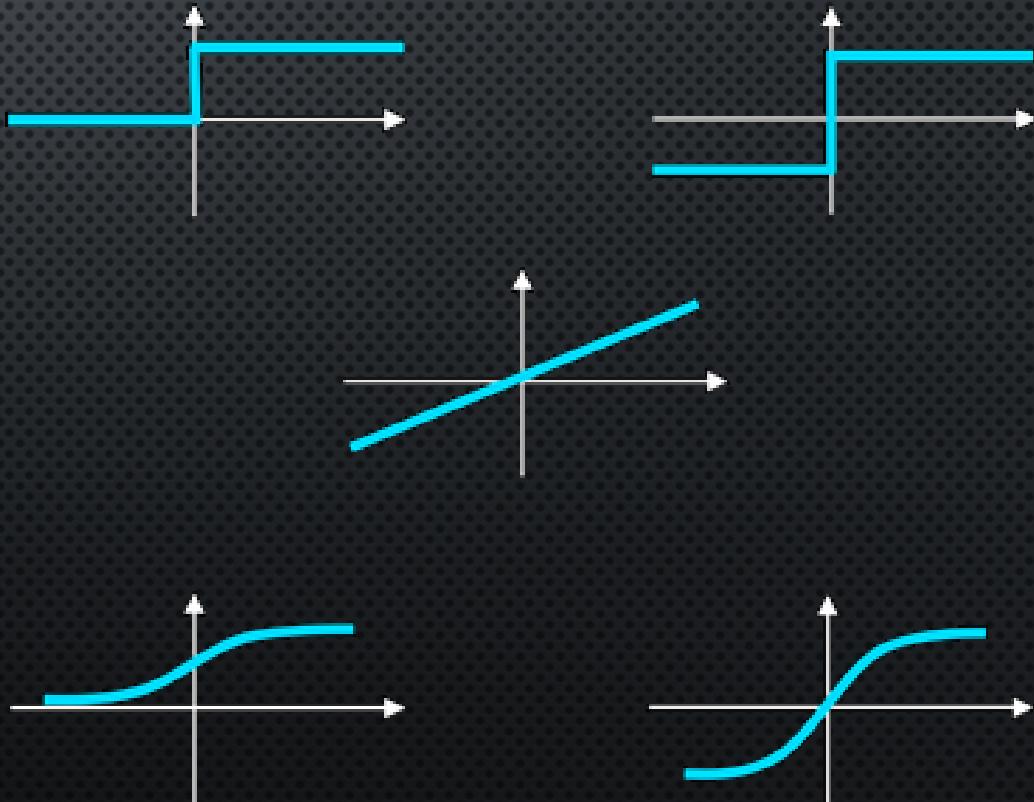


# Common activation functions

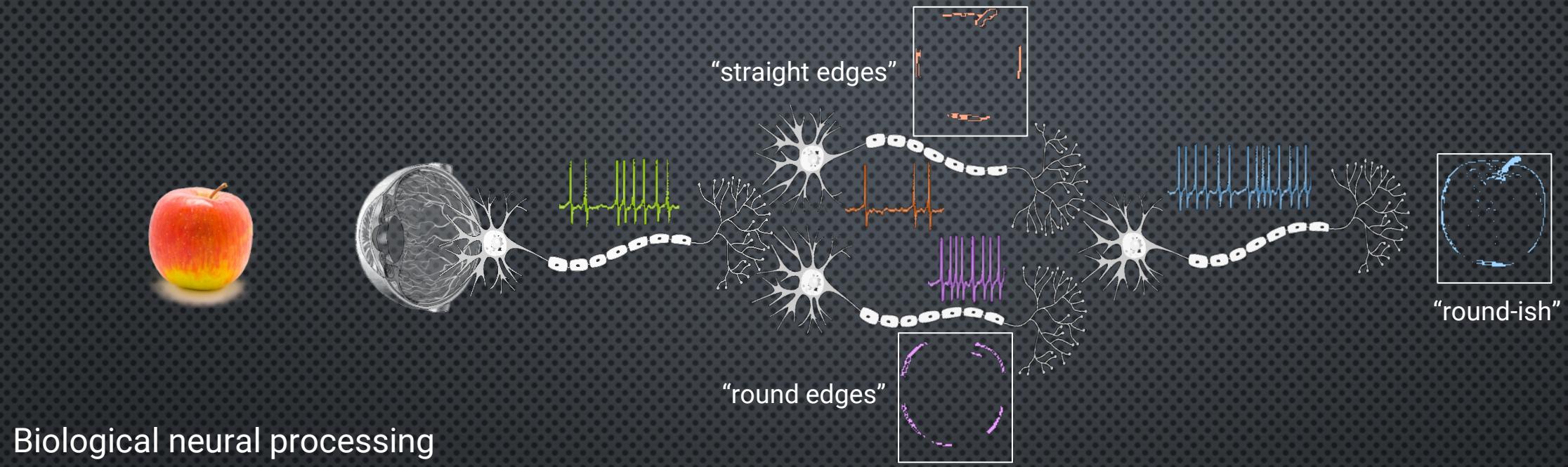
Activation function	Equation	Example	1D Graph
Unit step (Heaviside) Threshold Logic Unit (TLU)	$\phi(z) = \begin{cases} 0, & z < 0, \\ 0.5, & z = 0, \\ 1, & z > 0, \end{cases}$	Perceptron variant	
Sign (Signum)	$\phi(z) = \begin{cases} -1, & z < 0, \\ 0, & z = 0, \\ 1, & z > 0, \end{cases}$	Perceptron variant	
Linear	$\phi(z) = z$	Adaline, linear regression	
Piece-wise linear	$\phi(z) = \begin{cases} 1, & z \geq \frac{1}{2}, \\ z + \frac{1}{2}, & -\frac{1}{2} < z < \frac{1}{2}, \\ 0, & z \leq -\frac{1}{2}, \end{cases}$	Support vector machine	
Logistic (sigmoid)	$\phi(z) = \frac{1}{1 + e^{-z}}$	Logistic regression, Multi-layer NN	
Hyperbolic tangent	$\phi(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$	Multi-layer NN	

# Which activation function to choose?

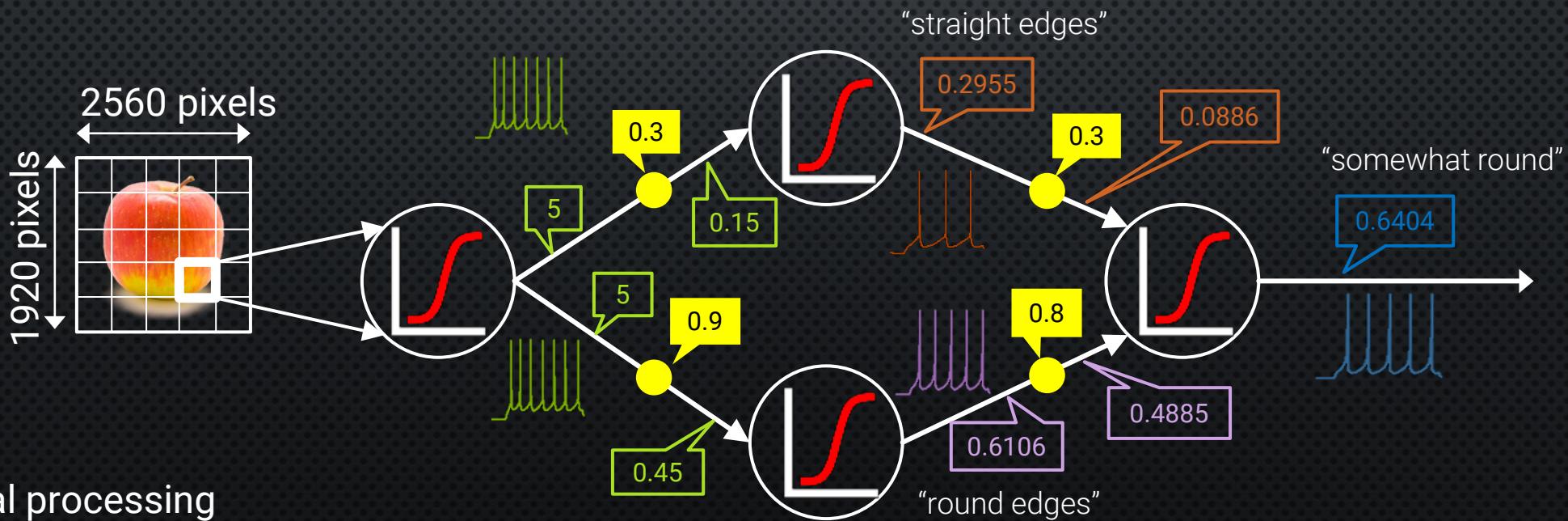
- Activation function decides what output the perceptron produces
- Choose an AF based on what you want your output to look like. For example, if your output:
  - can only be 0 or 1 (-1 or 1), choose TLU or sign
    - Good for binary classification problems but bad for everything else
  - can be any value between  $-\infty$  to  $+\infty$ , choose linear
    - Almost always a bad choice so try to avoid using it
  - can be any value between 0 or 1 (-1 or 1), choose logistic or hyperbolic tangent
    - Almost always the best choice for most ML problems



# Biologically-inspired computation

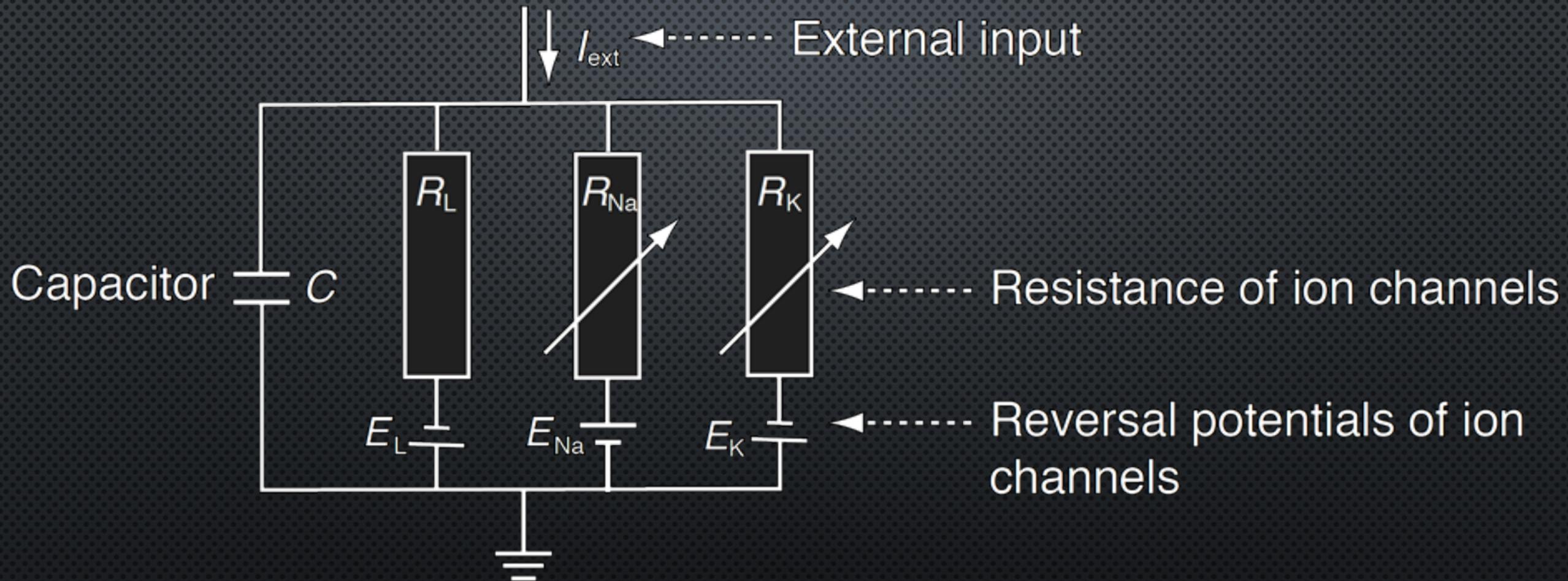


Biological neural processing



Artificial neural processing

# A biologically realistic neuron model: Hodgkin-Huxley model



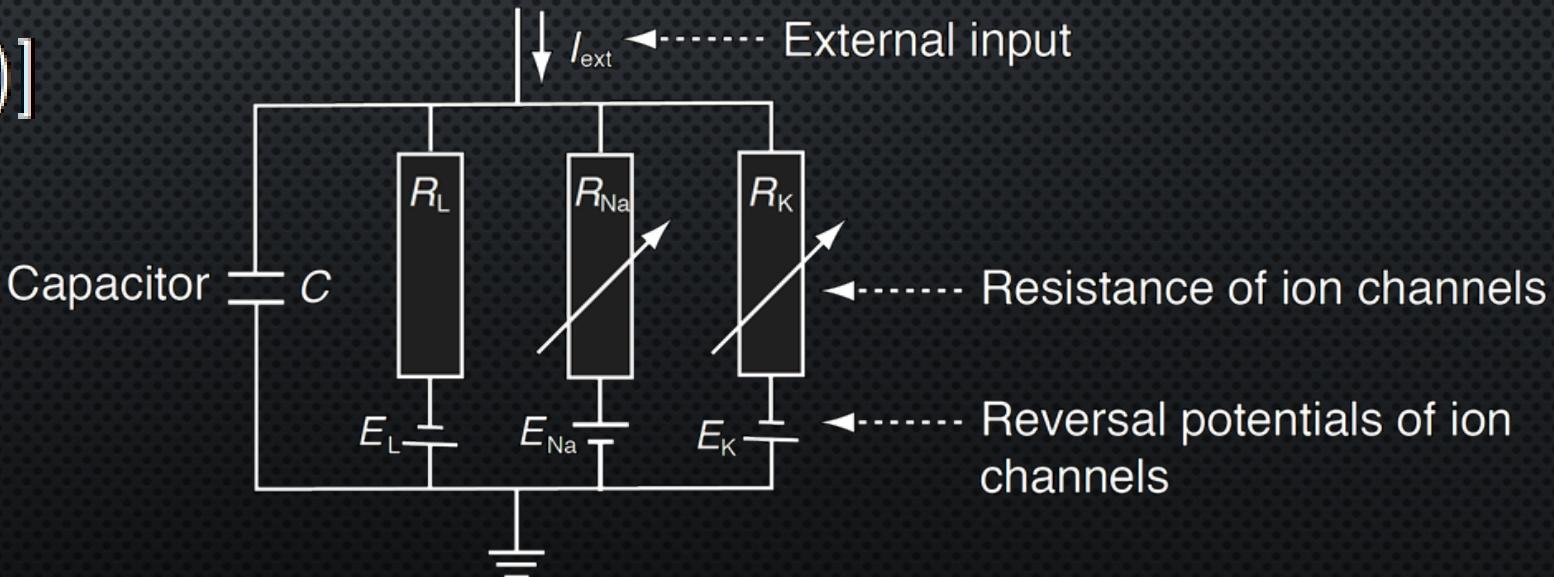
# A biologically realistic neuron model: Hodgkin-Huxley model

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

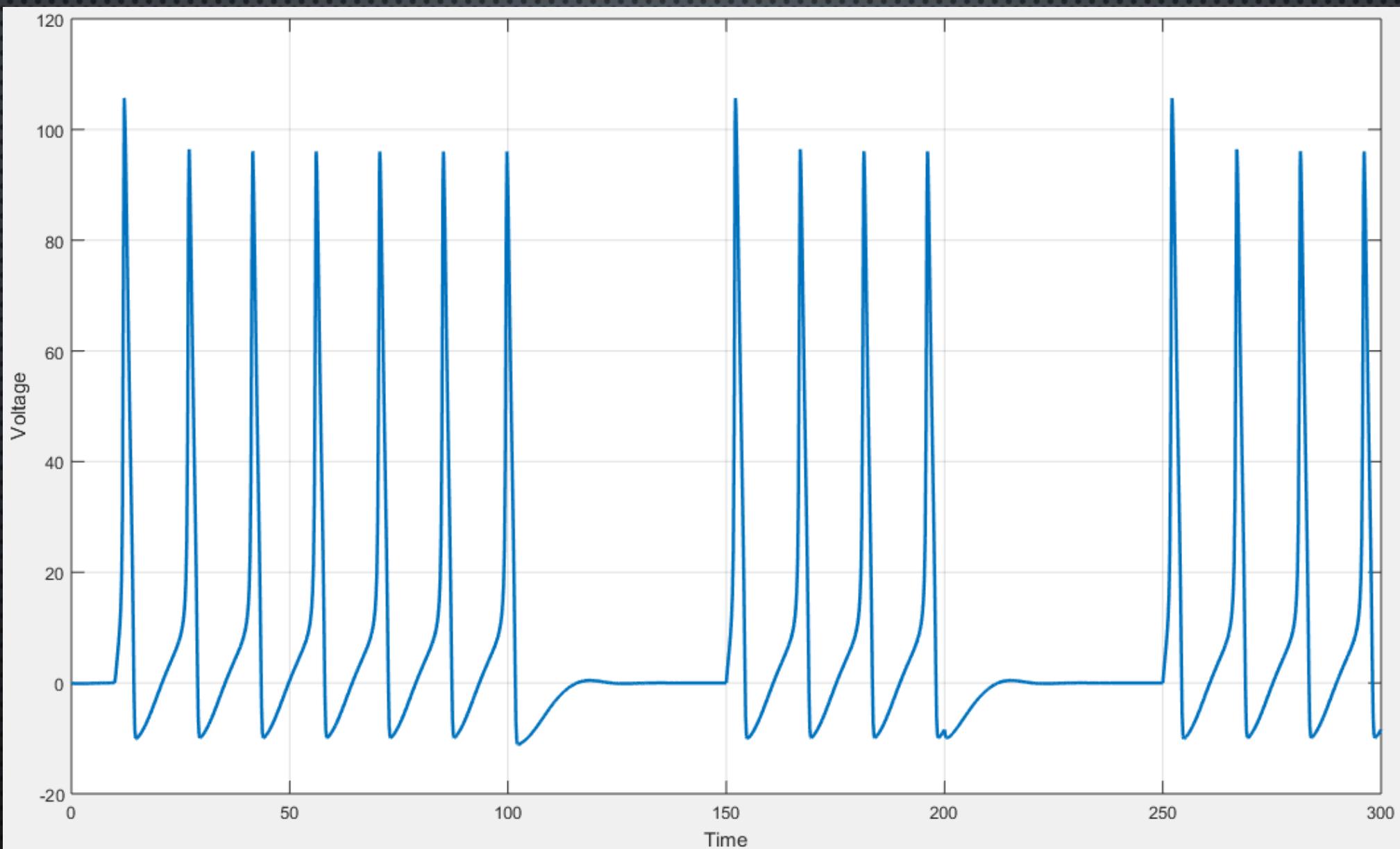
$$\tau_n(V) \frac{dn}{dt} = -[n - n_0(V)]$$

$$\tau_m(V) \frac{dm}{dt} = -[m - m_0(V)]$$

$$\tau_h(V) \frac{dh}{dt} = -[h - h_0(V)]$$



# A biologically realistic neuron model: Hodgkin-Huxley model



# Matlab exercises

- Download “Assignment 1.pdf” from Itslearning and follow the instructions in it.