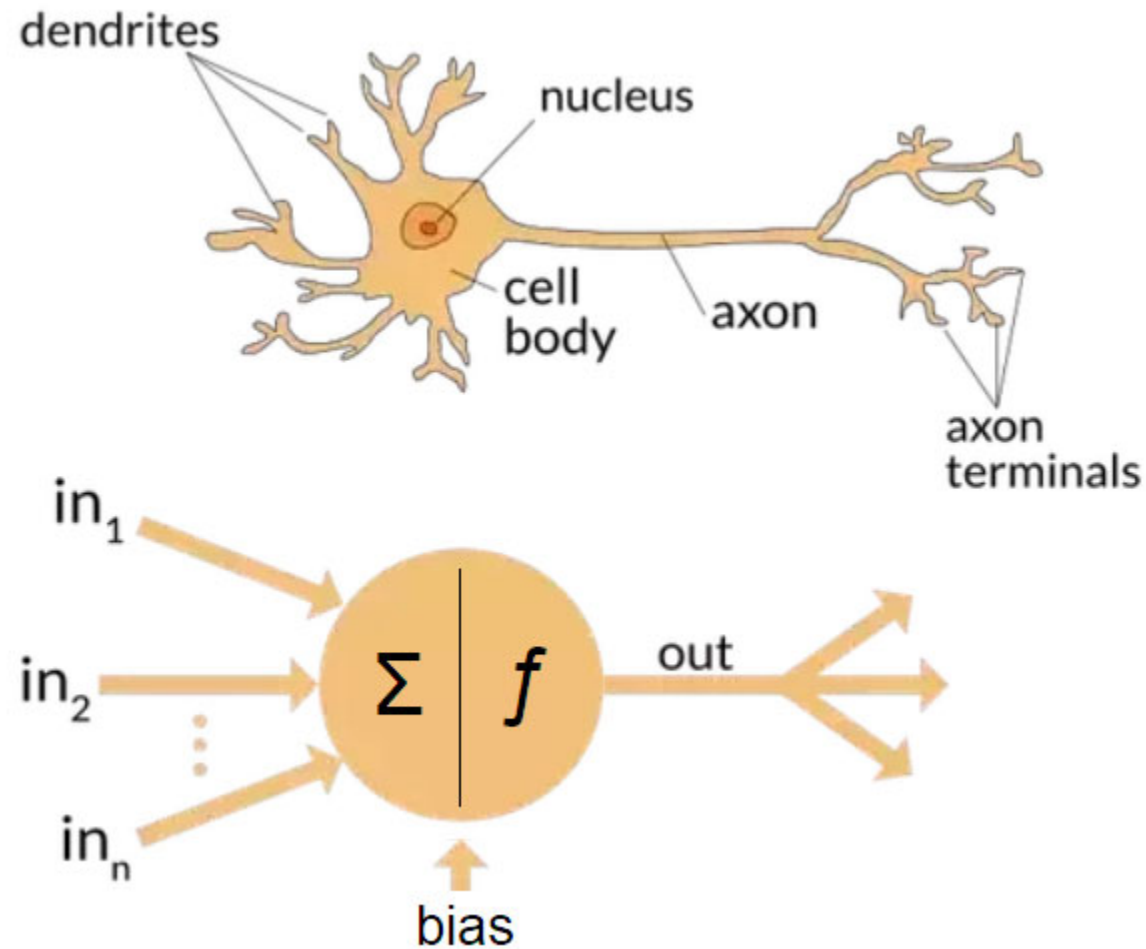


Machine Learning

Introduction to Neural Networks

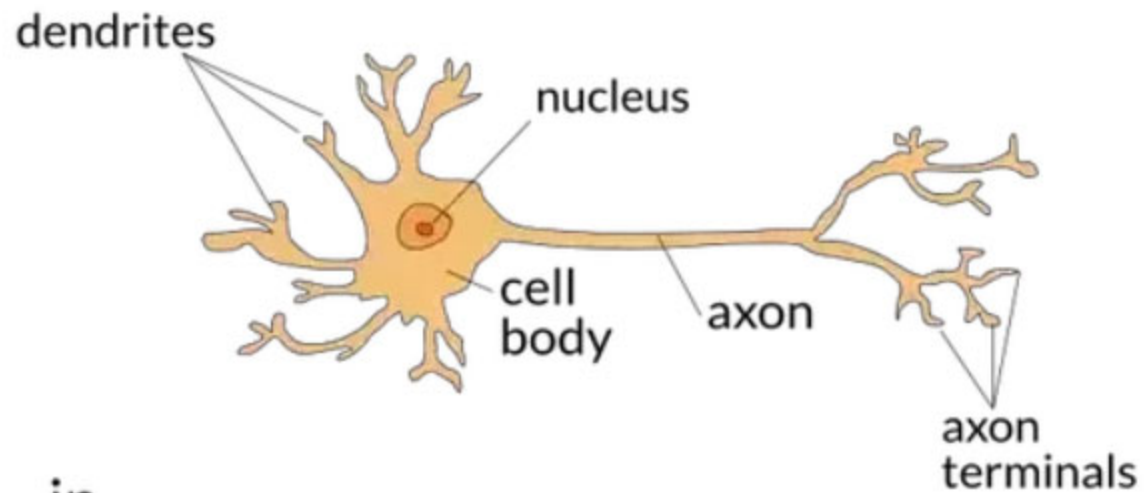
Jian Liu



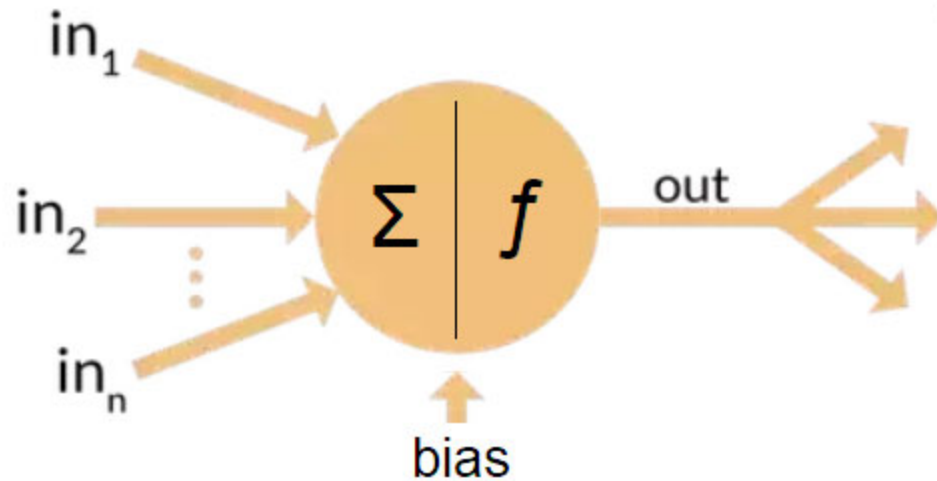
from towardsdatascience

Motivation for artificial neurons

Part 1: Biological neurons

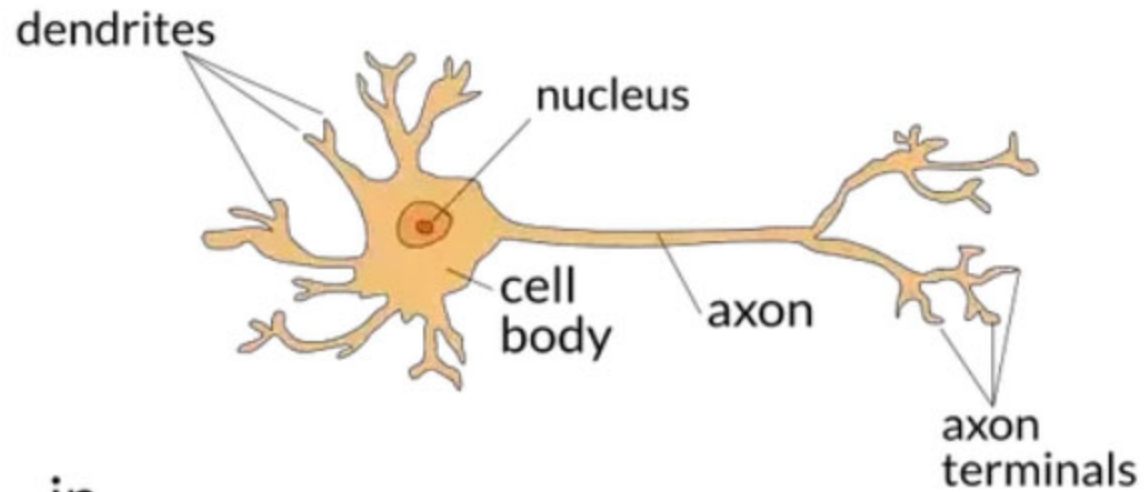


Part 2: Artificial neurons

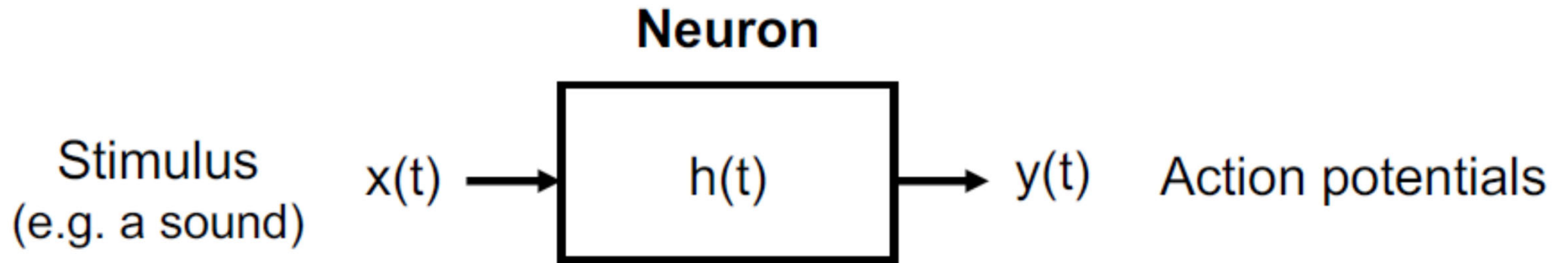


Motivation for artificial neurons

Part 1: Biological neurons



Neuron as an information processor



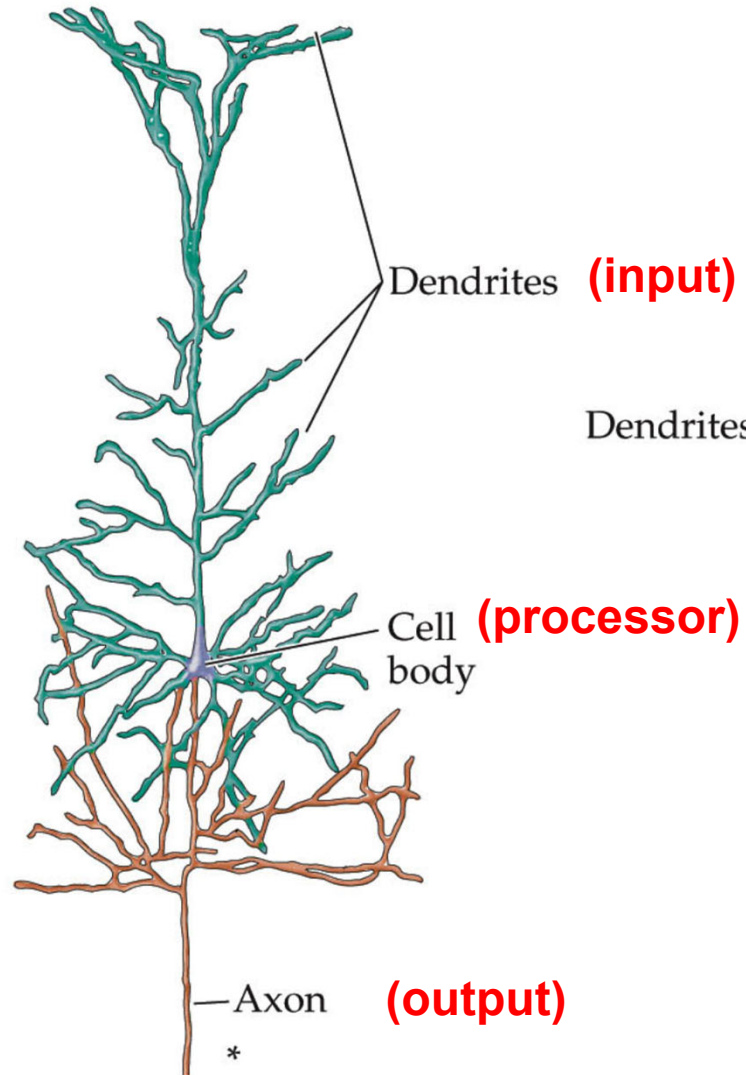
Neuron:

Structure: building block of the brain

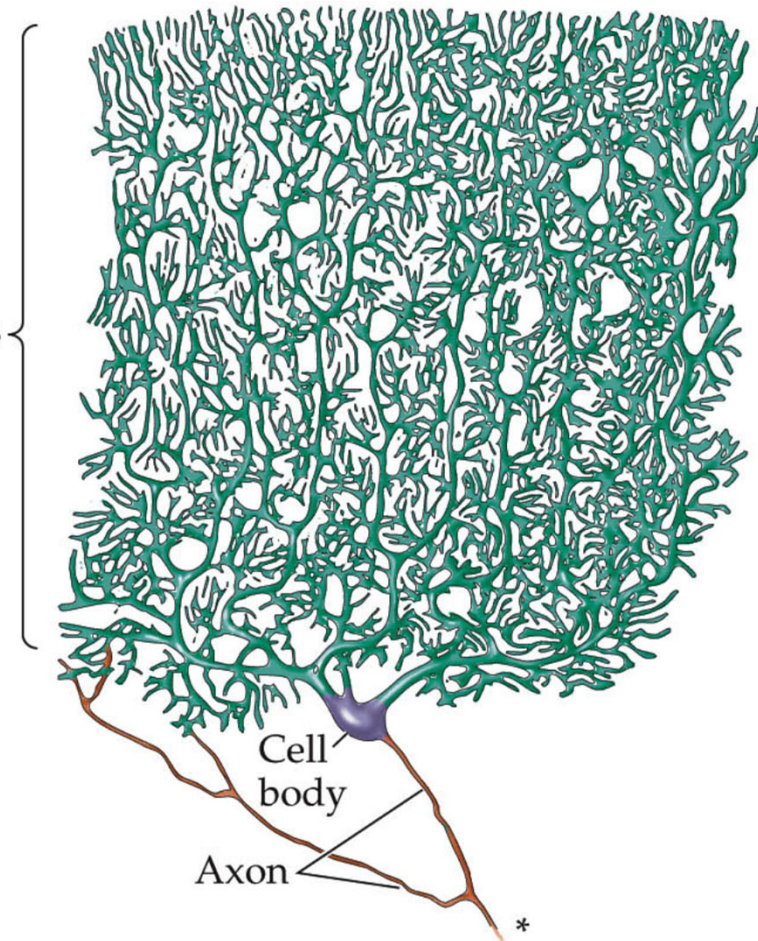
Function: basic computational unit

Neurons are building blocks of the brain

(E) Cortical pyramidal cell



(F) Cerebellar Purkinje cells

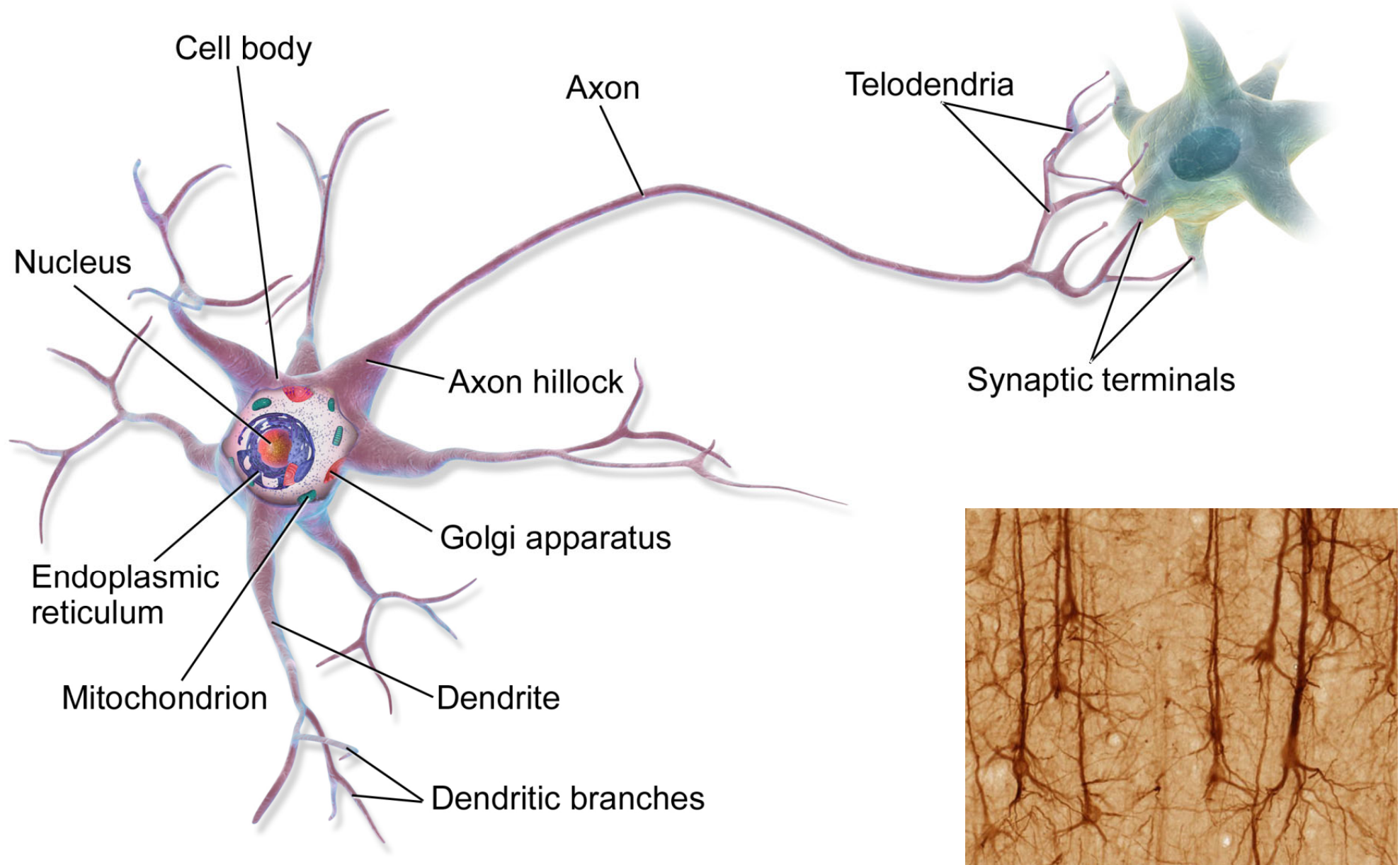


NEUROSCIENCE, Fourth Edition, Figure 1.2 (Part 2)

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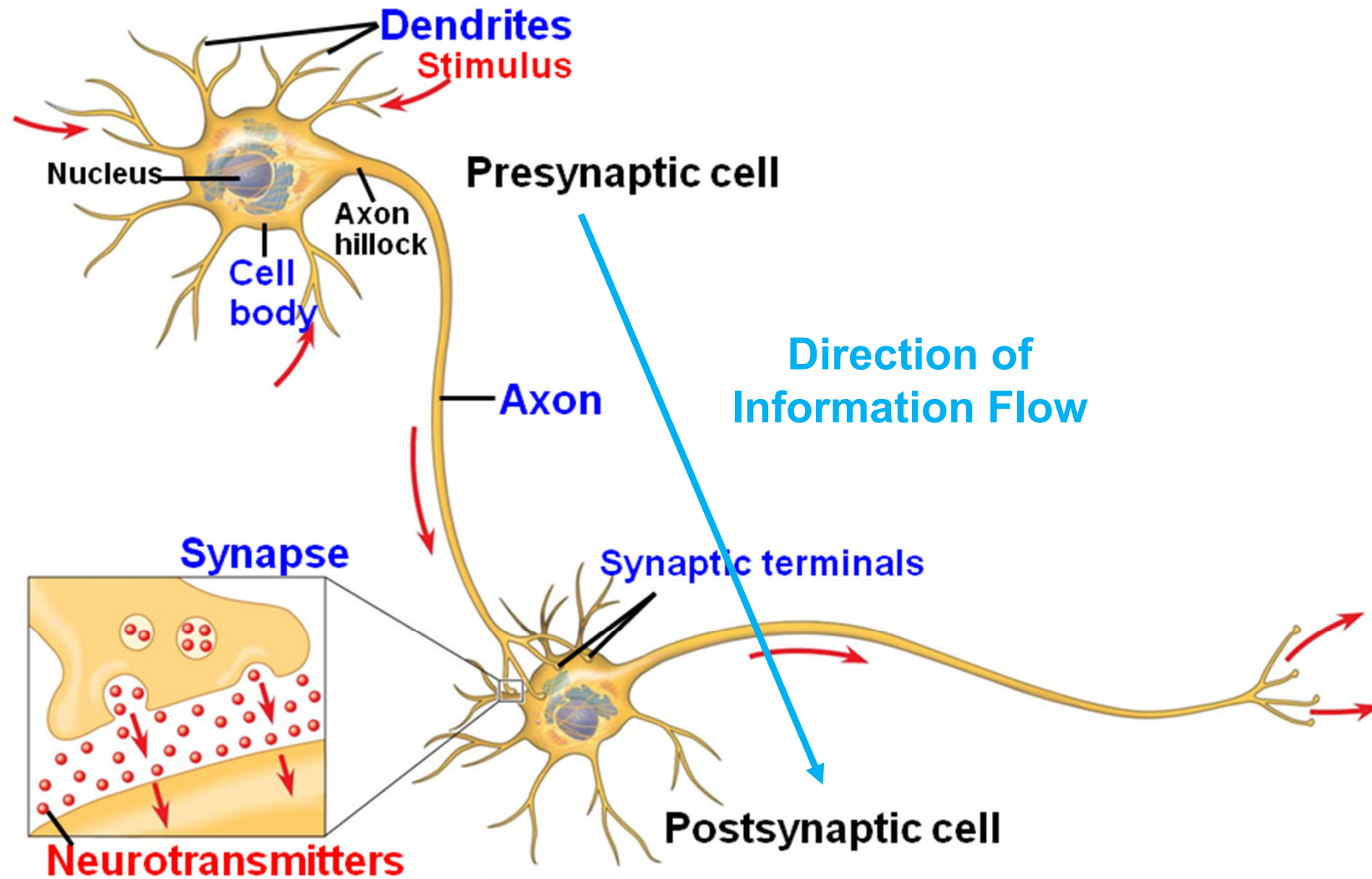
Neurons generally have similar “input-processor-output” structures

Structure of a typical neuron

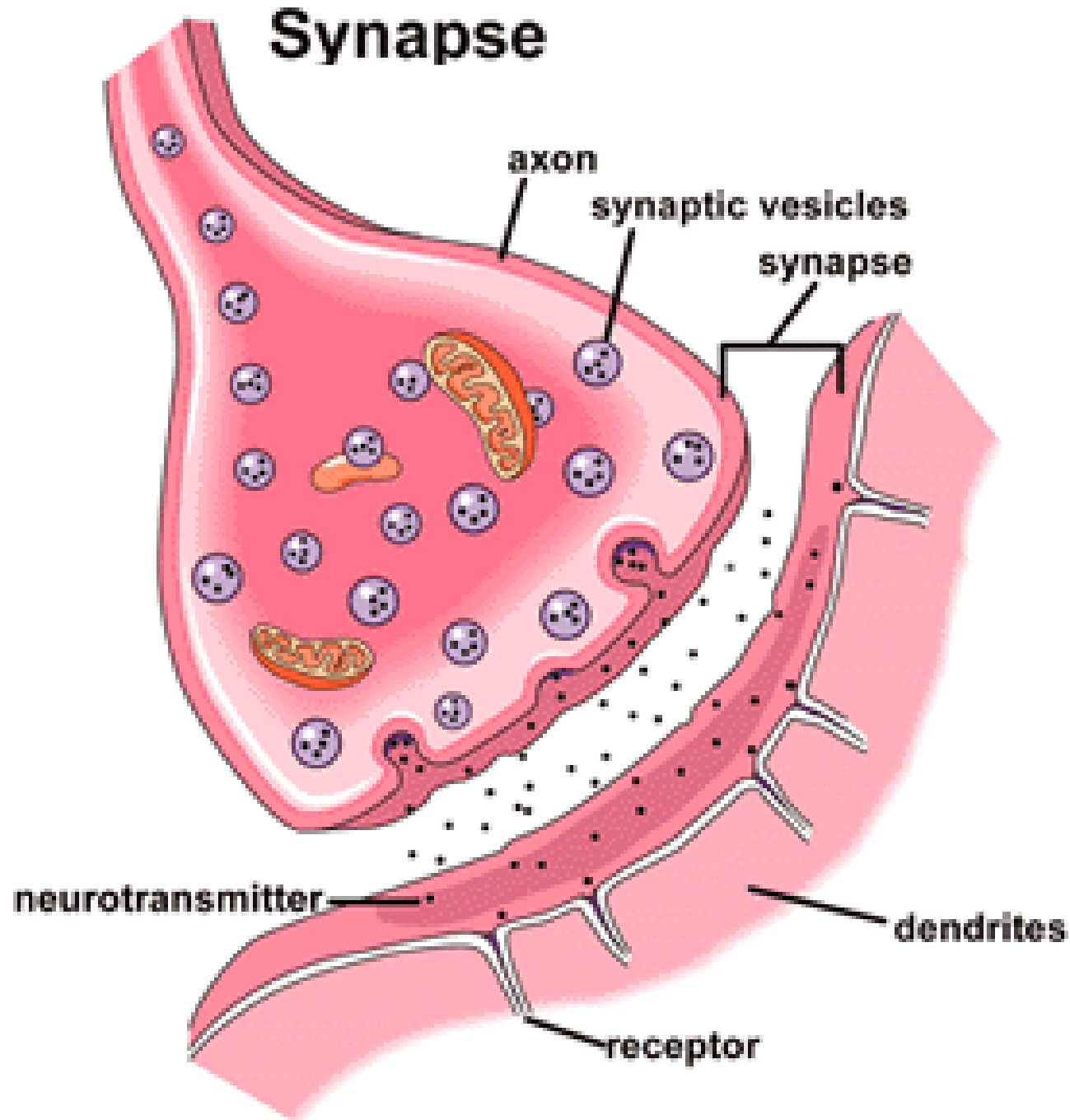


Neuronal information flow

Neurons

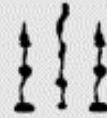


Synaptic weights

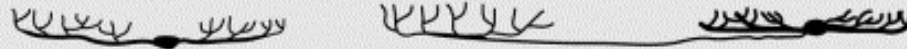


Retinal neurons

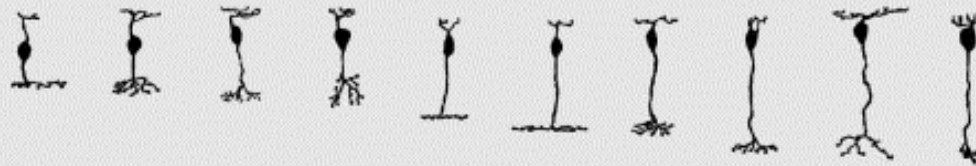
Photoreceptors



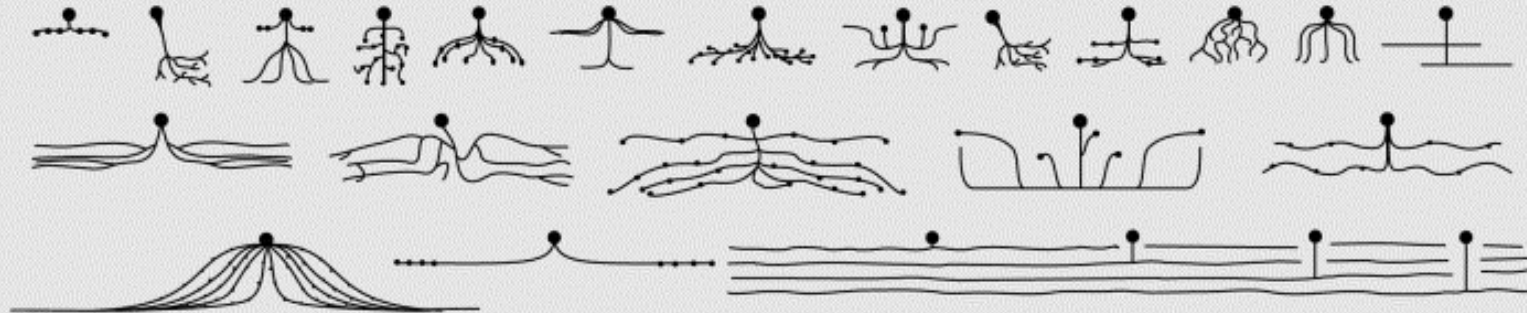
Horizontal cells



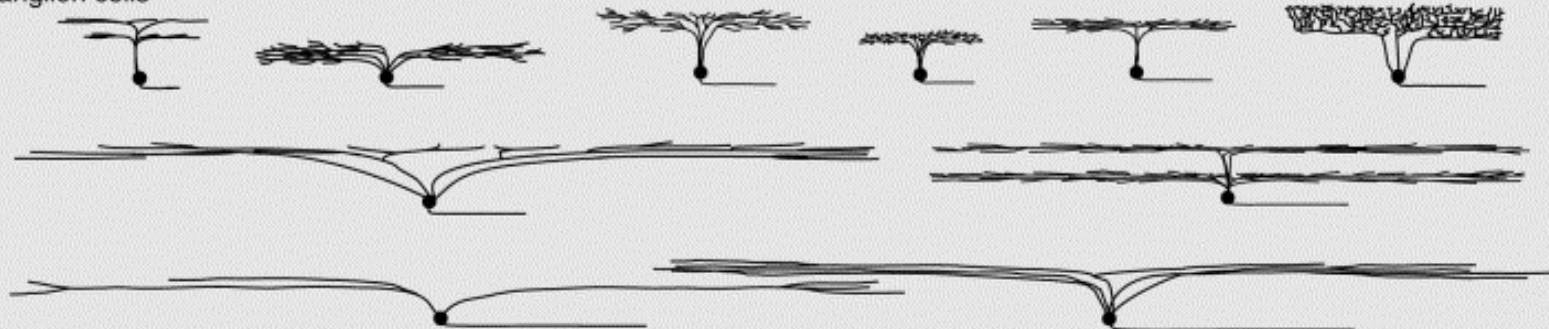
Bipolar cells



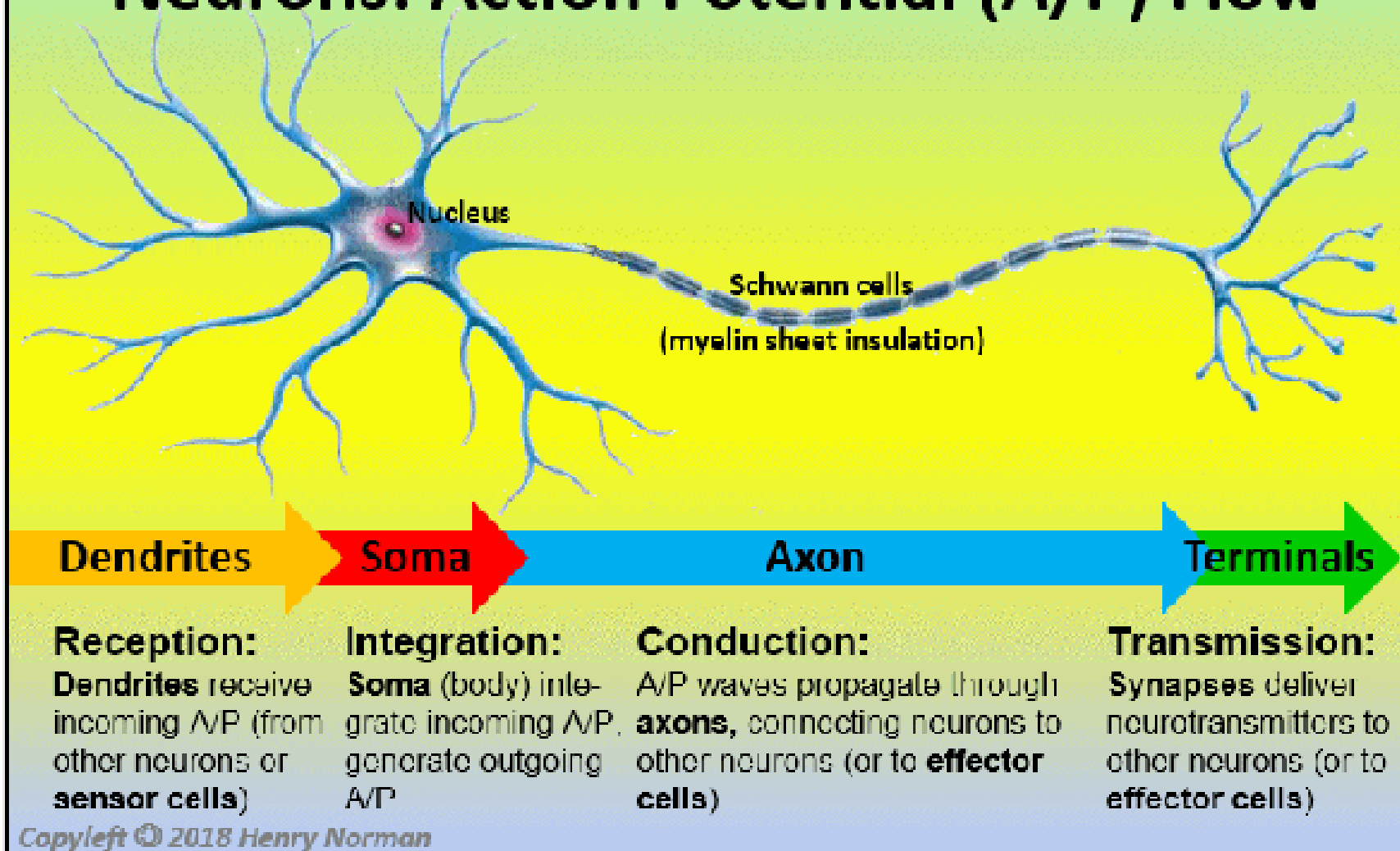
Amacrine cells



Ganglion cells

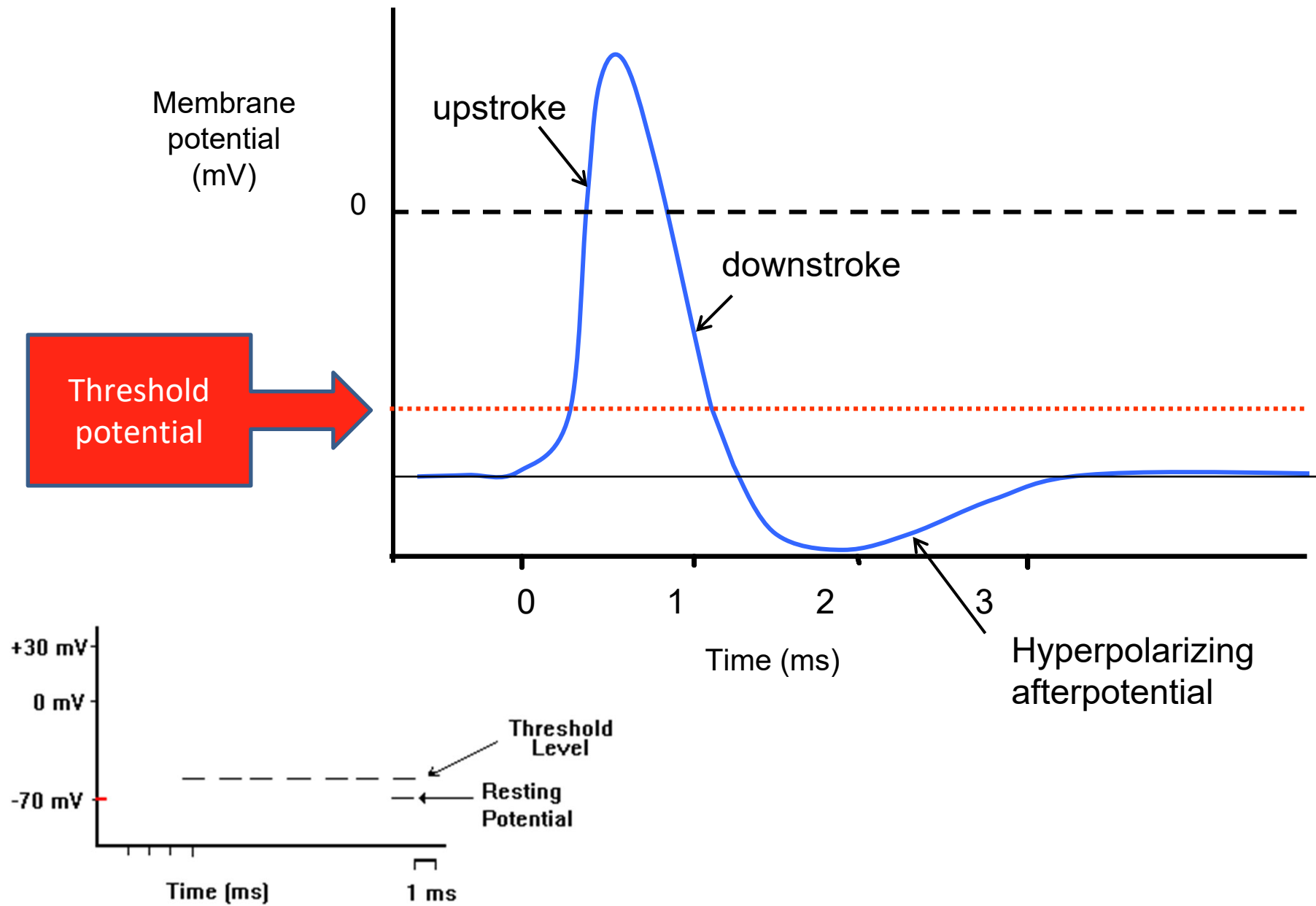


Neurons: Action Potential (A/P) Flow

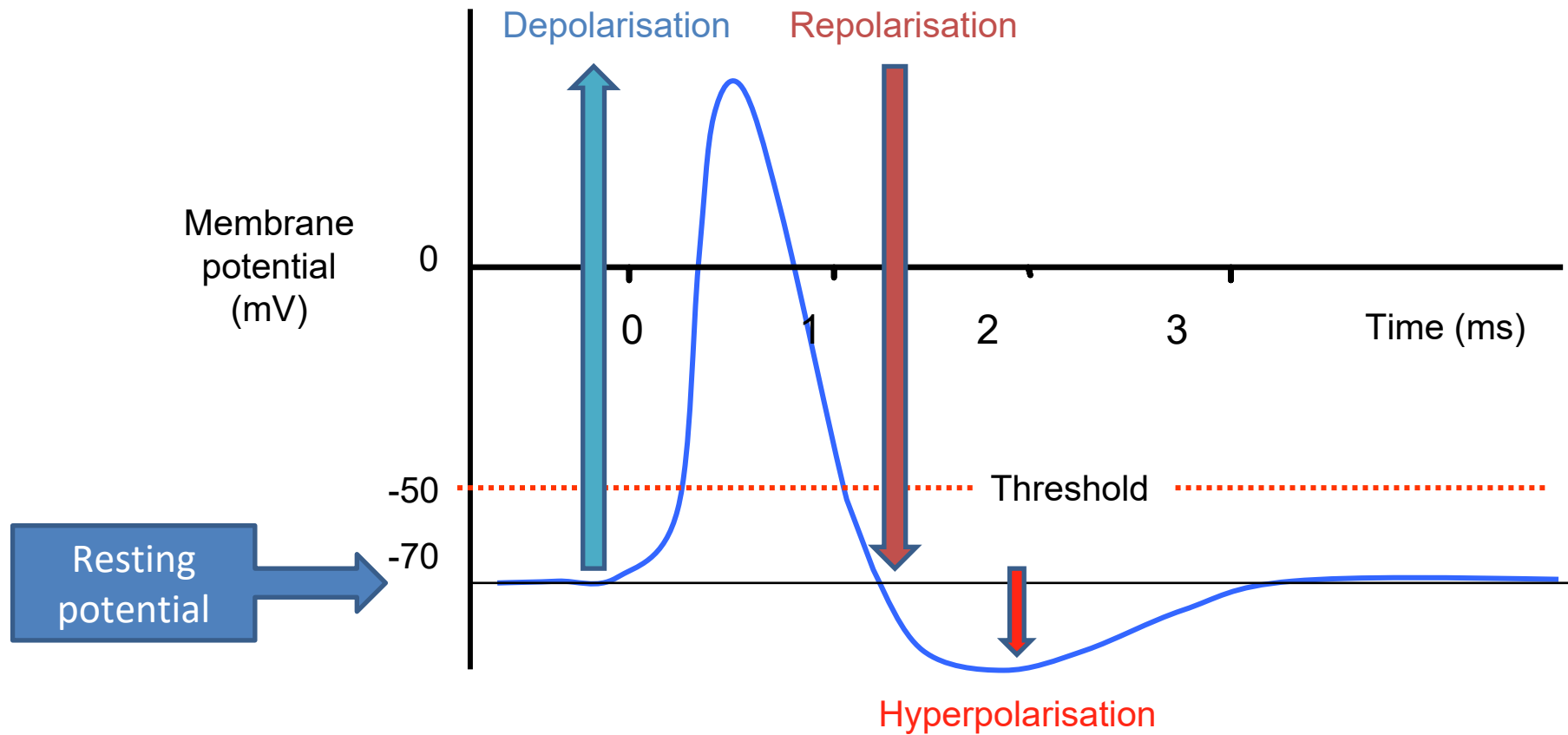


Structure → **Function**

The action potential

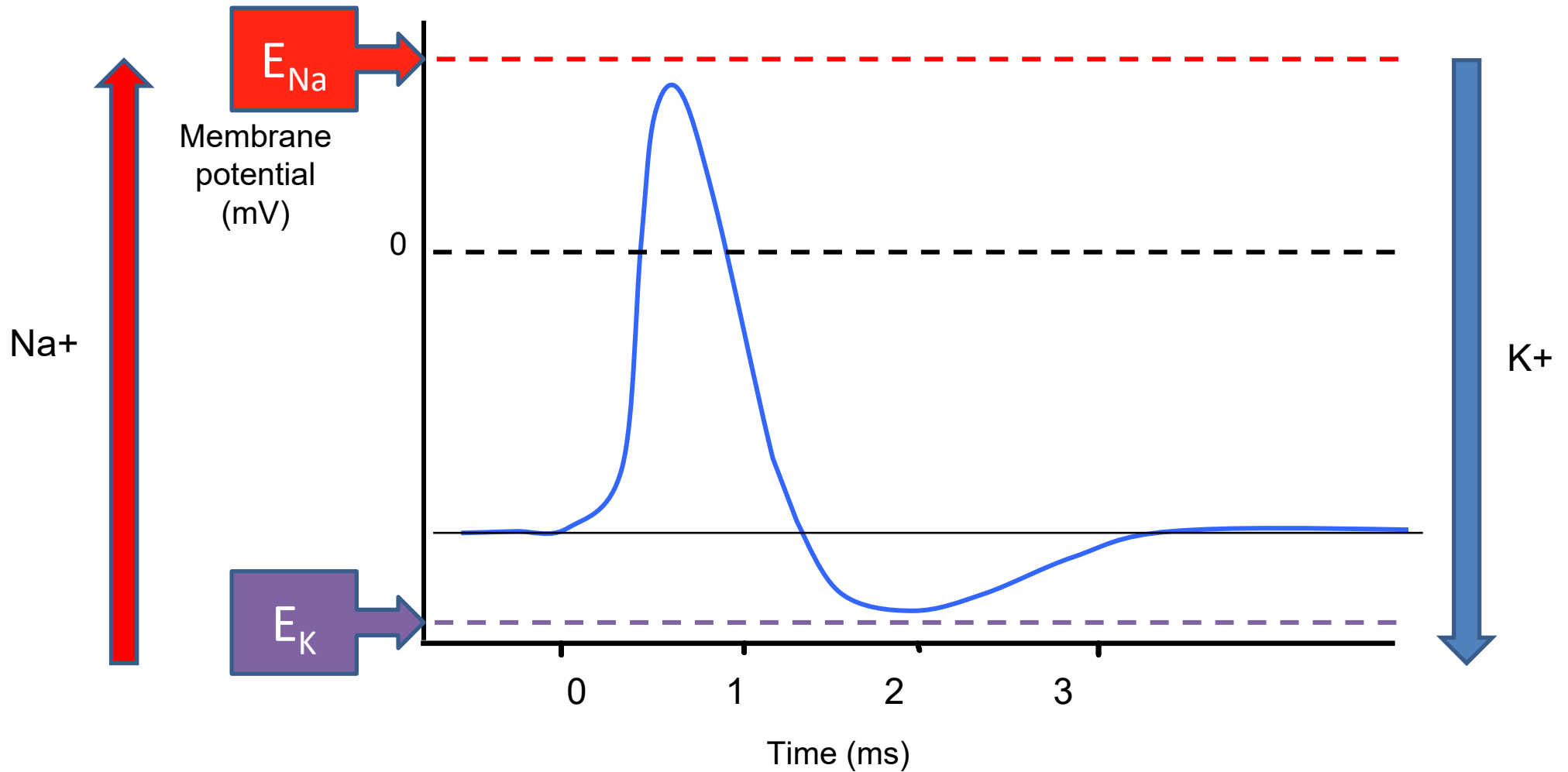


The action potential



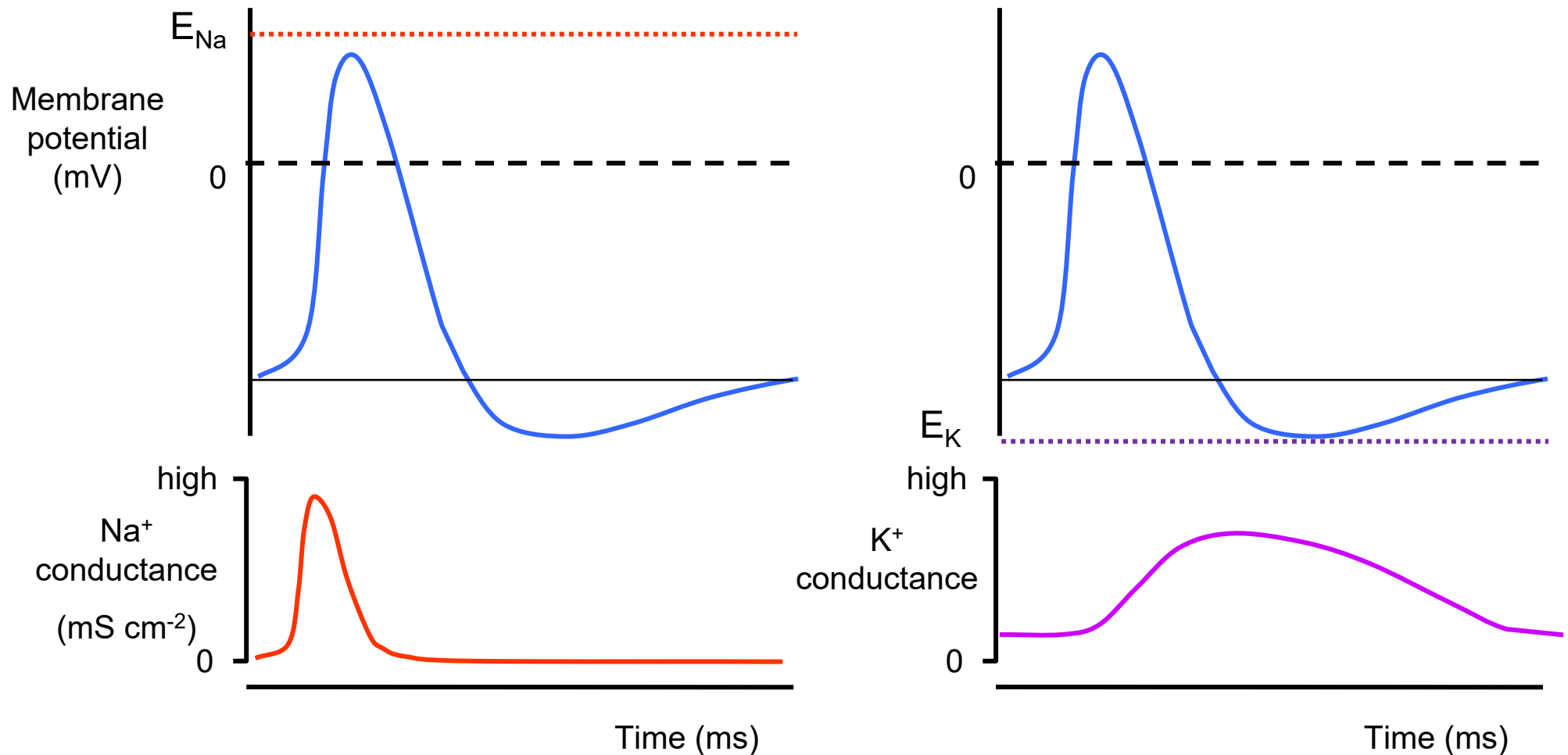
Synaptic inputs to neurons cause the membrane to depolarise and can lead to an action potential being triggered

The action potential



During action potential the selective permeability of membrane to ions changes due to activity of voltage-gated ion channels

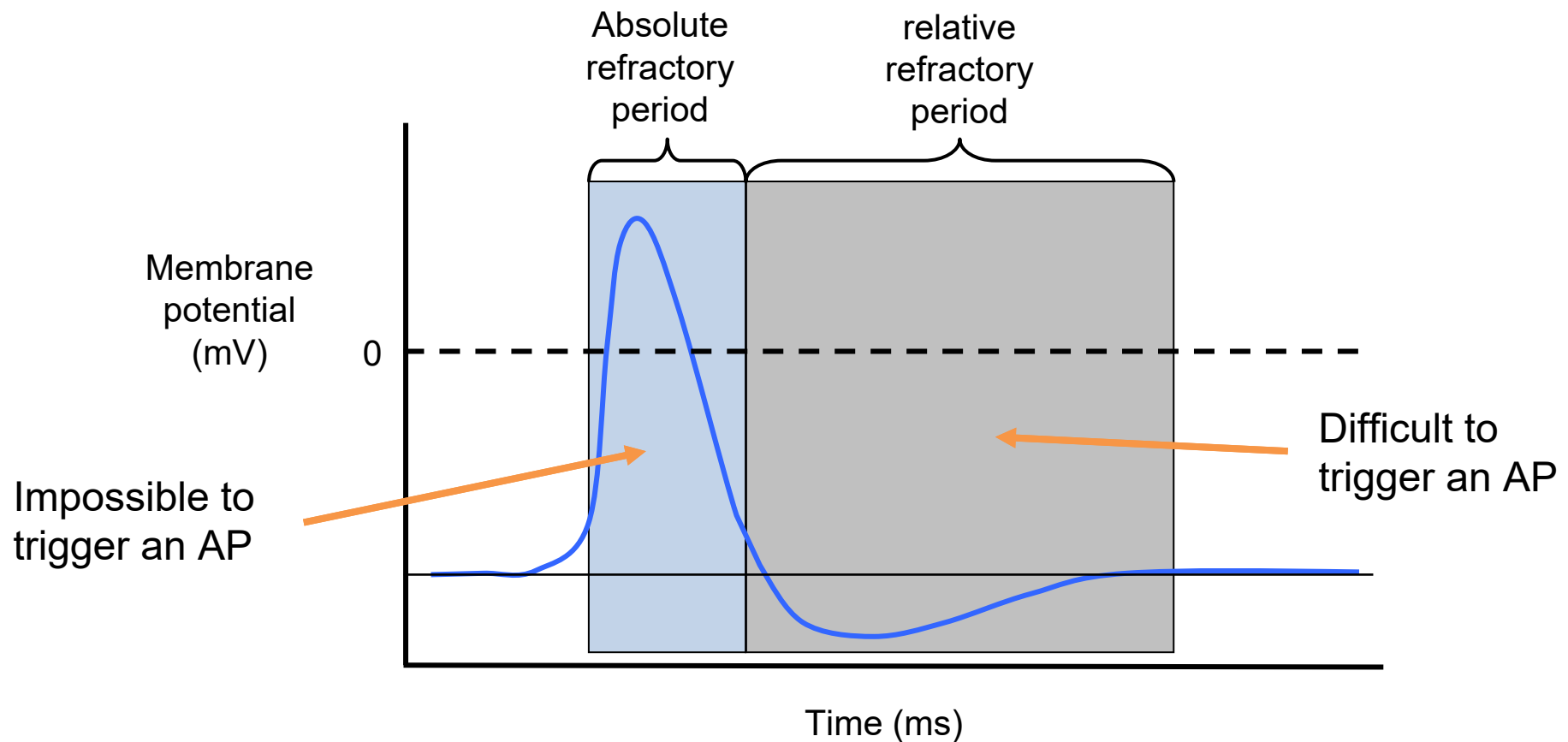
Changes of membrane permeability to sodium and potassium ions during the action potential



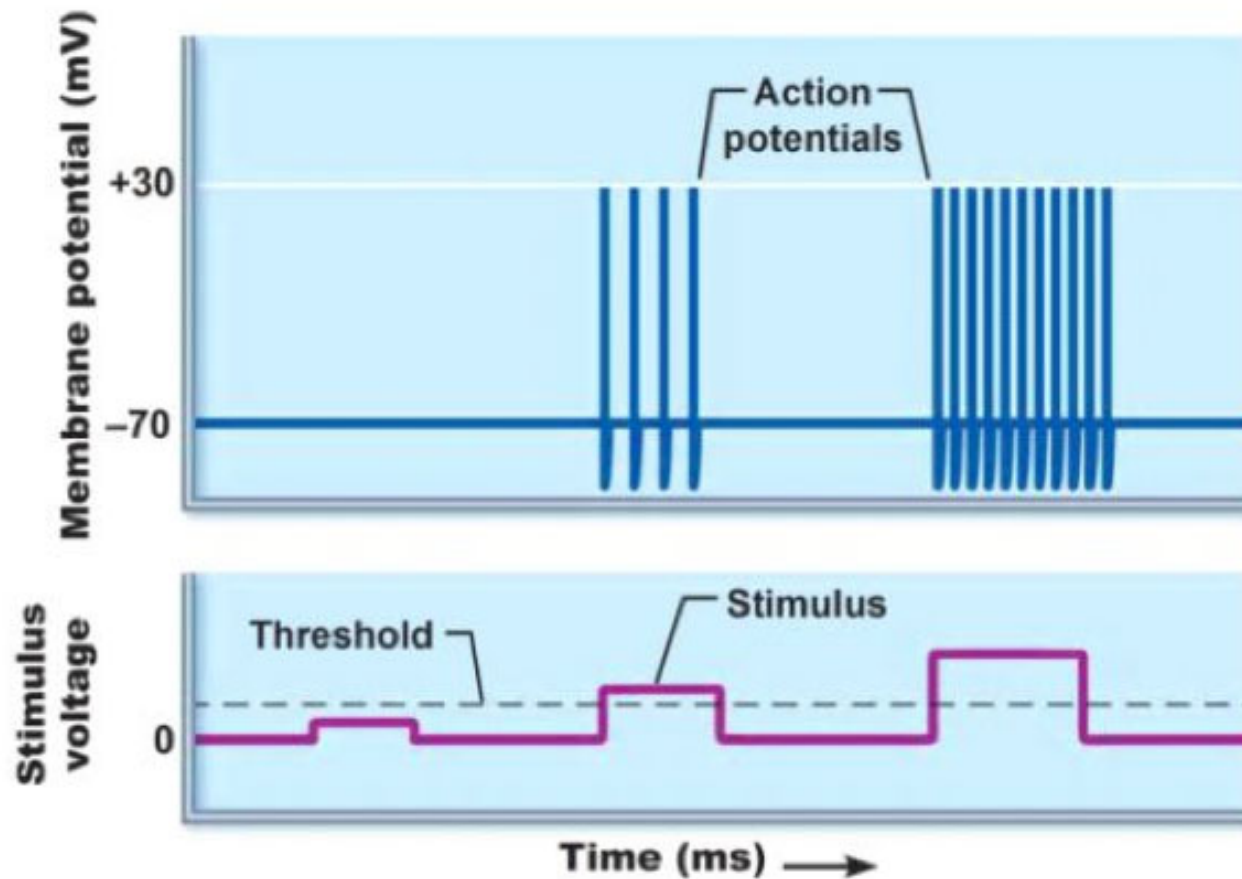
Voltage-gated ion channels permit the membrane to change its permeability

Properties of action potentials

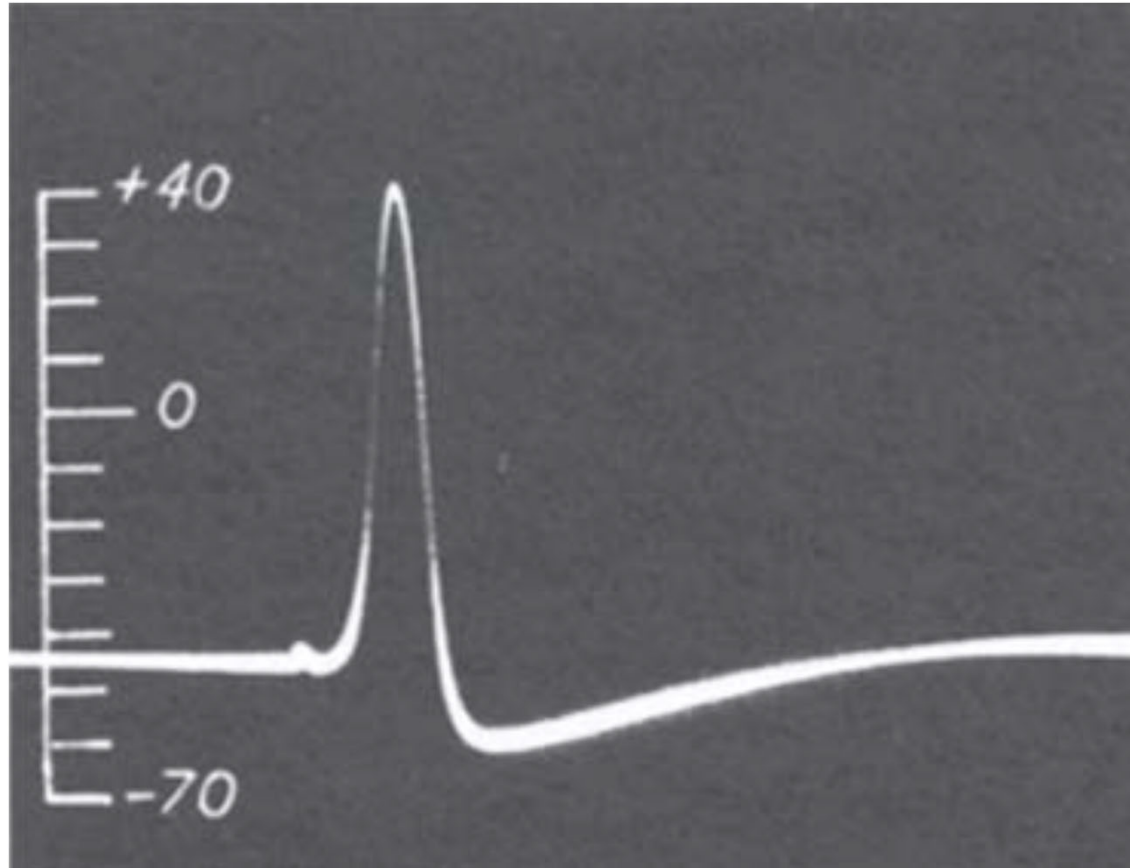
- 1) Threshold potential.
- 2) All or none – information coding is by frequency not amplitude.
- 3) Refractory periods – time during and after an AP when it is more difficult to fire a subsequent AP. Places an upper limit on AP firing frequency.



**Increasing stimulus intensity increases frequency,
not size of action potentials**



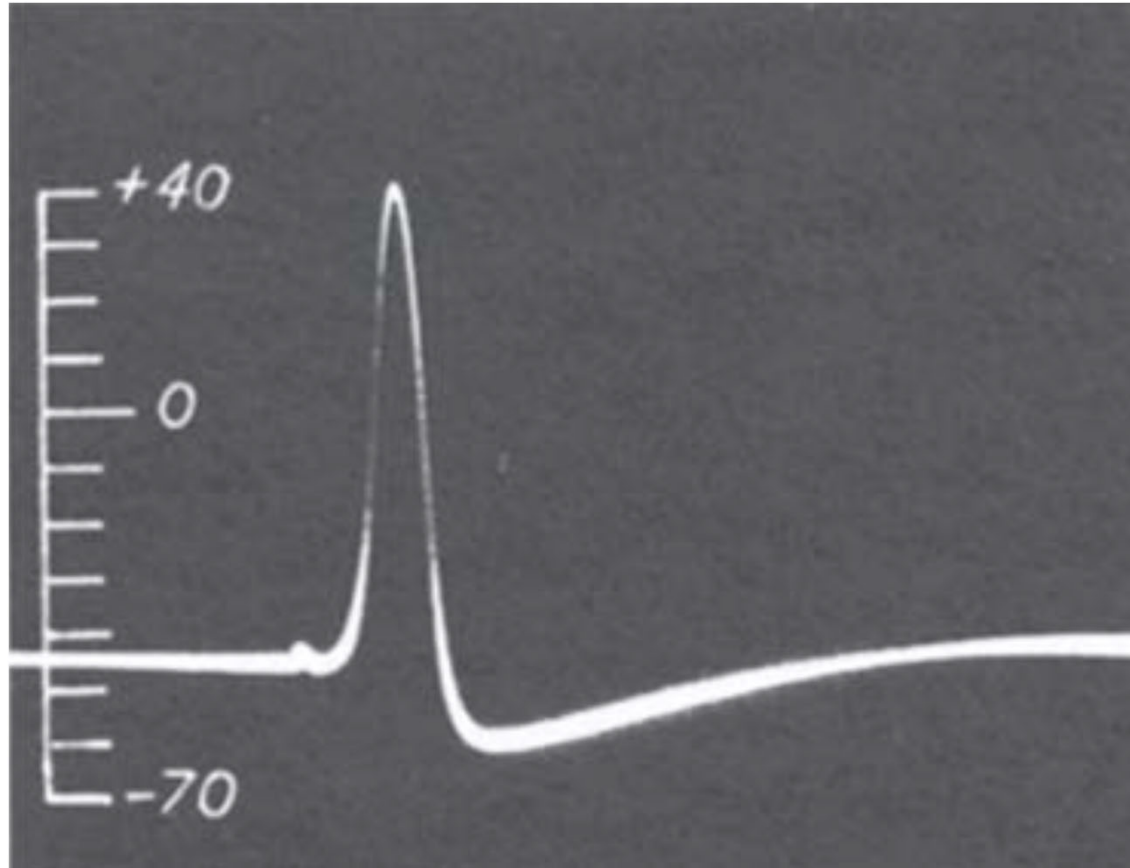
Hodgkin-Huxley model



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

Hodgkin-Huxley model

action potential generation



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

The action potential

The Action Potential

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Differential equations

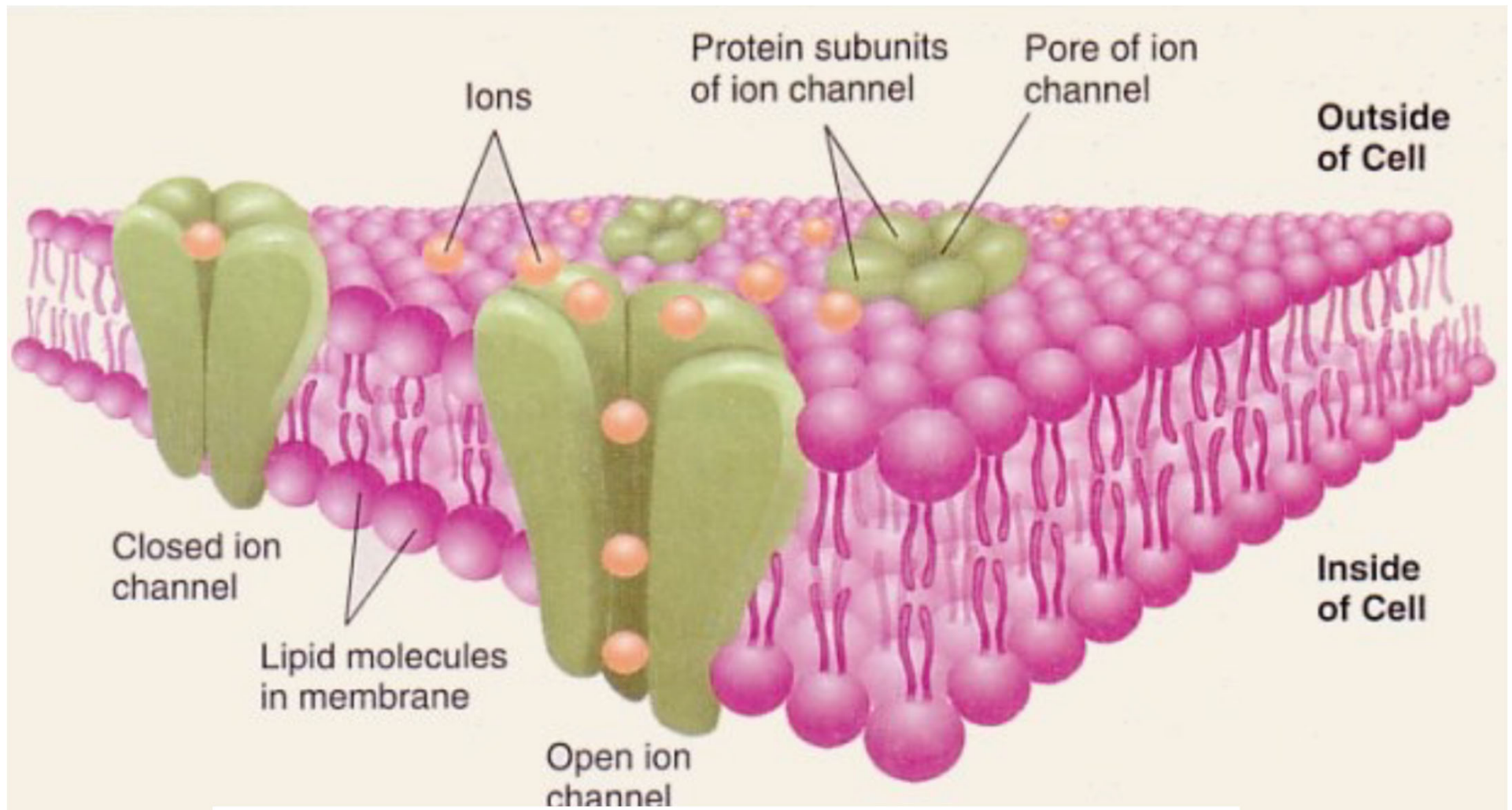
The most common and simplest types of dynamic models come in the form of differential equations

They describe the rate of change of some variable, say u , as a function u and other variables

$\frac{du}{dt}$ = change per unit time = sum of production rates – sum of removal rates.

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

Ion channels

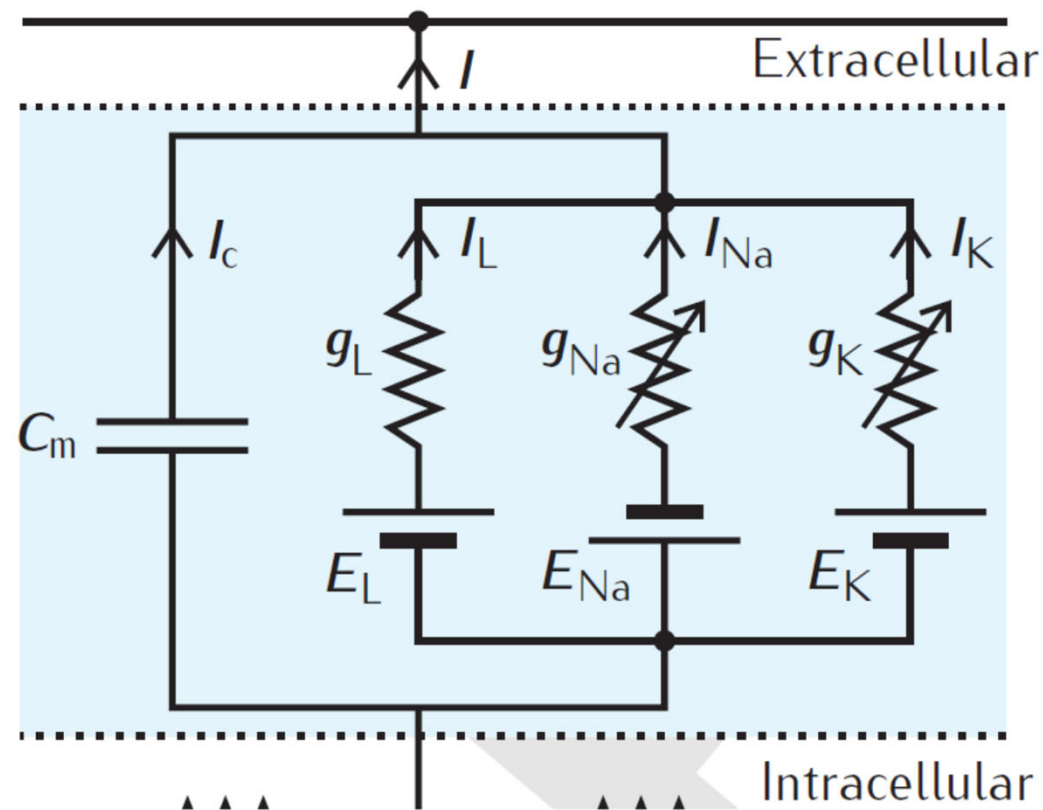


$$CV = Q$$

$$I = -g(V - E_r)$$

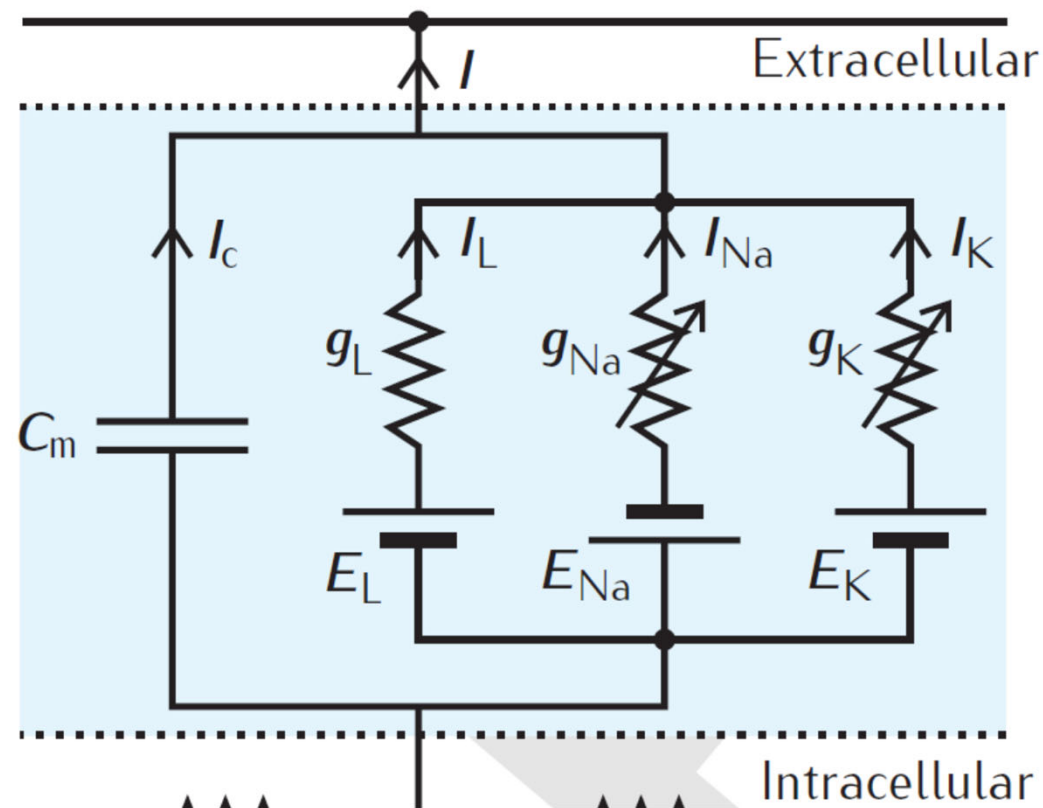
$$C \frac{dV}{dt} = \frac{dQ}{dt} = I$$

The Equivalent Electronic Circuit of a Neuron



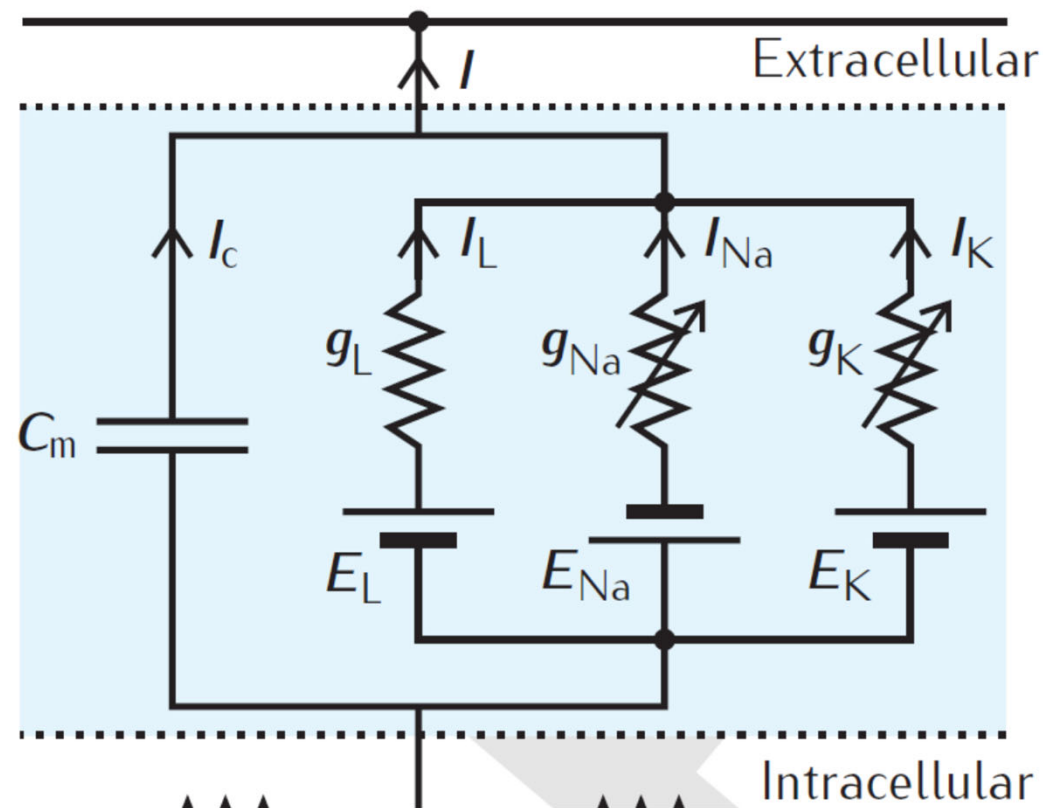
$$C_m \frac{dV}{dt} = - \sum_i g_i(V)(V - E_i) - \bar{g}_L(V - E_L) + I_e$$

The Equivalent Electronic Circuit of a Neuron



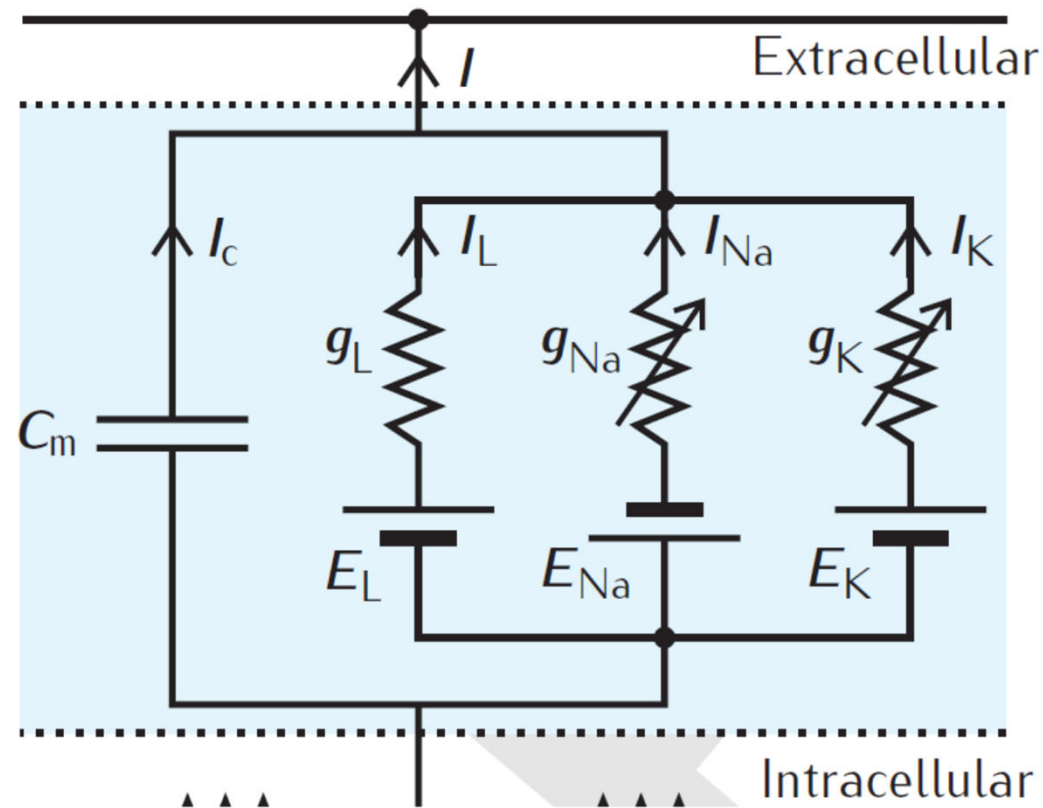
$$I_m(t) + C \frac{dV(t)}{dt} = I_e(t)$$

The Equivalent Electronic Circuit of a Neuron



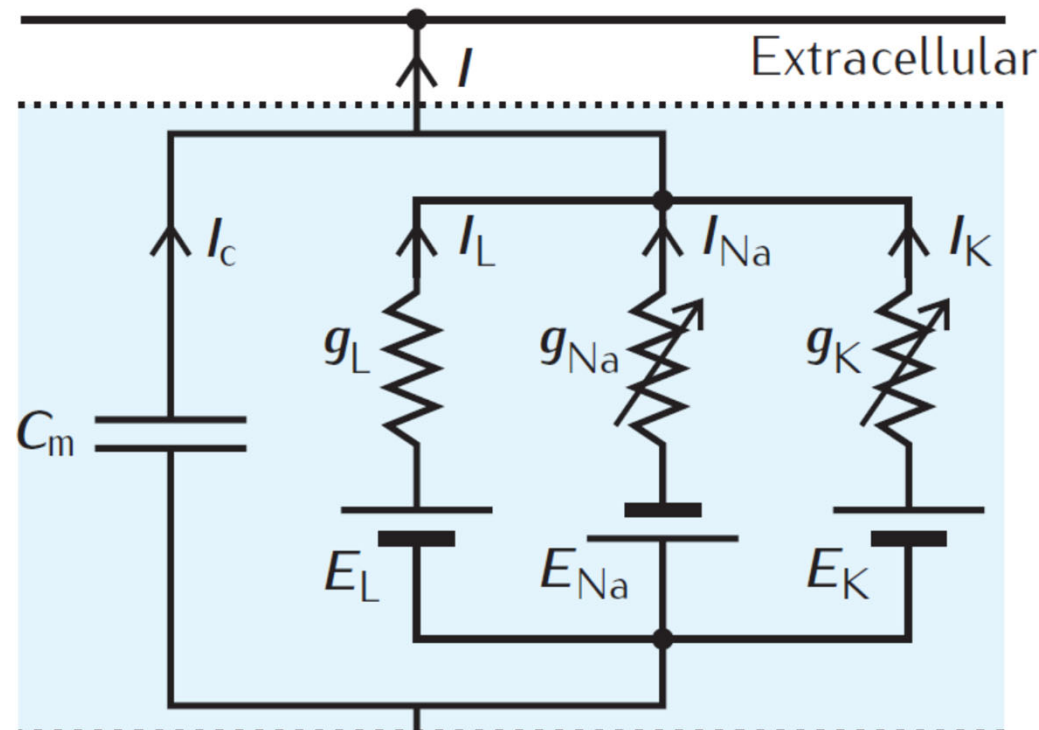
$$I_m = I_{Na} + I_K + I_L$$

The Equivalent Electronic Circuit of a Neuron



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

The Equivalent Electronic Circuit of a Neuron



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

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

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

$$E_K = -75mV$$

$$E_L = -50mV$$

Hodgkin-Huxley (heroic)

Empirical model that describes the ionic conductance and **generation of action potential**

Published in 1952

Nobel Prize in Medicine or Physiology in 1963

Work reflects a combination of experimental work, theoretical hypothesis, computational data fitting, and model prediction

HH model in action

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

Start with initial condition $V_m = V_0$ at time step t_0

Compute:

$$n_\infty(V) \text{ and } \tau_n(V) \quad m_\infty(V) \text{ and } \tau_m(V) \quad h_\infty(V) \text{ and } \tau_h(V)$$

$$n(t) = n(t-1) + \frac{dn}{dt} \Delta t \quad m(t) = m(t-1) + \frac{dm}{dt} \Delta t \quad h(t) = h(t-1) + \frac{dh}{dt} \Delta t$$

$$I_K = \bar{G}_K n^4 (V - E_K) \quad I_{Na} = \bar{G}_{Na} m^3 h (V - E_{Na}) \quad I_L = \bar{G}_L (V - E_L)$$

$$\text{Total membrane current} \quad I_m = I_K + I_{Na} + I_L$$

Compute τ_{mem} and V_∞

$$V_m(t) = V_m(t-1) + \frac{dV_m}{dt} \Delta t$$