

# *Modeling neural flow through linearization procedures*

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1. Neuron models: from real world to mathematical and technical models.
2. Membrane phenomena in neural cells.
3. Cable equation of the axon.
4. Neural networks. Hopfield type models.

# Models for neurons

From real world to mathematical and technical models

## Why models?

- Modeling – the natural mental action in understanding the world
- The model should be realistic
- It has to capture the biological complexity but to allow further investigations.

## Models may describe an object or a phenomenon as:

- Structure (mechanical model)
- Function (electronic or mathematical or computer model)
- Evolution (transport through membrane or along axon, synaptic transmission)

## The main neuron models are expressed in terms of

- Mathematical equations (Hodgkin-Huxley equations)
- An imaginary construction following the laws of physics (Eccles model)

# Membrane phenomena in neural cells

## Electric conduction through membrane

The energy spent as work of electrical force to move an electric charge along an electric circuit is given by:

$$W_e = Q(\Phi_0 - \Phi_i) = zF(\Phi_0 - \Phi_i)$$

$$\begin{aligned} \Phi_0 ; \Phi_i &= \text{extra and intracellular potentials} \\ z &= \text{valence of the ions} \end{aligned} \quad (1)$$

$$F = \text{Faraday's constant } [9.649 \times 10^4 \text{ C/mol}]$$

## Transport phenomena through membrane

According to Ohm's law, current density and electric field are related by

$$\bar{J} = \sigma \bar{E} = -\sigma \nabla \Phi \quad \text{where } \sigma \text{ is the conductivity of the medium} \quad (2)$$

For a system with many types of ions:

$$\sigma = \sum u_k \cdot \frac{z_k}{|z_k|} c_k$$

$$\begin{aligned} \text{where: } u_k &= \text{ionic mobility } [\text{cm}^2/(\text{V}\cdot\text{s})] \\ z_k &= \text{valence of the ion} \end{aligned} \quad (3)$$

According to Fick's Law:

$$\bar{j}_{kD} = -D_k \nabla c_k$$

$$D_k = \frac{u_k RT}{|z_k| F}$$

where:

$T$  = absolute temperature [K]

$R$  = gas constant [8.314 J/(mol·K)]

$J_{kD}$  = ionic flux (due to diffusion) [mol/(cm<sup>2</sup>·s)]

$D_k$  = Fick's constant (diffusion constant) [cm<sup>2</sup>/s]

$c_k$  = ion concentration [mol/cm<sup>3</sup>]

$c_{i,k}$  = intracellular concentration of the kth ion

$c_{o,k}$  = extracellular concentration of the kth ion

– Nernst-Planck Equation for neuron:

ionic flux

$$\bar{j}_{kD} + \bar{j}_{kE} = -D_k \left( \nabla c_k + \frac{c_k z_k F}{RT} \nabla \Phi \right)$$

ionic current density

$$-D_k z_k F \left( \nabla c_k + \frac{c_k z_k F}{RT} \nabla \Phi \right)$$

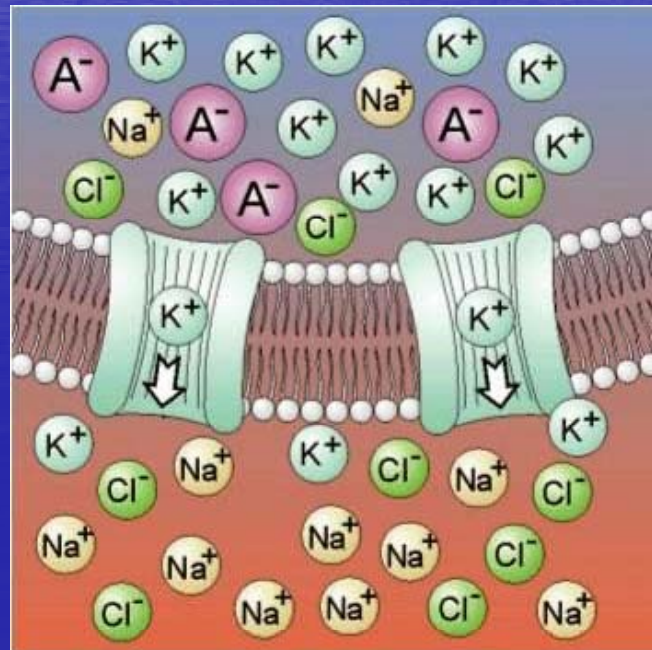
– Equilibrium condition for current density:

$$\bar{J}_k = 0 \Rightarrow$$

$$\nabla c_k = -\frac{c_k z_k F}{RT} \nabla \phi$$

$$V_k = -\frac{z_k F}{RT} (\Phi_o - \Phi_i)$$

$$V_k = -\frac{RT}{z_k F} \ln \frac{c_{i,k}}{c_{o,k}}$$



## Equivalent electric circuit for neural cell

A neuron membrane behaves as a capacitor with many conducting channels joining its two sides, the extracellular and the cytoplasmic ones.

An equivalent electric circuit is presented below and the Kirchhoff laws

$$\nabla \vec{j} = 0 \Rightarrow$$

$$\sum_{k=1}^n \varepsilon_k I_k = 0$$

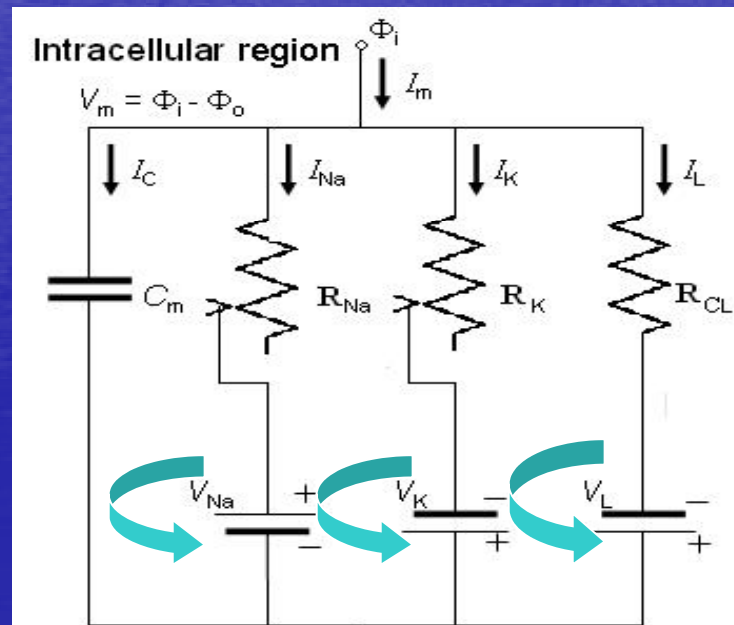
$$\frac{1}{\sigma} \vec{j} = \vec{E} \Rightarrow$$

$$\sum_{k=1}^n \varepsilon_k R_k I_k = \sum_{k=1}^m U_k$$

With the concrete notations:

$$I_C + I_{Na} + I_K + I_L = 0$$

$$\frac{dV}{dt} = -g_{Na}(V - V_{Na}) - g_K(V - V_K) - g_L(V - V_L) + I_m$$

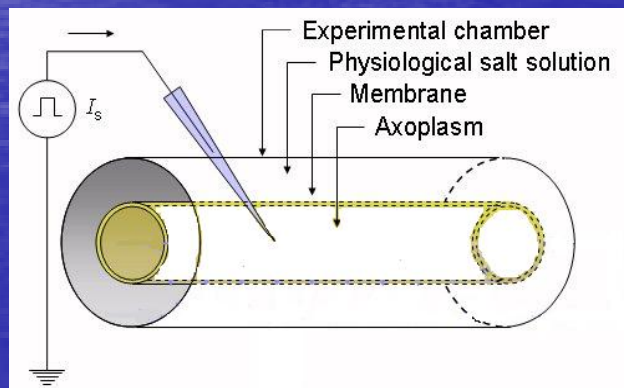


# Cable equation for the axon

## Assumptions:

the voltage depends on the point where is calculated  
the flow travel with constant velocity and the pulse maintains its original form during the propagation.

$$U = U(x, t)$$



**Conclusion:** the dynamics obeys the wave equation and the

- The propagation equation has the form:

$$\frac{\partial^2 U}{\partial x^2} = \frac{1}{\Theta^2} \frac{\partial^2 U}{\partial t^2}$$

- Suppose the variation of the potential along the axon satisfies the relation:

$$\frac{\partial U}{\partial x} = I_i r_i + I_0 r_0$$

With new derivation of the previous relation we get:

$$\frac{\partial^2 U}{\partial x^2} = r_i \frac{\partial I_i}{\partial x} + r_0 \frac{\partial I_0}{\partial x}$$

The conservation of total charge imposes

$$\frac{\partial I_i}{\partial x} = \frac{\partial I_0}{\partial x} = v_m$$

- Substituting this last equation in the wave equation, we obtain

$$(r_i + r_0) v_m = \frac{1}{\Theta^2} \frac{\partial^2 U}{\partial t^2} - \frac{\partial^2 U}{\partial x^2}$$

## Neural networks. Hopfield type models

Proposed in 1982 by John Hopfield to explain the functioning of the memory as an asynchronous neural network.

Consists in  $n$  totally coupled units, that is, each unit is connected to all other units except itself and change information recursively between them.

The individual units are randomly updated, preserving their individual states in the interval between two updates.

### Assumptions:

No synchronization requirements (not a universal time is needed)

The neurons are not updated simultaneously

Symmetric transfer of information among units



The starting point is the Hodgkin-Huxley model and the Kirchhoff law:

$$C_m \frac{dV}{dt} = -g_{Na}(V - V_{Na}) - g_K(V - V_K) - g_L(V - V_L) + I_m$$

**Suppositions:**

- the potential is constant along the axon.
- in the same unit area, there is a delay between the voltage  $U$  on the resistive channels and the capacitive potential  $V$ .

The cable equation of the axon can be written as:

$$C_m \frac{dV}{dt} = -f(V, U, t) + I_m(t)$$
$$\frac{dU}{dt} = g(V, U, t); g(V, U, t) \xrightarrow{U \rightarrow V} 0$$

The previous system gives an expression of the voltage  $V$  in terms of  $U$ .

# Mathematical form of the binary model

Experiments shown that there are two branches:

- a mostly linear branch
- a second nonlinear one

The two branches can be separated by using a binary variable  $S=+1$  or  $-1$ .

From the previous consideration we conclude that the functioning of one neuron can be described in good agreement with the real neuron behavior by the equation:

$$\frac{dU}{dt} = a - b \cdot U(t) + c(1 - S)I_m(t)$$

For many neurons, the Hopfield model proposed the equations:

$$\dot{u}_k(t) = -b_k u_k(t) + \sum_{j=1}^n w_{kj} C(u_j(t)) + i_k(t) \quad k=1, \dots, n$$

## Numerical simulations for the action potential

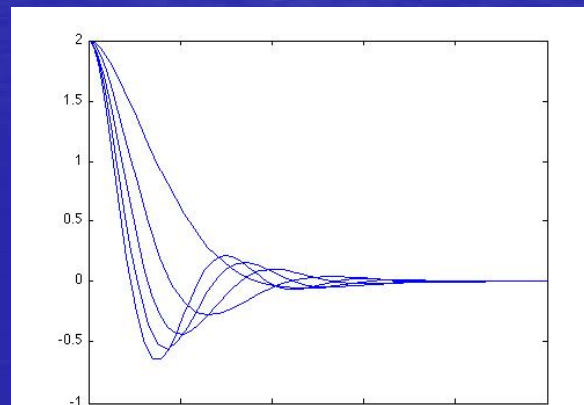
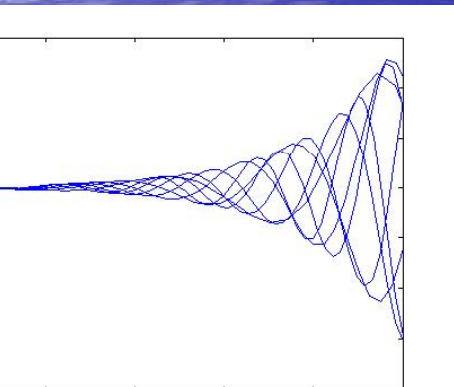
Returning back to the cable equation and passing to the "wave" coordinate  $\tau = t - \frac{x}{c}$ , the previous equation becomes:

$$U'' - AU' - BU = 0$$

where:

$$A = \frac{\tau}{\lambda^2}; B = \frac{1}{\lambda^2}$$

We have numerically studied the behavior of the voltage along the wave, and the results are presented in the figure below:



## Conclusions:

The propagation of neural flow and across neurons is a not yet elucidated problem.

There are many models from very complex as Hodgkin-Huxley, continuing with integrate or fire models, till very simple ones as binary Hopfield-type models.

The reduction of the complexity of a model has to allow detailed studies on network characteristics but not to spoil its biological content.

Using Hopfield model, we described the flow in terms of a simple parameter  $A$ : it expands for small values of  $A$  and strongly damp for

A blue-tinted photograph of a vast ocean under a cloudy sky. The text "Thank you !" is centered in the middle of the image.

Thank you !