

# Brain Inspired Computing (WS 21): Exercise sheet 2

Hand in on 09.11.2021, 14:00

Name(s):

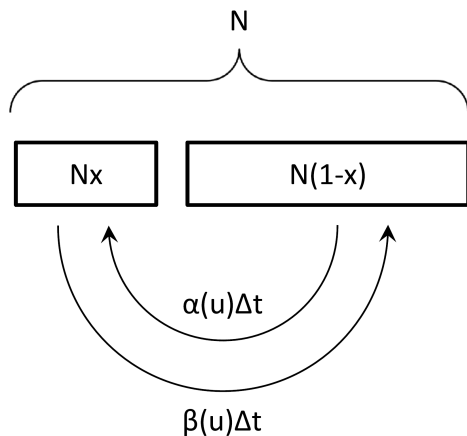
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Question:	1	2	3	4	Total
Points:	40	40	10	10	100
Score:					

## Exercise 1: Channel activation functions (40 points)

Since the variables  $m$ ,  $n$  and  $h$  represent probabilities of particular molecular gates being open, some authors prefer describing them with a different set of equations that more intuitively depict their stochastic switching between “open” and “closed” states.



Assume that a protein is “open” with the probability  $x$ . Out of  $N$  such proteins that are embedded in a membrane,  $Nx$  will be open and  $N(1 - x)$  closed, on average, at any point in time. If, during a very short time interval  $\Delta t$ , the protein may switch from “open” to “closed” with the probability  $\beta_x(u)\Delta t$  and from “closed” to “open” with the probability  $\alpha_x(u)\Delta t$ , then the population of open proteins will lose  $Nx \cdot \beta_x(u)\Delta t$  and gain  $N(1 - x) \cdot \alpha_x(u)\Delta t$  members during this  $\Delta t$ .

- (a) (10 points) Write this down as a differential equation for  $x$  (we can replace  $\Delta$ s with “ $d$ ”s in the limit of  $\Delta t \rightarrow 0$ ).

Hint: the total number of channels does not change, so  $\Delta(Nx) = N\Delta x$ .

- (b) (15 points) Transform the ODE in a) to the form

$$\dot{x} = \frac{1}{\tau_x(u)}[x_0(u) - x]$$

(which we used in the lecture to describe the dynamics of the activation variables  $m$ ,  $n$  and  $h$ ).

What are the required transformations from  $\alpha_x$  and  $\beta_x$  to  $\tau_x$  and  $x_0$ ?

(c) (15 points) Show that with the switching rates

$$\alpha(u) = \frac{1}{1 + e^{-\frac{u+a}{b}}} \quad \text{and} \quad \beta(u) = \frac{1}{1 + e^{\frac{u+a}{b}}}$$

the stationary value of the activation variables  $x$  can be written as

$$x_0(u) = \frac{1}{2} [1 + \tanh [\beta(u - \Theta_{\text{act}})]] .$$

Determine the activation threshold  $\Theta_{\text{act}}$  and the activation slope  $\beta$ .

## Exercise 2: Euler moving forward (1.3 continued) (40 points)

(a) (30 points) **4D case (Hodgkin-Huxley neuron)**

Simulate a Hodgkin-Huxley neuron with forward Euler for a cell with membrane capacitance  $C_m = 1$  nF. As an external stimulus  $I^{\text{ext}}(t)$ , use a step current of amplitude 7.5 nA and width 50 ms. This stimulus is strong enough to elicit spikes. Use integration time step  $\Delta t = 0.01$  ms, and initial conditions  $u(0) = -65$  mV,  $n(0) = 0.3$ ,  $m(0) = 0.1$ ,  $h(0) = 0.6$ . Plot the time course of the input, the membrane potential and the gating variables.

The dynamics of the gating variables  $x \in \{n, m, h\}$  are provided in the table below. They are given in the original formulation of A. Hodgkin and A. Huxley: Gates open and close stochastically with switching probabilities  $\alpha_x(u)$  and  $\beta_x(u)$  respectively. In this formulation, the differential equations of the gating variables read  $\dot{x} = \alpha_x(u)(1 - x) - \beta_x(u)x$ . The formulation used in the lecture (and that you should use in your implementation) is  $\dot{x} = \tau_x^{-1}(u)(x_0(u) - x)$ . Both formulations are equivalent, and  $x_0(u)$  and  $\tau_x^{-1}(u)$  can be expressed in terms of  $\alpha_x(u)$  and  $\beta_x(u)$  as follows:

$$x_0(u) = \alpha_x(u) / (\alpha_x(u) + \beta_x(u)) \quad \text{and} \quad \tau_x^{-1}(u) = \alpha_x(u) + \beta_x(u) \quad . \quad (15)$$

The resulting values for  $\tau_x$  with the equations in the table are in ms. Hint: First define functions for all  $\alpha_x(u)$  and  $\beta_x(u)$  by copying the equations below. Then define functions for all six variables  $x_0(u)$  and  $\tau_x^{-1}(u)$  using the translation eq. (15).

$x$	$E_x$	$g_x$	$x$	$\alpha_x(u[\text{mV}])$	$\beta_x(u[\text{mV}])$
Na	50 mV	120 $\mu\text{S}$	$n$	$\frac{-0.55 - 0.01u}{\exp(-5.5 - 0.1u) - 1}$	$0.125 \exp(-(u + 65)/80.0)$
K	-77 mV	36 $\mu\text{S}$	$m$	$\frac{-4.0 - 0.1u}{\exp(-4.0 - 0.1u) - 1.0}$	$4.0 \exp(-(u + 65)/18.0)$
l	-54.4 mV	0.3 $\mu\text{S}$	$h$	$0.07 \exp(-(u + 65)/20.0)$	$1.0 / (\exp(-3.5 - 0.1u) + 1.0)$

Reversal potentials and conductances.

Gating variable ODEs, given in the original formulation of the HH model as chemical kinetic equations.

(b) (10 points) **Hodgkin-Huxley neuron: Post-inhibitory rebound spike**

Again, use your forward Euler Hodgkin-Huxley integrator to reproduce another effect mentioned in the lecture called “Post-inhibitory rebound”. To this end, apply a step current of amplitude -3.0 nA for 30 ms. Plot the stimulus, the membrane potential and the gating variables.

**Exercise 3: LIF Firing Rate** (10 points)

Derive the following equation for the firing rate  $\nu$  of a LIF neuron stimulated by the constant current  $I^{\text{ext}}$ :

$$\nu(I^{\text{ext}}) = \left( \tau_{\text{ref}} + \tau_{\text{m}} \ln \left( \frac{E_{\text{reset}} - E_{\text{leak}} - \frac{I^{\text{ext}}}{g_{\text{l}}}}{\theta - E_{\text{leak}} - \frac{I^{\text{ext}}}{g_{\text{l}}}} \right) \right)^{-1}, \quad (16)$$

with the refractory period  $\tau_{\text{ref}}$ , the membrane time constant  $\tau_{\text{m}}$ , the reset potential  $E_{\text{reset}}$ , the resting potential  $E_{\text{leak}}$ , the firing threshold  $\theta$  and the leak conductance  $g_{\text{l}}$ .

**Exercise 4: Asymptotic behavior of LIF neurons** (10 points)

Use the activation function  $\nu(I^{\text{ext}})$  of LIF neurons in response to a constant input current  $I^{\text{ext}}$  from the previous exercise. For vanishing refractory period,  $\tau_{\text{ref}} = 0$ , the activation  $\nu(I^{\text{ext}})$  diverges in the limit of large input currents  $I^{\text{ext}} \rightarrow \infty$ . However, this divergence happens linearly (the activation function has an oblique asymptote).

Find the asymptotic behavior of  $\nu(I^{\text{ext}})$  as  $I^{\text{ext}} \rightarrow \infty$  in the limit of  $\tau_{\text{ref}} \rightarrow 0$ . Give the full function in the form  $\nu_{\text{asyp}}(I^{\text{ext}}) = K \cdot I^{\text{ext}} + C$ , with the slope  $K$  and the y-intercept  $C$ .

Hint: Consider the Taylor series for  $\ln(1 - y)$  and  $\frac{1}{1 - y}$ :

$$\ln(1 - y) = -y - \frac{y^2}{2} - \frac{y^3}{3} + O(y^4), \quad (32)$$

$$\frac{1}{1 - z} = 1 + z + z^2 + z^3 + O(z^4). \quad (33)$$