

# Neurodynamics - Fall 2019

## BENG 260 / BGGN 260 / PHYS 279

Homework 4: Due November 4

### 1 Computational Lab

In this assignment we will explore the synchronization of two Hodgkin-Huxley model neurons connected by reciprocal, inhibitory synapses. For both neurons use the full HH model.

The equations describing the HH dynamics are replicated here for convenience (**Note: Parameters and equations are slightly changed for a baseline voltage of -70 mV, it is recommended you use these parameters**):

$$\frac{dV}{dt} = \frac{1}{C} (-I_{Na} - I_K - I_L + I_{ext}) \quad (1)$$

$$I_{Na} = g_{Na} m^3 h (V - E_{Na}) \quad (2)$$

$$I_K = g_K n^4 (V - E_K) \quad (3)$$

$$I_L = g_L (V - E_L) \quad (4)$$

with parameters:

$$\begin{aligned} C &= 1 \mu F/cm^2 \\ E_{Na} &= 45 mV; & g_{Na} &= 120 mS/cm^2 \\ E_K &= -82 mV; & g_K &= 36 mS/cm^2 \\ E_L &= -59.387 mV; & g_L &= 0.3 mS/cm^2 \end{aligned} \quad (5)$$

where the dynamics of gating variables:

$$\frac{dm}{dt} = \alpha_m(V) (1 - m) - \beta_m(V) m \quad (6)$$

$$\frac{dh}{dt} = \alpha_h(V) (1 - h) - \beta_h(V) h \quad (7)$$

$$\frac{dn}{dt} = \alpha_n(V) (1 - n) - \beta_n(V) n \quad (8)$$

is determined by rate functions (note: these equations are different from Homework 2):

$$\alpha_m(V) = 0.1(V + 45)/(1 - \exp(-(V + 45)/10)) \quad (9)$$

$$\beta_m(V) = 4 \exp(-(V + 70)/18) \quad (10)$$

$$\alpha_h(V) = 0.07 \exp(-(V + 70)/20) \quad (11)$$

$$\beta_h(V) = 1/(1 + \exp((-V + 40)/10)) \quad (12)$$

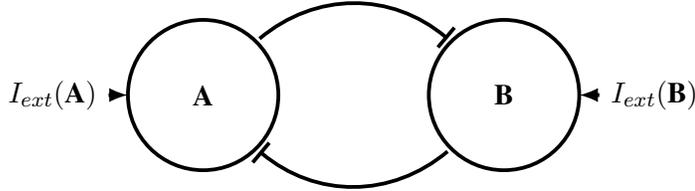
$$\alpha_n(V) = 0.01(V + 60)/(1 - (\exp((-V + 60)/10) - 1)) \quad (13)$$

$$\beta_n(V) = 0.125 \exp(-(V + 70)/80) \quad (14)$$

1. *Uncoupled Neurons* [15 points]. Create two Hodgkin-Huxley model neurons and run the simulation of both uncoupled neurons for 500 ms, injecting a constant  $10 \mu A/cm^2$  current into one neuron and  $20 \mu A/cm^2$  into the other. Plot the last 100ms of the simulations and explain your results focusing on how their spiking frequencies and phases.

*Hint: It is convenient to vectorize variables so that  $V[1], m[1], \dots$  are for neuron A and  $V[2], m[2], \dots$  are for neuron B.*

2. *Inhibitory Synaptic Current* [15 points].



Add reciprocal, inhibitory ( $GABA_A$ ) synapses between the two neurons:

$$I_{syn} = g_{GABA_A} r (V_{post} - E_{Cl}) \quad (15)$$

with receptor channel kinetics  $r$  governing the [synaptic dynamics](#):

$$\frac{dr}{dt} = \alpha_r [T] (1 - r) - \beta_r r \quad (16)$$

$$[T] = [T]_{max} / (1 + \exp(-(V_{pre} - V_p)/K_p)) \quad (17)$$

with:

$$\begin{aligned} E_{Cl} &= -80 \text{ mV} \\ \alpha_r &= 5 \text{ mM}^{-1} \text{ ms}^{-1}; \quad \beta_r = 0.18 \text{ ms}^{-1} \\ [T]_{max} &= 1.5 \text{ mM} \\ K_p &= 5 \text{ mV}; \quad V_p = 7 \text{ mV} \end{aligned} \quad (18)$$

In these equations,  $V_{pre}$  and  $V_{post}$  are the pre- and postsynaptic membrane voltage, which are  $V(1)$  and  $V(2)$  for the first synapse, and  $V(2)$  and  $V(1)$  for the second synapse, respectively.

Run the simulation of the synaptically coupled neurons a number of times, increasing the peak synaptic conductance  $g_{GABA_A}$  from zero to  $0.5 \text{ mS/cm}^2$  by steps of  $0.1 \text{ mS/cm}^2$  while injecting  $10 \mu A/cm^2$  into one neuron and  $20 \mu A/cm^2$  into the other. Plot the last 100ms of each simulation and describe how the spiking frequency and phase change with changing  $g_{GABA_A}$ . It can be helpful to plot  $r_1$  and  $r_2$  as well as  $V_1$  and  $V_2$ .

Next, increase  $g_{GABA_A}$  from  $0.5 \text{ mS/cm}^2$  to  $3.5 \text{ mS/cm}^2$  by steps of  $0.5 \text{ mS/cm}^2$ . Plot the last 100ms of each simulation and describe the changing behavior. Why does this phenomena happen?

Plot the spiking frequency of the two neurons as a function of  $g_{GABA_A}$ . In general, it is a good idea to have your program estimate the frequency (spike count over a time interval) after the neurons have had time to recover (e.g. after 250 ms). You can use the `isi` function to calculate the average and standard deviation of the interspike intervals.

3. *In-phase oscillations* [15 points].

When the current injected into the two neurons is more similar, another interesting phenomenon can be observed. Set the  $I_{ext}$  currents to 10.0 and 10.1  $\mu A/cm^2$  respectively and hold  $g_{GABA_A}$

at  $1.0 \text{ mS/cm}^2$ . Run multiple simulations, decreasing the value of the backward rate constant,  $\beta_r$ , from  $0.2 \text{ ms}^{-1}$  to  $0.1 \text{ ms}^{-1}$  by steps of  $0.01 \text{ ms}^{-1}$ , plot the last 100ms of each simulation.

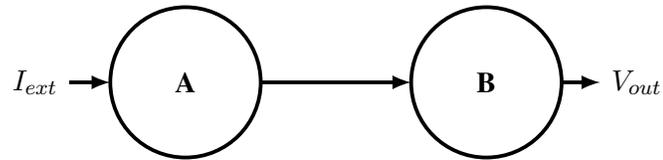
This increases the decay time of the current, and at some value of  $\beta_r$ , the neurons should settle into a nearly in-phase (as opposed to anti-phase) spiking pattern. Why do you think this happens?

Plot the phase of the two neurons as a function of  $\beta_r$ . You can use the `spk_phase` function to calculate the phase between spiking patterns.

## 2 Homework

Expanding on the two state model, you will develop a model for the dynamics of an excitatory post-synaptic current and use it to investigate the dynamics of two 3-neuron network motifs.

4. *Excitatory synapse model* [20 points].

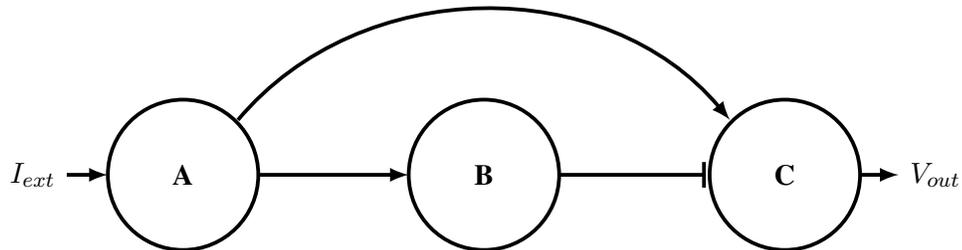


Create an excitatory synapse using the same form as the inhibitory synapse but with the parameters:

$$\begin{aligned}
 E &= -38 \text{ mV} \\
 \alpha_r &= 2.4 \text{ mM}^{-1}\text{ms}^{-1}; \quad \beta_r = 0.56 \text{ ms}^{-1} \\
 [T]_{max} &= 1.0 \text{ mM} \\
 g_{Glu} &= 0 \text{ to } 0.5 \text{ mS/cm}^2
 \end{aligned} \tag{19}$$

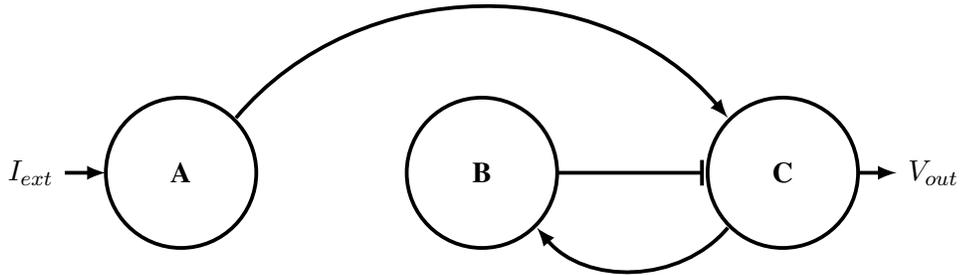
Test the excitatory synapse by injecting current into **A** and recording spike(s) in **B**. First simulate with  $I_{ext} = 10 \mu\text{A/cm}^2$  and  $g_{Glu_{AB}} = 0.3$ . Plot the results and compare the spike rates in **A** and **B**. Try different values of  $g_{Glu_{AB}}$  and compare the spike rates in **A** and **B**. Tuning the strength of the connections will be important for the subsequent parts.

5. *Feedforward inhibition* [20 points].



Feedforward inhibition is when a primary neuron has input current into an inhibitory neuron as well as your output neuron. First simulate with  $I_{ext} = 10 \mu\text{A/cm}^2$  and  $g_{GABA_{BC}} = 1$ ,  $g_{Glu_{AB}} = g_{Glu_{AC}} = 0.4$ . Plot the results and comment on how this connectivity affects the dynamics of the input/output function? What is the relationship between the spike train in neuron **A** vs. **C** for various currents? You should play around with  $g_{Glu}$  and  $g_{GABA_A}$  to alter the connection strengths of the network.

6. Feedback inhibition [10 points]

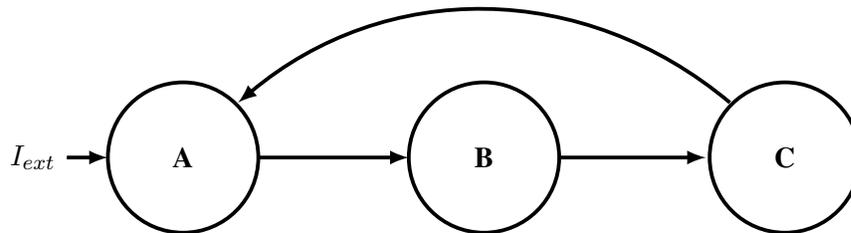


Feedback inhibition occurs through the connectivity shown below. First simulate with  $I_{ext} = 10\mu A/cm^2$  and  $g_{GABA_{BC}} = 1$ ,  $g_{Glu_{AC}} = g_{Glu_{CB}} = 0.5$ . Plot the results. Try other connectivity weights and comment on how the spike frequency of the output varies with the input?

7. Function of mini-networks [5 points]

Explore and describe the potential uses of these feedforward and feedback network motifs in neuroscience and other fields.

8. Loop [Bonus Problem: 20 points]



You can connect many cells in a loop with excitatory synapses to get continual firing from just a small pulse to one cell ( $10\mu A/cm^2$  for  $1ms$ ). The more cells you have the easier it is to do. A 5-cell loop can be done with just changing  $g_{Glu}$ . A 4-cell loop needs a few more tweaks. A 3-cell loop should be possible. Once you get continual firing, what can you add to stop it?

### Submission Guidelines

Solutions without work or explanations where applicable will receive no credit. Submit a single .zip file containing a single PDF with all your solutions, plots, and any handwritten code as well as your Matlab/Python code to both computational lab and homework problems by 3:00pm of due date on Canvas.

The submission file should follow the naming scheme `LastFirst_A12345678_HW2.zip`