

# Computational Neurodynamics

## Solutions for Exercise Sheet 2 (Unassessed) Neural populations and Braitenberg vehicles

### Solutions to written questions

*Question 4.*

- a) The LIF model is comprised of three equations:

$$\tau \frac{dv}{dt} = v_r - v + RI$$
$$\text{if } v \geq \vartheta \text{ and } t - t_{\text{spike}} > \alpha \text{ then } \begin{cases} v \leftarrow v_r \\ t_{\text{spike}} \leftarrow t \end{cases}$$

The first equation describes the temporal evolution of the membrane potential in the absence of a spike. The second equation describes the behaviour of the neuron as the potential reaches a threshold  $\theta$ , at which point the neuron is considered to spike and the voltage is reset to its resting value  $v_r$ . This equation also describes the refractory period of the neuron, whereby the neuron is prevented from spiking more than once in a time interval of duration  $\alpha$ .

- b) Given the simplicity of the ODE of the LIF neuron, we can analytically calculate its time course  $v(t)$ . We do this by first separating the ODE into  $v$  and  $t$ :

$$\frac{1}{\tau} dt = \frac{dv}{v_r - v + RI}$$

We now perform a definite integral on both sides of the equation. On the RHS, we integrate from the resting potential (where the neuron gets reset after a spike)  $v = v_r$  to the firing threshold  $v = \theta$ . On the LHS, we integrate from  $t = 0$  to the critical time  $t_c$  when the neuron reaches the threshold.

$$\int_0^{t_c} \frac{dt}{\tau} = \int_{v_r}^{\theta} \frac{dv}{v_r - v + RI}$$

$$\frac{t_c}{\tau} = \ln(RI) - \ln(v_r - \theta + RI)$$

Solving for  $t_c$  in this equation will show what time the neuron reaches the threshold. Plugging the parameters on the RHS yields  $\ln(RI) - \ln(v_r - \theta + RI) = 1.4$ , and solving for  $t_c$  shows that  $t_c = 7\text{ms}$ .

Finally, if the neuron spikes once every 7ms, we conclude that its firing rate will be  $\frac{1}{7 \text{ ms}} = 0.142 \text{ kHz} = 142 \text{ Hz}$ .

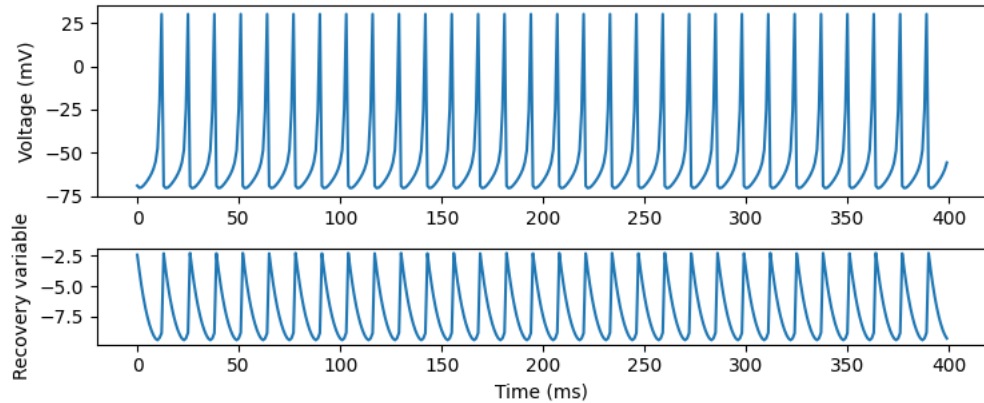
*Question 5.*

- a) The missing element is the spike update (or reset) rule, that is run whenever the neuron records a spike:

$$\text{if } v \geq 30 \text{ then } \begin{cases} v \leftarrow c \\ u \leftarrow u + d \end{cases}$$

This is necessary to reproduce spiking dynamics appropriately – without it,  $v$  would diverge to infinity and the recovery variable would not be able to reproduce the refractory period of real neurons.

- b) If we were to increase  $a$ , the recovery variable  $u$  would decay much faster, allowing the voltage to increase more quickly following a spike, and thereby effectively shortening the neuron's refractory period. This would result in a neuron with a higher firing rate. A neuron with increased  $a$  would behave as follows:



*Question 6.*

- a) The parameter  $d$  represents the after-spike jump of the recovery variable  $u$ . Increasing  $d$  would result in a larger jump, therefore a longer refractory period, and less likelihood that the neuron will fire twice in a short period of time.
- b) In this scenario the recovery variable  $u$  does not play any role, so the model simplifies to a Quadratic Integrate and Fire (QIF) model.

By solving the second-degree equation  $0.04v^2 + 5v + 140 = 0$ , we can rewrite the voltage ODE as

$$\frac{dv}{dt} = 0.04(v + 42.3)(v + 82.7)$$

This motivates the definition of two values: a resting potential  $v_r = -82.7\text{mV}$  and a critical potential  $v_c = -42.3\text{mV}$ . The neuron will show different behaviours depending on the relationship between  $v(0)$ ,  $v_r$ , and  $v_c$ :

- If  $v(0) < v_r$ , then  $dv/dt > 0$  and the voltage will monotonically increase until a stable equilibrium is achieved at  $v = v_r$ .
- If  $v_r < v(0) < v_c$ , then  $dv/dt < 0$  and the voltage will monotonically decrease until a stable equilibrium is achieved at  $v = v_r$ .
- If  $v(0) > v_c$ , then  $dv/dt > 0$  and the voltage will diverge to infinity.