

The code for this homework is published at:

<https://github.com/trienthuyendang/M6397-StochasticProcesses/tree/master/Homework>

Question 1. a) We follow the argument in Section 8.4 of the book of Gerstner et al.

On the one hand, the transition law P^{trans} , which represents the probability density of finding a membrane potential u at time $t + \Delta t$ if at time t membrane potential is u' , is given by

$$P^{\text{trans}}(u, t + \Delta t | u', t) = \left[1 - \Delta t \sum_k \nu_k(t) \right] \delta(u - u' e^{-\Delta t / \tau_m}) + \Delta t \sum_k \nu_k(t) \delta(u - u' e^{-\Delta t / \tau_m} - w_k). \quad (1)$$

On the other hand, we know that the membrane potential is given by a differential equation (i.e., (8.20) in the book) with Poisson distribution input spikes. So Picard's theorem and continuation of solutions implies that the evolution of the membrane potential is a Markov process (the memoryless property comes from the fact that we can continue the solution trajectory by regarding the old end time/value as the new initial time/value). So the probability that membrane potential is u at time $t + \Delta t$ is given by Chapman-Kolmogorov equation,

$$p(u, t + \Delta t) = \int_{\mathbb{R}} P^{\text{trans}}(u, t + \Delta t | u', t) p(u', t) du'. \quad (2)$$

Substituting (1) into (2), we obtain,

$$\begin{aligned} p(u, t + \Delta t) &= \left[1 - \Delta t \sum_k \nu_k(t) \right] \int_{\mathbb{R}} \delta(u - u' e^{-\Delta t / \tau_m}) p(u', t) du' \\ &\quad + \Delta t \sum_k \nu_k(t) \int_{\mathbb{R}} \delta(u - u' e^{-\Delta t / \tau_m} - w_k) p(u', t) du' \\ &= \left[1 - \Delta t \sum_k \nu_k(t) \right] e^{\Delta t / \tau_m} p(e^{\Delta t / \tau_m} u, t) \\ &\quad + \Delta t \sum_k \nu_k(t) e^{\Delta t / \tau_m} p(e^{\Delta t / \tau_m} (u - w_k), t). \end{aligned} \quad (3)$$

Here we use two changes of variables: $t_1 = u - u' e^{-\Delta t / \tau_m}$ for the first integral and $t_2 = u - u' e^{-\Delta t / \tau_m} - w_k$ for the second integral, then apply $\int_{\mathbb{R}} \delta(s) f(s) ds = f(0)$ and $\delta(\alpha s) = |\alpha|^{-1} \delta(s)$.

Since $|\Delta t| \ll 1$, we have the Taylor expansions

$$\begin{aligned} e^{\Delta t / \tau_m} &= 1 + \frac{\Delta t}{\tau_m} + O(\Delta t^2), \\ p(e^{\Delta t / \tau_m} u, t) &= p(u, t) + \frac{\Delta t}{\tau_m} u \frac{\partial}{\partial u} p(u, t) + O(\Delta t^2), \end{aligned}$$

$$p\left(e^{\Delta t/\tau_m}(u - w_k), t\right) = p(u - w_k, t) + \frac{\Delta t}{\tau_m}(u - w_k) \frac{\partial}{\partial u} p(u - w_k, t) + O(\Delta t^2).$$

Thus (3) becomes

$$\begin{aligned} p(u, t + \Delta t) &= e^{\Delta t/\tau_m} p\left(e^{\Delta t/\tau_m} u, t\right) \\ &\quad + \Delta t \sum_k \nu_k(t) e^{\Delta t/\tau_m} \left[p\left(e^{\Delta t/\tau_m}(u - w_k), t\right) - p\left(e^{\Delta t/\tau_m} u, t\right) \right] \\ &= \left(1 + \frac{\Delta t}{\tau_m}\right) \left(p(u, t) + \frac{\Delta t}{\tau_m} u \frac{\partial}{\partial u} p(u, t) \right) \\ &\quad + \Delta t \sum_k \nu_k(t) [p(u - w_k, t) - p(u, t)] + O(\Delta t^2) \\ &= \left(1 + \frac{\Delta t}{\tau_m}\right) p(u, t) + \frac{\Delta t}{\tau_m} u \frac{\partial}{\partial u} p(u, t) \\ &\quad + \Delta t \sum_k \nu_k(t) [p(u - w_k, t) - p(u, t)] + O(\Delta t^2), \end{aligned}$$

which implies

$$\begin{aligned} \frac{p(u, t + \Delta t) - p(u, t)}{\Delta t} &= \frac{1}{\tau_m} p(u, t) + \frac{1}{\tau_m} u \frac{\partial}{\partial u} p(u, t) \\ &\quad + \sum_k \nu_k(t) [p(u - w_k, t) - p(u, t)] + O(\Delta t) \end{aligned}$$

Letting $\Delta t \rightarrow 0$, we conclude that

$$\frac{\partial}{\partial t} p(u, t) = \frac{1}{\tau_m} p(u, t) + \frac{1}{\tau_m} u \frac{\partial}{\partial u} p(u, t) + \sum_k \nu_k(t) [p(u - w_k, t) - p(u, t)] \quad (4)$$

If we further assume that $|w_k| \ll 1$, then by the asymptotic expansion

$$p(u - w_k, t) = p(u, t) - w_k \frac{\partial}{\partial u} p(u, t) + \frac{1}{2} w_k^2 \frac{\partial^2}{\partial u^2} p(u, t) + O(w_k^3),$$

we can write (4) as (ignoring $O(w^3)$),

$$\begin{aligned} \frac{\partial}{\partial t} p(u, t) &= \left[\frac{1}{\tau_m} + \left(\frac{1}{\tau_m} u - \sum_k \nu_k(t) w_k \right) \frac{\partial}{\partial u} \right] p(u, t) + \frac{1}{2} \sum_k \nu_k(t) w_k^2 \frac{\partial^2}{\partial u^2} p(u, t) \\ &= \frac{\partial}{\partial u} \left[\left(\frac{1}{\tau_m} u - \sum_k \nu_k(t) w_k \right) p(u, t) \right] + \frac{1}{2} \sum_k \nu_k(t) w_k^2 \frac{\partial^2}{\partial u^2} p(u, t) \end{aligned}$$

Multiplying both sides by τ_m , we obtain equation (8.41) in the book.

b) If

$$p(u, t) = \frac{1}{\sqrt{2\pi \langle \Delta u^2(t) \rangle}} \exp \left[-\frac{(u(t | \hat{t}) - u_0(t))^2}{2 \langle \Delta u^2(t) \rangle} \right], \quad (5)$$

then

$$\begin{aligned}\frac{\partial}{\partial u}p(u, t) &= \frac{(u_0(t) - u(t | \hat{t}))}{\langle \Delta u^2(t) \rangle^{\frac{3}{2}} \sqrt{2\pi}} \exp \left[-\frac{(u(t | \hat{t}) - u_0(t))^2}{2 \langle \Delta u^2(t) \rangle} \right] \\ \frac{\partial^2}{\partial u^2}p(u, t) &= \frac{(u_0(t) - u(t | \hat{t}))^2 - \langle \Delta u^2(t) \rangle}{\langle \Delta u^2(t) \rangle^{\frac{5}{2}} \sqrt{2\pi}} \exp \left[-\frac{(u(t | \hat{t}) - u_0(t))^2}{2 \langle \Delta u^2(t) \rangle} \right] \\ \frac{\partial}{\partial t}p(u, t) &= \frac{(u_0(t) - u(t | \hat{t})) (u'_0(t) - u'(t | \hat{t}))}{\langle \Delta u^2(t) \rangle^{\frac{3}{2}} \sqrt{2\pi}} \exp \left[-\frac{(u(t | \hat{t}) - u_0(t))^2}{2 \langle \Delta u^2(t) \rangle} \right]\end{aligned}$$

Plugging into (8.41), we see that p defined as in (5) satisfies this equation. □

Question 2. a) The expected time for a LIF neuron to fire satisfies the equation

$$\begin{aligned}A(x)\partial_x T(x) + \frac{1}{2}B(x)\partial_x^2 T(x) &= -1, \\ T(0) = T(\theta) &= 0.\end{aligned}\tag{6}$$

Using the formula in slide 7 lecture 16, we get $A(x) = RI_0 - x$, $B(x) = \sigma^2$.

b) We use the codes of Group 4 - Challenge 6 to plot the histogram of 1000 trials:

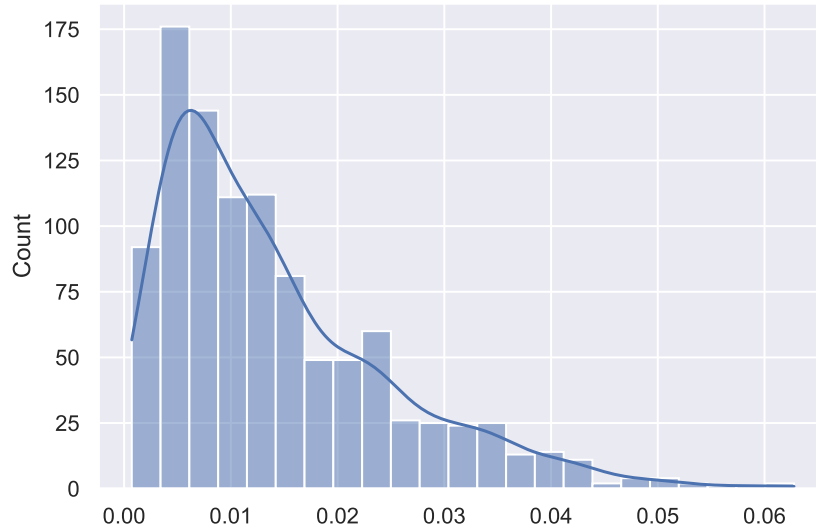


Figure 1 – Histogram of expected time for a LIF neuron to fire. □

Question 3. a) If the times between spikes are distributed according to a gamma distribution $\Gamma(\alpha, \beta)$, then the coefficient of variation of the inter-spike interval is

$$c_v = \frac{\sigma}{\mu} = \frac{\sqrt{\frac{\alpha}{\beta^2}}}{\frac{\alpha}{\beta}} = \frac{1}{\sqrt{\alpha}}.$$

b) If inter-event time follows the Erlang distribution with fixed λ and increasing k , then the coefficient of variation of the inter-spike interval is

$$c_v = \frac{\sigma}{\mu} = \frac{\sqrt{\frac{k}{\lambda^2}}}{\frac{k}{\lambda}} = \frac{1}{\sqrt{k}}.$$

The Erlang distribution is a special case of Gamma distribution, i.e., if we let the shape parameter $\alpha = k$, where the latter is a positive integer, then $\text{Gamma}(k, \lambda) \sim \text{Erlang}(k, \lambda)$. In this case, we get the same coefficient of variation. □

Question 4. Let $n = Nx$, $u = 2\mathcal{U}\frac{1}{N}$, $v = 2\mathcal{V}\frac{1}{N}$, we have

$$\begin{aligned} T(n+1 | n) &= (1-u)x(1-x) + v(1-x)^2 \\ &= \left(1 - 2\mathcal{U}\frac{1}{N}\right)x(1-x) + 2\mathcal{V}\frac{1}{N}(1-x)^2, \\ T(n-1 | n) &= (1-v)(1-x)x + ux^2 \\ &= \left(1 - 2\mathcal{V}\frac{1}{N}\right)x(1-x) + 2\mathcal{U}\frac{1}{N}x^2, \\ T(n | n+1) &= (1-v)\left(1-x-\frac{1}{N}\right)\left(x+\frac{1}{N}\right) + u\left(x+\frac{1}{N}\right)^2 \\ &= \left(1 - 2\mathcal{V}\frac{1}{N}\right)\left[x(1-x) + (1-2x)\frac{1}{N} - \frac{1}{N^2}\right] + 2\mathcal{U}\frac{1}{N}\left(x^2 + 2x\frac{1}{N} + \frac{1}{N^2}\right) \\ &= x(1-x) + [(1-2x) - 2\mathcal{V}x(1-x) + 2\mathcal{U}x^2]\frac{1}{N} \\ &\quad + [-2\mathcal{V}(1-2x) - 1 + 4\mathcal{U}x]\frac{1}{N^2} + O\left(\frac{1}{N^3}\right), \\ T(n | n-1) &= (1-u)\left(x-\frac{1}{N}\right)\left(1-x+\frac{1}{N}\right) + v\left(1-x+\frac{1}{N}\right)^2 \\ &= \left(1 - 2\mathcal{U}\frac{1}{N}\right)\left[x(1-x) + (2x-1)\frac{1}{N} - \frac{1}{N^2}\right] + 2\mathcal{V}\frac{1}{N}\left((1-x)^2 + 2(1-x)\frac{1}{N} + \frac{1}{N^2}\right) \\ &= x(1-x) + [(2x-1) - 2\mathcal{U}x(1-x) + 2\mathcal{V}(1-x)^2]\frac{1}{N} \\ &\quad + [2\mathcal{U}(1-2x) - 1 + 4\mathcal{V}(1-x)]\frac{1}{N^2} + O\left(\frac{1}{N^3}\right). \end{aligned}$$

To avoid abusing notation, let

$$Q(x, t) := P(Nx, t) = P(n, t),$$

then

$$\begin{aligned} \frac{dP(n, t)}{dt} &= \frac{\partial}{\partial t}Q(x, t), \\ P(n+1, t) &= Q\left(x + \frac{1}{N}, t\right) \end{aligned}$$

$$\begin{aligned}
&= Q(x, t) + \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) + O\left(\frac{1}{N^3}\right) \\
P(n-1, t) &= Q\left(x - \frac{1}{N}, t\right) \\
&= Q(x, t) - \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) + O\left(\frac{1}{N^3}\right)
\end{aligned}$$

We now compute each term in the right hand side of the master equation:

$$\begin{aligned}
&P(n, t) [T(n+1 | n) + T(n-1 | n)] \\
&= Q(x, t) [(1-u)x(1-x) + v(1-x)^2 + (1-v)(1-x)x + ux^2] \\
&= [2x(1-x) + (2x-1)(ux - v(1-x))] Q(x, t) \\
&= \left[2x(1-x) + (2x-1) \left(2\mathcal{U} \frac{1}{N} x - 2\mathcal{V} \frac{1}{N} (1-x) \right) \right] Q(x, t), \\
&P(n+1, t) T(n | n+1) \\
&= \left(Q(x, t) + \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) \right) \\
&\quad \times \left[(1-v) \left(1-x - \frac{1}{N} \right) \left(x + \frac{1}{N} \right) + u \left(x + \frac{1}{N} \right)^2 \right] + O\left(\frac{1}{N^3}\right) \\
&= \left(Q(x, t) + \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) \right) \\
&\quad \times \left\{ x(1-x) + [(2x-1) - 2\mathcal{U}x(1-x) + 2\mathcal{V}(1-x)^2] \frac{1}{N} \right. \\
&\quad \left. + [2\mathcal{U}(1-2x) - 1 + 4\mathcal{V}(1-x)] \frac{1}{N^2} \right\} + O\left(\frac{1}{N^3}\right), \\
&P(n-1, t) T(n | n-1) \\
&= \left(Q(x, t) - \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) \right) \\
&\quad \times \left[(1-v) \left(1-x - \frac{1}{N} \right) \left(x + \frac{1}{N} \right) + u \left(x + \frac{1}{N} \right)^2 \right] + O\left(\frac{1}{N^3}\right) \\
&= \left(Q(x, t) - \frac{1}{N} \frac{\partial}{\partial x} Q(x, t) + \frac{1}{2N^2} \frac{\partial^2}{\partial x^2} Q(x, t) \right) \\
&\quad \times \left\{ x(1-x) + [(2x-1) - 2\mathcal{U}x(1-x) + 2\mathcal{V}(1-x)^2] \frac{1}{N} \right. \\
&\quad \left. + [2\mathcal{U}(1-2x) - 1 + 4\mathcal{V}(1-x)] \frac{1}{N^2} \right\} + O\left(\frac{1}{N^3}\right).
\end{aligned}$$

Substituting into the master equation and simplifying, we obtain

$$\frac{\partial}{\partial t} Q(x, t) = \frac{1}{N^2} \frac{\partial^2}{\partial x^2} [x(1-x)Q(x, t)] + \frac{2}{N^2} \frac{\partial}{\partial x} \{ [\mathcal{U}x - \mathcal{V}(1-x)] Q(x, t) \} + O\left(\frac{1}{N^3}\right).$$

Let $\tau = \frac{2t}{N^2}$ be the rescaled time unit, then $\frac{\partial}{\partial t} = \frac{2}{N^2} \frac{\partial}{\partial \tau}$, so the above equation becomes

$$\frac{\partial}{\partial \tau} Q(x, \tau) = \frac{1}{2} \frac{\partial}{\partial x^2} [x(1-x)Q(x, \tau)] + \frac{\partial}{\partial x} \{[\mathcal{U}x - \mathcal{V}(1-x)] Q(x, \tau)\} + O\left(\frac{1}{N}\right).$$

Passing to limit $N \rightarrow \infty$, we obtain the Fokker-Planck equation,

$$\frac{\partial}{\partial \tau} Q(x, \tau) = \frac{1}{2} \frac{\partial}{\partial x^2} [x(1-x)Q(x, \tau)] + \frac{\partial}{\partial x} \{[\mathcal{U}x - \mathcal{V}(1-x)] Q(x, \tau)\}.$$

□