Neural Networks and Biological Modeling

Professor Wulfram Gerstner Laboratory of Computational Neuroscience

Answers to question set 11

Exercise 1: Poisson neuron

1.1 We present two methods to solve this problem.

Method 1: The probability that the neuron does not fire during a *small* time interval Δt is given by $S(\Delta t) = 1 - \rho \Delta t$. Since a Poisson process is independent of its past history, the probability that the neuron does not fire during *n* such time intervals is the product of the probabilities for each time intervals, i.e.,

$$S(n\Delta t) = (1 - \rho\Delta t)^n \,. \tag{1}$$

Although this expression is correct for a discrete process, it has the drawback of being dependent on the discretization time step Δt . Thus it is desirable to take the limit as $\Delta t \to 0$. This can be done by setting $t = n\Delta t$ and taking the limit as $n \to \infty$ with t fixed. Remembering the formula $\lim_{n\to\infty} (1 + \frac{a}{n})^n = e^a$, one concludes that

$$S(t) = \lim_{n \to \infty} \left(1 - \frac{\rho t}{n} \right)^n = e^{-\rho t} \,. \tag{2}$$

Alternatively, one can use the identity

$$(1 - \rho\Delta t)^n = \exp\left[\sum_{i=1}^n \log\left(1 - \rho\Delta t\right)\right],\tag{3}$$

and expand the logarithm as $\log(1 + x) = x + \dots$, which yields

$$S(t) = \lim_{n \to \infty} \exp\left[-\sum_{i=1}^{n} \rho \Delta t\right] \to \exp\left[-\int_{0}^{t} \rho dt\right] = \exp\left[-\rho t\right].$$
(4)

The latter calculation has the advantage that it also works for time dependent rates $\rho = \rho(t)$, which is less obvious from Eq.(2).

Method 2 A different way to obtain this result is to consider the variation of S(t) during a small time interval Δt . Because of independence, we have

$$S(t + \Delta t) = S(t)S(\Delta t), \qquad (5)$$

where $S(\Delta t) = 1 - \rho \Delta t$ by assumption. Rearranging. we obtain

$$\frac{S(t+\Delta t) - S(t)}{\Delta t} = -\rho S(t), \qquad (6)$$

which becomes as $\Delta t \to 0$

$$\frac{d}{dt}S(t) = -\rho S(t), \qquad (7)$$

the solution of which is indeed $S(t) = e^{-\rho t}$.

1.2 Again, due to independence, we have

$$P(t, t + \Delta t) \equiv P \text{ (fire for the first time in } (t, t + \Delta t)) = P \text{ (not fire until } t) \times P \text{ (fire in } (t, t + \Delta t))$$
$$= e^{-\rho t} \times \rho \Delta t.$$
(8)

As $\Delta t \to 0$, this probability vanishes; however, the probability density, defined by p(t)dt = P(t, t + dt), has finite value,

$$p(\text{fire at } t) = \lim_{\Delta t \to 0} \frac{P(t, t + \Delta t)}{\Delta t} = \rho e^{-\rho t} \,. \tag{9}$$

1.3

(i) The interval distribution was calculated earlier, $P(t) = \rho e^{-\rho t}$.

(ii) The probability to observe an interspike interval smaller than 20 ms is

$$P(\text{ISI} < 20\text{ms}) = \int_0^{20\text{ms}} \rho e^{-\rho s} ds = \left[-e^{-\rho s}\right]_{s=0}^{20\text{ms}} = 1 - e^{-20\rho}.$$
 (10)

Due to independence, the probability of getting a burst of two such intervals is just the square of this probability. Thus, for $\rho = 2\text{Hz} = 2 \cdot 10^{-3}\text{ms}^{-1}$, we get $p_{\text{burst}} \simeq 0.0015$, whereas for $\rho = 20\text{Hz}$, $p_{\text{burst}} \simeq 0.109$.

(iii) Given knowledge of the interspike interval distribution and survivor function as a function of the firing rate ρ , the observer can determine the strength of the input with fair confidence after observing a few spikes.

1.4 Let us label the spike trains corresponding to each neuron S_1 and S_2 . The percentage is the number of spikes in S_1 coincident with a spike in S_2 , N_{coinc} , divided by the total number of spikes (N) in spike train one:

$$P = \frac{\langle N_{coinc} \rangle}{N} \,. \tag{11}$$

And $\langle N_{coinc} \rangle$ is just the probability to observe a spike in S_2 within a small observation window size $2\Delta = 4$ ms, times the number of spikes in S_1 :

$$P \approx \frac{2\Delta\rho_0 N}{N} = 2\rho_0 \Delta = 8\%.$$
⁽¹²⁾

Here, we had to assume that the observation windows do not overlap, i. e. $\Delta \ll \rho_0$.

Exercise 2: Stochastic spike arrival

We first need to solve the linear equation

$$\tau \frac{du}{dt} = -(u - u_{\text{rest}}) + RI(t) \tag{13}$$

We know (c.f. exercise set 1) that the solution is given by

$$u(t) = u_{\text{rest}} + \frac{R}{\tau} \int_{-\infty}^{t} e^{-(t-t')/\tau} I(t') dt'.$$
 (14)

Let us first solve the general problem with arbitrary presynaptic current shape $\alpha(t - t^f)$. The case of problem 2.1 then corresponds to the choice $\alpha(t - t^f) = q\delta(t - t^f)$. So for $I(t) = \sum_{f} \alpha(t - t^f)$ we have:

$$u(t) = u_{\text{rest}} + R \int_{-\infty}^{t} \frac{e^{-(t-t')/\tau}}{\tau} \sum_{f} \alpha(t'-t^{f}) dt'.$$
 (15)

Writing $\alpha(t'-t^f) = \int_{-\infty}^{\infty} \alpha(s) \delta(s-(t'-t^f)) ds$, we obtain

$$u(t) = u_{\text{rest}} + R \int_{-\infty}^{t} dt' \int_{-\infty}^{\infty} ds \frac{e^{-(t-t')/\tau}}{\tau} \alpha(s) \sum_{f} \delta(s - (t' - t^{f})) \,. \tag{16}$$

Taking the average over all possible spike trains,

$$\langle u(t) \rangle = u_{\text{rest}} + R \int_{-\infty}^{t} dt' \int_{-\infty}^{\infty} ds \frac{e^{-(t-t')/\tau}}{\tau} \alpha(s) \left\langle \sum_{f} \delta(s - (t' - t^{f})) \right\rangle$$
(17)

because all the deterministic quantities can be pulled out of the average. Now since $\left\langle \sum_{f} \delta(s - (t' - t^{f})) \right\rangle = \nu$,

$$\langle u(t) \rangle = u_{\text{rest}} + R\nu \underbrace{\int_{-\infty}^{t} dt' \frac{e^{-(t-t')/\tau}}{\tau}}_{=1} \int_{-\infty}^{\infty} ds \alpha(s)$$

$$= u_{\text{rest}} + R\nu \int_{-\infty}^{\infty} \alpha(s) ds .$$
(18)

2.1 With $\alpha(t - t^f) = q\delta(t - t^f)$, we obtain:

$$\langle u(t) \rangle = u_{\text{rest}} + R\nu q.$$
 (19)

2.2 The general solution is given by Eq. (18).

Exercise 3: Renewal process

Given an output spike at $t = \hat{t}$, the survivor function $S(t - \hat{t})$ is given by

$$S(t-\hat{t}) = \exp\left[-\int_{\hat{t}}^{t} \rho(t'|\hat{t})dt'\right] = \exp\left[-\int_{\hat{t}}^{t} \rho(t'-\hat{t})dt'\right] = \exp\left[-\int_{0}^{t-\hat{t}} \rho(s)ds\right].$$

where we made the variable change $s = t' - \hat{t}$. The interspike interval distribution is $P(t - \hat{t}) = \rho(t - \hat{t})S(t - \hat{t})$. Thus we only need to calculate the integral of the hazard function $\rho(t - \hat{t})$. This gives

$$\int_{0}^{t-\hat{t}} \rho(s) ds = \begin{cases} \int_{0}^{t_{\rm abs}} \rho(s) ds = 0 & \text{for } s < t_{\rm abs} \\ \int_{0}^{t_{\rm abs}} \rho(s) ds + \int_{t_{\rm abs}}^{t-\hat{t}} \rho(s) ds = \frac{\rho_0}{4} \left(t - \hat{t} - t_{\rm abs}\right)^2 & \text{for } t_{\rm abs} < s < t_{\rm abs} + 2 \\ \int_{0}^{t_{\rm abs}} \rho(s) ds + \int_{t_{\rm abs}}^{t_{\rm abs}+2} \rho(s) ds + \int_{t_{\rm abs}+2}^{t-\hat{t}} \rho(s) ds = \rho_0 \left(-1 + t - \hat{t} - t_{\rm abs}\right) & \text{for } t_{\rm abs} + 2 < s \,. \end{cases}$$

¹ this can be seen by remarking that $\int \delta(s) ds = 1$ so that $\frac{1}{T} \sum_f \int_0^T \delta(s - t^f) ds = \frac{\# \text{ of spikes in } (0,T)}{T} = \nu$.

Exercise 4: Homework

4.1 We take the limit and use Stirling's approximation and $\lim_{n\to\infty}(1-x/n)^n = e^{-x}$:

$$P_k(T) = \lim_{N \to \infty} \frac{N!}{k!(N-k)!} \left(1 - \frac{\nu T}{N}\right)^{N-k} \left(\frac{\nu T}{N}\right)^k \tag{20}$$

$$= \frac{(\nu T)^k}{k!} \lim_{N \to \infty} \frac{N^N e^{-N}}{(N-k)^{N-k} e^{-N+k}} \left(1 - \frac{\nu T}{N}\right)^{N-k} \left(\frac{1}{N}\right)^k \tag{21}$$

$$= \frac{(\nu T)^{k} e^{-k}}{k!} \lim_{N \to \infty} \frac{\left(1 - \frac{\nu T}{N}\right)^{N-k}}{\left(1 - k/N\right)^{N-k}}$$
(22)

$$=\frac{(\nu T)^{k}e^{-k}}{k!}\frac{e^{-\nu T}}{e^{-k}}$$
(23)

$$=\frac{(\nu T)^k}{k!}e^{-\nu T} \tag{24}$$

The expected number of spikes in an interval of duration T can be calculated from the definition of expectation,

$$\langle k \rangle = \sum_{k=0}^{\infty} k P_k(T) \tag{25}$$

$$=\sum_{k=0}^{\infty} k \frac{(\nu T)^{k}}{(k)!} e^{-\nu T}$$
(26)

$$= e^{-\nu T} \sum_{k=1}^{\infty} k \frac{(\nu T)^k}{(k)!}$$
(27)

$$=e^{-\nu T}\sum_{k=1}^{\infty}\frac{(\nu T)^{k}}{(k-1)!}$$
(28)

$$= e^{-\nu T} (\nu T) \sum_{k=0}^{\infty} \frac{(\nu T)^k}{k!}$$
(29)

$$=\nu T.$$
(30)

For the third equality we considered that for k = 0 the sum is 0, so we can start with k = 1. For the fourth equality we performed a change of variables and for the last one we used the definition of the exponential function $e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$.