Notes on Noise and First-Passage Times - Larry Abbott

The Model

The membrane potential V, obeys the stochastic, first-order differential equation.

$$\tau \frac{dV}{dt} = V_{\infty} - V + \eta(t), \qquad (1.1)$$

where $V_{\infty} = V_{\text{rest}} + I$, for constant I. In addition, when V reaches a threshold value, V_{th} , an action potential is generated and V is reset to a lower value V_{reset} .

 η is a white-noise random variable satisfying (see the following section),

$$\langle \eta(t) \rangle = 0$$
 and $\langle \eta(t) \eta(t') \rangle = 2D\delta(t - t')$. (1.2)

D is a constant with dimensions of voltage squared multiplied by time.

The problem is to compute the average time between resets, that is, the average time it takes for V to travel from V_{reset} to reach, for the first time, V_{th} .

White Noise

A white noise stimulus, $\eta(t)$, has zero mean

$$\langle \eta(t) \rangle = 0, \tag{1.3}$$

where the brackets stand for the average over multiple trials. This equation applies for all t values. In addition to the above equation, white-noise is characterized by averages of pairs of η variables. When $t \neq t'$, $\eta(t)$ and $\eta(t')$ are completely independent, so the average value of their product (again averaged over a large number of different white-noise trials), which is their cross-correlation, is zero. This is stated as

$$\langle \eta(t)\eta(t')\rangle = 0 \quad \text{if} \quad t \neq t'.$$
 (1.4)

On the other hand, if t=t', the value of $\langle \eta^2(t) \rangle$ is proportional to $1/\Delta t$, assuming, for the moment, that time is divided into discrete bins. The reason for this proportionality is that it prevents the fluctuations from averaging out in the $\Delta t \to 0$ limit. Consider the average of η over a time period of duration T,

$$\overline{\eta}(t) = \frac{\Delta t}{T} \sum_{i=1}^{T/\Delta t} \eta(t + (i-1)\Delta t). \tag{1.5}$$

This quantity has zero mean, $\langle \overline{\eta} \rangle = 0$, and variance

$$\langle \overline{\eta}^2 \rangle = \left(\frac{T}{\Delta t}\right) \left(\frac{\Delta t}{T}\right)^2 \langle \eta^2 \rangle = \left(\frac{\Delta t}{T}\right) \langle \eta^2 \rangle.$$
 (1.6)

Thus, for consistency across scales, we must require that

$$\langle \eta^2 \rangle \propto \frac{1}{\Delta t}.$$
 (1.7)

The constant of proportionality is written, by convention, as 2*D*.

The function that is zero except when its argument is zero and then is $1/\Delta t$ at that point is the δ function, defined by

$$\int dt \, \delta(t) = 1 \quad \text{and} \quad \int dt \, f(t) \, \delta(t - t') = f(t') \,. \tag{1.8}$$

In other words, $\delta(t-t')$ is zero when $t \neq t'$ and is infinite when t=t' in such a way that its total integral over time is one. Thus,

$$\langle \eta(t)\eta(t')\rangle = 2D\delta(t-t'). \tag{1.9}$$

D is called the diffusion constant and has the units of voltage squared multiplied by time.

Physical processes typically produce filtered versions of white noise. Consider, the membrane potential satisfying 1.1 for a neuron that is not spiking. Then, take t=0 as an arbitrary time point,

$$V(0) = V_{\infty} + \frac{1}{\tau} \int_{-\infty}^{0} dt' \, \eta(t') e^{t'/\tau}, \qquad (1.10)$$

so

$$\langle V(t) \rangle = \langle V(0) \rangle = V_{\infty}$$
 (1.11)

and

$$\sigma_{V}^{2} = \langle (V(t) - V_{\infty})^{2} \rangle = \langle (V(0) - V_{\infty})^{2} \rangle$$

$$= \frac{1}{\tau^{2}} \int_{-\infty}^{0} dt' \int_{-\infty}^{0} dt'' \langle \eta(t') \eta(t'') \rangle e^{(t'+t'')/\tau} = \frac{2D}{\tau^{2}} \int_{-\infty}^{0} dt' e^{2t'/\tau} . \quad (1.12)$$

The variance of the variable *V* from this equation is thus given by

$$\sigma_V^2 = \frac{D}{\tau} \,. \tag{1.13}$$

Suppose we know the value of V at time t, V(t). What will the value be a short time Δt later. We denote this by $V(t+\Delta t)$, but we cannot determine its value unless we know what the random variable η is doing. However, we can compute its average value $\langle V(t+\Delta t)\rangle$. Actually, we will compute the average of $\Delta V=V(t+\Delta t)-V(t)$ instead. To do this we integrate equation 1.1 over the small interval from t to $t+\Delta t$,

$$\int_{t}^{t+\Delta t} dt' \frac{dV(t')}{dt'} = \Delta V = \frac{1}{\tau} \int_{t}^{t+\Delta t} dt' \left(V_{\infty} - V(t') + \eta(t') \right). \tag{1.14}$$

Taking the average of both sides of this equation, and recalling that $\langle \eta \rangle = 0$, we find

$$\langle \Delta V \rangle = \frac{1}{\tau} \int_{t}^{t+\Delta t} dt' \left(V_{\infty} - V(t') \right). \tag{1.15}$$

Finally, for small Δt , we can approximate the integral as the integrand times Δt , so we obtain the final result

$$\langle \Delta V \rangle = (V_{\infty} - V(t)) \left(\frac{\Delta t}{\tau} \right),$$
 (1.16)

where in this and the following equations ΔV stands for $\Delta V(t)$.

We can also compute

$$\langle (\Delta V)^2 \rangle = \frac{1}{\tau^2} \int_t^{t+\Delta t} dt' \int_t^{t+\Delta t} dt'' \left\langle (V_{\infty} - V(t') + \eta(t')) \left(V_{\infty} - V(t'') + \eta(t'') \right) \right\rangle. \tag{1.17}$$

Taking the averages and using equation 1.9 we find

$$\langle (\Delta V)^2 \rangle = \frac{1}{\tau^2} \int_t^{t+\Delta t} dt' \int_t^{t+\Delta t} dt'' \left((V_{\infty} - V(t'))(V_{\infty} - V(t'')) + 2D\delta(t' - t'') \right),$$

$$(1.18)$$

which gives, integrating over the δ function and using the same approximation as before,

$$\langle (\Delta V)^2 \rangle = (V_{\infty} - V(t))^2 \left(\frac{\Delta t}{\tau}\right)^2 + 2\sigma_V^2 \left(\frac{\Delta t}{\tau}\right). \tag{1.19}$$

Furthermore, for small Δt , we can ignore the term of order $(\Delta t)^2$ relative to the term linear in Δt and write

$$\langle (\Delta V)^2 \rangle = 2\sigma_V^2 \left(\frac{\Delta t}{\tau}\right). \tag{1.20}$$

Results 1.16 and 1.20 are needed for the first-passage time calculation.

The Mean First-Passage Time

We denote the average first-passage from an arbitrary value V to the threshold as T(V), so that the mean first-passage time we seek is $T = T(V_{\text{reset}})$. Suppose that, on a given trial, the variable satisfying equation 1.1 moves from V to $V + \Delta V$ in time Δt . On average, the time it takes to get to the threshold from $V + \Delta V$ must be Δt less than the time it takes from V, so

$$\langle T(V + \Delta V) \rangle = T(V) - \Delta t. \tag{1.21}$$

Expanding in a Taylor series,

$$\langle T(V + \Delta V) \rangle \approx T(V) + T'(V) \langle \Delta V \rangle + \frac{1}{2} T''(V) \langle (\Delta V)^2 \rangle,$$
 (1.22)

where the primes denote derivatives with respect to V. Using equations 1.16 and 1.20, we find, from 1.21, that

$$\sigma_V^2(V)T''(V)\left(\frac{\Delta t}{\tau}\right) + (V_\infty - V)T'(V)\left(\frac{\Delta t}{\tau}\right) + \Delta t = 0, \qquad (1.23)$$

or

$$\sigma_V^2 T''(V) + (V_\infty - V)T'(V) + \tau = 0.$$
 (1.24)

Defining

$$f(V) = -\frac{(V_{\infty} - V)^2}{2\sigma_V^2}$$
 so that $f'(V) = \frac{V_{\infty} - V}{\sigma_V^2}$, (1.25)

we can write down the solution to this equation using standard integration factors,

$$T'(V) = -\left(\frac{\tau}{\sigma_V^2}\right) e^{-f(V)} \int_{-\infty}^{V} dy \ .e^{f(y)} , \qquad (1.26)$$

Integrating the above result, we find

$$T(V) = -\left(\frac{\tau}{\sigma_V^2}\right) \int_{V_{\text{th}}}^{V} dx \, e^{-f(x)} \int_{-\infty}^{x} dy \, e^{f(y)} \,, \tag{1.27}$$

where we have imposed the additional boundary condition $T(V_{th}) = 0$, which means that once you are there it takes no time to get there. This means that the answer we seek is

$$T = \left(\frac{\tau}{\sigma_V^2}\right) \int_{V_{\text{reset}}}^{V_{\text{th}}} dx \, e^{-f(x)} \int_{-\infty}^{x} dy \, e^{f(y)}$$

$$= \left(\frac{\tau}{\sigma_V^2}\right) \int_{V_{\text{reset}}}^{V_{\text{th}}} dx \, \exp\left(\frac{(V_\infty - x)^2}{2\sigma_V^2}\right) \int_{-\infty}^{x} dy \, \exp\left(-\frac{(V_\infty - y)^2}{2\sigma_V^2}\right) .$$
(1.28)

Changing variables $y \to y\sqrt{2}\sigma_V + V_{\infty}$ and $x \to x\sqrt{2}\sigma_V + V_{\infty}$, we find

$$T = 2\tau \int_{(V_{\text{reset}} - V_{\infty})/\sqrt{2}\sigma_V}^{(V_{\text{th}} - V_{\infty})/\sqrt{2}\sigma_V} dx \exp(x^2) \int_{-\infty}^x dy \exp(-y^2).$$
 (1.29)

Using the fact that

$$\int_{-\infty}^{x} dy \, \exp(-y^2) = \frac{\sqrt{\pi} (1 + \operatorname{erf}(x))}{2}, \qquad (1.30)$$

we obtain the result

$$T = \tau \sqrt{\pi} \int_{(V_{\text{reset}} - V_{\infty})/\sqrt{2}\sigma_{V}}^{(V_{\text{th}} - V_{\infty})/\sqrt{2}\sigma_{V}} dx \, \exp(x^{2}) \, (1 + \text{erf}(x)) \,. \tag{1.31}$$

Useful Numerical Approximation

The integral in equation 1.31 is difficult to compute numerically because of the nature of the integrand $\exp(x^2)(1 + \operatorname{erf}(x))$. To compute this integral using standard methods, use the following approximation.

$$\exp(x^2)(1 + \operatorname{erf}(x)) \approx \begin{cases} f_1 & \text{if } x \le 0\\ 2\exp(x^2) - f_1 & \text{if } x > 0, \end{cases}$$
 (1.32)

where

$$f_1 = t \exp(\alpha), \qquad t = \frac{1}{1 + 0.5|x|},$$
 (1.33)

and

$$\alpha = a_1 + t(a_2 + t(a_3 + t(a_4 + t(a_5 + t(a_6 + t(a_7 + t(a_8 + t(a_9 + ta_{10}))))))))) (1.34)$$

with

$$a_1 = -1.26551223$$
 $a_2 = 1.00002368$ $a_3 = 0.37409196$ (1.35)
 $a_4 = 0.09678418$ $a_5 = -0.18628806$ $a_6 = 0.27886087$
 $a_7 = -1.13520398$ $a_8 = 1.48851587$ $a_9 = -0.82215223$
 $a_{10} = 0.17087277$