Mathematics for Neuroscientists

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Chapter 27.

Exercise 2. We compute

$$\begin{split} E(j+1) - E(j) &= -(\mathbf{s}^j)^T \mathbf{W} \mathbf{s}^{j+1} + (\mathbf{s}^{j-1})^T \mathbf{W} \mathbf{s}^j + \mathbf{b}^T (\mathbf{s}^{j+1} - \mathbf{s}^{j-1}) \\ &= -(\mathbf{s}^{j+1})^T \mathbf{W}^T \mathbf{s}^j + (\mathbf{s}^{j-1})^T \mathbf{W} \mathbf{s}^j + \mathbf{b}^T (\mathbf{s}^{j+1} - \mathbf{s}^{j-1}) \\ &= -(\mathbf{s}^{j+1} - \mathbf{s}^{j-1})^T (\mathbf{W} \mathbf{s}^j - \mathbf{b}) \end{split}$$

as claimed. Now, if $\mathbf{s}^{j+1} \neq \mathbf{s}^{j-1}$ then $s_i^{j+1} = -s_i^{j-1}$ for some i. There are only two possibilities. If $s_i^{j+1} = 1$ then $(\mathbf{W}\mathbf{s}^j)_i > b_i$ and so

$$(\mathbf{s}^{j+1} - \mathbf{s}^{j-1})_i (\mathbf{W}\mathbf{s}^j)_i = 2(\mathbf{W}\mathbf{s}^j - \mathbf{b})_i > 0,$$

while if $s_i^{j+1} = -1$ then $(\mathbf{W}\mathbf{s}^j)_i < b_i$ and so, again,

$$(\mathbf{s}^{j+1} - \mathbf{s}^{j-1})_i (\mathbf{W} \mathbf{s}^j)_i = -2(\mathbf{W} \mathbf{s}^j - \mathbf{b})_i > 0.$$

Summing over i we find that if $\mathbf{s}^{j+1} \neq \mathbf{s}^{j-1}$ then $\Delta E < 0$. Hence, if there exists a period greater than 2, i.e., if $\mathbf{s}^{j-1} = \mathbf{s}^{j+1+k} \neq \mathbf{s}^{j+1}$ for some k > 0 then E(j) = E(j+2+k), is contradiction of E(j+2+k) < E(j).

Exercise 3. If we integrate Eq. (27.2) on a small interval, say $(P - \varepsilon, P + \varepsilon)$, about the first input spike we find

$$\tau_E(g_{E,1}(P+\varepsilon) - g_{E,1}(P-\varepsilon)) = -\int_{P-\varepsilon}^{P+\varepsilon} g_{E,1}(t) dt + w_{inp}.$$
 (1)

Prior to the first spike we expect $g_{E,1}$ to remain at its initial value and so $g_{E,1}(P-\varepsilon))=0$ regardless of ε . Now, as $\varepsilon \to 0$ in Eq. (1) the integral vanishes and we arrive at $\tau_E g_{E,1}(P^+)=w_{inp}$ as claimed. This provides the initial data for $g_{E,1}$ from t=P up to (but not including) t=2P. In that interval $g_{E,1}$ obeys $\tau_E g'_{E,1}(t)=-g_{E,1}(t)$ and so, proceeding as in §3.1 we find

$$g_{E,1}(t) = \exp((P - t)/\tau_E)w_{inp}/\tau_E \quad P \le t < 2P$$
 (2)

and so $g_{E,1}(2P^-) = \exp(-P/\tau_E)w_{inp}/\tau_E$. If we next integrate Eq. (27.2) across t = 2P we find $\tau_E(g_{E,1}(2P^+) - g_{E,1}(2P^-)) = w_{inp}$, and so conclude that $g_{E,1}(2P^+) = g_{E,1}(2P^-) + w_{inp}/\tau_E = (1 + \exp(-P/\tau_E))w_{inp}/\tau_E$ as claimed. Arguing as above we find

$$g_{E,1}(t) = g_{E,1}(2P^+) \exp((2P - t)/\tau_E)$$

$$= \exp((P - t)/\tau_E)(1 + \exp(P/\tau_E))w_{inp}/\tau_E, \quad 2P \le t < 3P$$
(3)

as claimed. Upon comparing Eqs. (2) and (3) we deduce that

$$g_{E,1}(t) = \exp((P-t)/\tau_E)(w_{inp}/\tau_E) \sum_{m=0}^{n-2} \exp(P/\tau_E)^m, \quad (n-1)P \le t < nP.$$
(4)

We recognize this sum as a finite geometric series and find

$$\sum_{m=0}^{n-2} \exp(P/\tau_E)^m = \frac{1 - \exp((n-1)P/\tau_E)}{1 - \exp(P/\tau_E)}.$$

On inserting this back into Eq. (4) we arrive at the desired Eq. (27.41).

Exercise 8. If we define $V(\theta) \equiv W(\theta) \star f(\theta) - U(\theta)$ then the convolution theorem and linearity yield $\hat{V}_n = \hat{W}_n \hat{f}_n - \hat{U}_n$. This, together with Parseval's equality yields

$$E(W) = \sum_{n = -\infty}^{\infty} |\hat{V}_n|^2 + \lambda |\hat{W}_n|^2 = \sum_{n = -\infty}^{\infty} |\hat{W}_n \hat{f}_n - \hat{U}_n|^2 + \lambda |\hat{W}_n|^2,$$
 (5)

as claimed. Now, for n fixed, we choose \hat{W}_n as the minimizer of

$$E_n(W) \equiv |W\hat{f}_n - \hat{U}_n|^2 + \lambda |W|^2.$$

We write W = a + ib and expand the squared magnitudes and find

$$E_n(W) = E_n(a,b) = (|\hat{f}_n|^2 + \lambda)(a^2 + b^2) - (a+ib)\hat{U}_n^* \hat{f}_n - (a-ib)\hat{U}_n \hat{f}_n^*.$$

It remains to determine its critical point. We compute

$$\partial_a E_n = 2a(|\hat{f}_n|^2 + \lambda) - (\hat{U}_n^* \hat{f}_n + \hat{U}_n \hat{f}_n^*)$$
 and $\partial_b E_n = 2b(|\hat{f}_n|^2 + \lambda) - i(\hat{U}_n^* \hat{f}_n - \hat{U}_n \hat{f}_n^*).$

These vanish when

$$a = \frac{\hat{U}_n^* \hat{f}_n + \hat{U}_n \hat{f}_n^*}{2(|\hat{f}_n|^2 + \lambda)} = \frac{\Re(\hat{f}_n^* \hat{U}_n)}{|\hat{f}_n|^2 + \lambda} \quad \text{and} \quad b = \frac{\hat{U}_n^* \hat{f}_n - \hat{U}_n \hat{f}_n^*}{2(|\hat{f}_n|^2 + \lambda)} = \frac{\Im(\hat{f}_n^* \hat{U}_n)}{|\hat{f}_n|^2 + \lambda}.$$

On combining these as W = a + ib we recover Eq. (27.33).

Exercise 10. (i) We proceed from the definition

$$\hat{f}_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{-inx} dx = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(-x) e^{-inx} dx = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{inx} dx = \hat{f}_{-n}.$$

- (ii) We note that $U(-\theta) = \sigma^{-1}(f(-\theta)) = \sigma^{-1}(f(\theta)) = U(\theta)$.
- (iii) The result follows from $\hat{f}_n^* = \hat{f}_{-n}$.
- (iv) We compute

$$W'(\theta) = \lim_{h \to 0} \frac{W(\theta + h) - W(\theta)}{h}$$

$$= \lim_{h \to 0} \frac{W(-\theta - h) - W(-\theta)}{h}$$

$$= -\lim_{h \to 0} \frac{W(-\theta - h) - W(-\theta)}{-h} = -W'(-\theta).$$