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## PDF and firing rate of a LIF neuron

Reference: Brunel & Hakim (1999)

The membrane potential V(t) of a LIF neuron is governed by

$$\tau_m \dot{V}(t) = -V(t) + RI(t), \tag{1}$$

where the input synaptic current

$$RI(t) = \tau_m J_E \sum_j \delta(t - t_j) - \tau_m J_I \sum_k \delta(t - t_k).$$
 (2)

 $au_m = RC$  is the membrane time constant. R and C are the membrane resistance and capacitance, respectively.  $J_E\left(J_I\right)$  is the amplitude of an excitatory (inhibitory) post-synaptic potential, whereas  $t_j$  ( $t_k$ ) represents the time of the jth (kth) excitatory (inhibitory) input spike. When  $V(t) = \theta$ , V(t) is reset to  $V_r$  and a pause for synaptic integration  $\tau_r$  is imposed to mimic the refractory period. In the high input regime, the sum of synaptic inputs to a neuron can be approximated by a fluctuating input noise:

$$RI(t) \equiv \tau_m[\mu + \sigma\eta(t)],$$
 (3)

where

$$\mu = J_E \nu_E - J_I \nu_I, \tag{4}$$

$$\sigma = \sqrt{J_E^2 \nu_E + J_I^2 \nu_I}. \tag{5}$$

 $\eta(t)$  is a white noise random process such that  $\langle \eta(t)\eta(t')\rangle = \delta(t-t')$ .  $\nu_E$  ( $\nu_I$ ) is the firing rate of the excitatory (inhibitory) input.

Normally  $\mu$  and  $\sigma$  are constant but here we assume they are functions of V. After rescaling, we have

$$\dot{V}(t) = -\frac{V(t)}{\tau_m} + \mu(V) + \sigma(V)\eta(t) \tag{6}$$

and the associated Fokker-Planck equation is

$$\frac{\partial P(V,t)}{\partial t} = \frac{\partial}{\partial V} \left(\frac{V}{\tau_m} - \mu(V)\right) P(V,t) + \frac{\sigma(V)^2}{2} \frac{\partial^2 P(V,t)}{\partial V^2}.$$
 (7)

The continuity equation gives

$$\frac{\partial P(V,t)}{\partial t} = -\frac{\partial \Phi(V,t)}{\partial V},\tag{8}$$

in which  $\Phi(V,t)$  denotes the flux,

$$\Phi(V,t) = -(\frac{V}{\tau_m} - \mu(V))P(V,t) - \frac{\sigma(V)^2}{2} \frac{dP(V,t)}{dV}.$$
 (9)

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Consider the boundary conditions (BC). First, an absorbing BC at the threshold  $V = \theta$ ,

$$P(\theta, t) = 0, (10)$$

and thus using Eq. (9),

$$\frac{\partial P(\theta, t)}{\partial V} = -\frac{2\nu(t)}{\sigma(\theta)^2}.$$
(11)

 $u(t) = \Phi(\theta, t)$  is the output rate of the neuron. At the reset  $V_r$ ,

$$\frac{\partial P(V_r^+, t)}{\partial V} - \frac{\partial P(V_r^-, t)}{\partial V} = -\frac{2\nu(t)}{\sigma(\theta)^2}.$$
 (12)

The normalization condition gives

$$\int_{-\infty}^{\theta} P(V, t)dV = 1. \tag{13}$$

In addition, P(V, t) is continuous and satisfies

$$\lim_{V \to -\infty} P(V, t) = 0, \tag{14}$$

and

$$\lim_{V \to -\infty} VP(V, t) = 0. \tag{15}$$

For steady state,

$$\frac{\partial P(V,t)}{\partial t} = 0 \tag{16}$$

$$\frac{d}{dV}\left[\left(\frac{V}{\tau_m} - \mu(V)\right)P(V) + \frac{\sigma(V)^2}{2}\frac{dP(V)}{dV}\right] = 0.$$
(17)

There is no net flux for  $V < V_r$  in steady state. Net flux flows from  $\theta$  to  $V_r$  and to  $\theta$  again.

$$\frac{d}{dV} \left[ P(V) \exp\left(2 \int^V \frac{V - \mu(V) \tau_m}{\sigma(V)^2 \tau_m} dV \right) \right] = -\frac{2\nu \Theta(V - V_r)}{\sigma(V)^2} \exp\left(2 \int^V \frac{V - \mu(V) \tau_m}{\sigma(V)^2 \tau_m} dV \right), \tag{18}$$

which is defined for  $\sigma(V) \neq 0$ . Integrate both sides from V to  $\theta$ :

$$P(V) \exp\left(2\int^{V} \frac{V' - \mu(V')\tau_{m}}{\sigma(V')^{2}\tau_{m}} dV'\right) = \int_{V}^{\theta} \frac{2\nu\Theta(V' - V_{r})}{\sigma(V')^{2}} \exp\left(2\int^{V'} \frac{V'' - \mu(V'')\tau_{m}}{\sigma(V'')^{2}\tau_{m}} dV''\right) dV'(19)$$

$$P(V) = \exp\left(-2\int^{V} \frac{V' - \mu(V')\tau_{m}}{\sigma(V')^{2}\tau_{m}} dV'\right) \int_{V}^{\theta} \frac{2\nu\Theta(V' - V_{r})}{\sigma(V')^{2}} \exp\left(2\int^{V'} \frac{V'' - \mu(V'')\tau_{m}}{\sigma(V'')^{2}\tau_{m}} dV''\right) dV'(20)$$

If  $\mu$  and  $\sigma$  do not depend on V,

$$P(V) = \frac{2\nu}{\sigma^2} \exp\left(-\frac{(V - \mu \tau_m)^2}{\sigma^2 \tau_m}\right) \int_V^\theta \Theta(V' - V_r) \exp\left(\frac{(V' - \mu \tau_m)^2}{\sigma^2 \tau_m}\right) dV'. \tag{21}$$

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To obtain the output rate, we make use of the normalization condition Eq. (13) and we have

$$\frac{1}{\nu} = \frac{2}{\sigma^2} \int_{-\infty}^{\theta} dV \exp\left(-\frac{(V - \mu \tau_m)^2}{\sigma^2 \tau_m}\right) \int_{V}^{\theta} \Theta(V' - V_r) \exp\left(\frac{(V' - \mu \tau_m)^2}{\sigma^2 \tau_m}\right) dV' \quad (22)$$

$$= 2\tau_m \int_{-\infty}^{\frac{\theta - \mu \tau_m}{\sigma \sqrt{\tau_m}}} dv \int_{v}^{\frac{\theta - \mu \tau_m}{\sigma \sqrt{\tau_m}}} du \Theta\left(u - \frac{V_r - \mu \tau_m}{\sigma \sqrt{\tau_m}}\right) e^{u^2 - v^2}$$
(23)

$$= 2\tau_m \int_{\frac{V_r - \mu \tau_m}{\sigma \sqrt{\tau_m}}}^{\frac{\theta - \mu \tau_m}{\sigma \sqrt{\tau_m}}} du e^{u^2} \int_{-\infty}^{u} dv e^{-v^2}$$
(24)

Since the integrand is symmetric about v = 0,

$$\frac{1}{\nu} = 2\tau_m \int_{y_r}^{y_{\theta}} du e^{u^2} \int_0^{\infty} dv e^{-(v-u)^2}$$
 (25)

$$= \tau_m \int_0^\infty dv e^{-v^2} \left[ \frac{e^{2y_\theta v} - e^{2y_r v}}{v} \right], \tag{26}$$

where  $y_{\theta} = \frac{\theta - \mu \tau_m}{\sigma \sqrt{\tau_m}}$ .