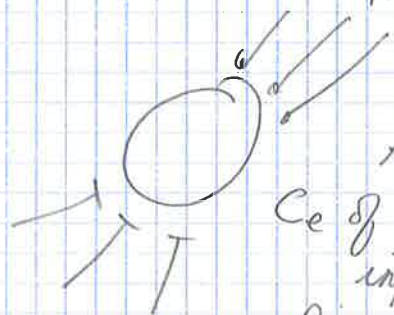


Notes for the tutorial on Meanfield Models for Networks of Spiking Neurons, ICMNS, May 29th 2016.
Given by Alex Roxin.

May 29th 2016

A. The Diffusion Approximation



A single cell receiving a large number C_e of excitatory and C_i of inhibitory inputs. The inputs are independent Poisson processes with rates λ_e and λ_i , respectively.

We will consider a leaky integrate-and-fire neuron (LIF)

$$\tau_m \dot{V} = -V + I(t) \quad (1)$$

$$I(t) = J_e \sum_{i=1}^{C_e} \sum_j \delta(t - t_j^i) - J_i \sum_{i=1}^{C_i} \sum_j \delta(t - t_j^i)$$

There is a slight difference in notation compared to the chapter by Renart et al which is following.

→ clin taking all synapses to have the same efficacy.

The j th spike of input i happens at a time t_j^i and causes a jump of amplitude J_e (or $-J_i$) in the voltage.

The dynamics of V are stochastic, so let's consider the evolution of the probability distribution for V , $p(V, t)$.

$$p(V, t + \Delta t | V_0, t_0) = \int_{-\infty}^{\infty} \underbrace{-V' p(V, t + \Delta t | V', t) p(V', t | V_0, t_0)}_{\text{sum of all the ways to get from } (V_0, t_0) \text{ to } (V, t + \Delta t) \text{ for a Markov process}}$$

sum of all the ways to get from (V_0, t_0) to $(V, t + \Delta t)$ for a Markov process (1)

How can you go from (V', t) to $(V, t + \Delta t)$ when Δt is very small ($\Delta t \ll \tau_m$)?

We assume it is so small that either there is 1 exc. input, 1 inh. input, or no inputs at all.

$$\text{So, } p(V, t + \Delta t | V', t) = P_0 \cdot \delta(V - V_0') + P_+ \cdot \delta(V - V_+') + P_- \cdot \delta(V - V_-')$$

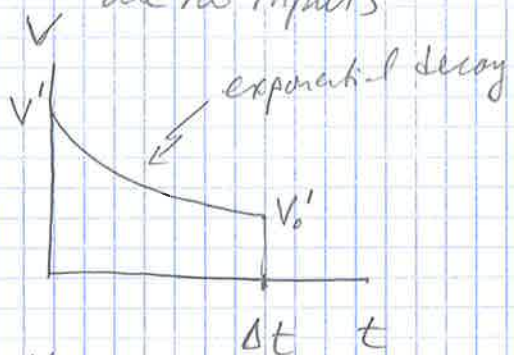
where

$$P_0 = P_r(\text{no input}) = 1 - [C_e V_e + C_i V_i] \Delta t$$

$$P_+ = P_r(1 \text{ exc. input}) = C_e V_e \Delta t$$

$$P_- = P_r(1 \text{ inh. input}) = C_i V_i \Delta t$$

$V_0' \equiv$ voltage after Δt when you start at V' and there are no inputs

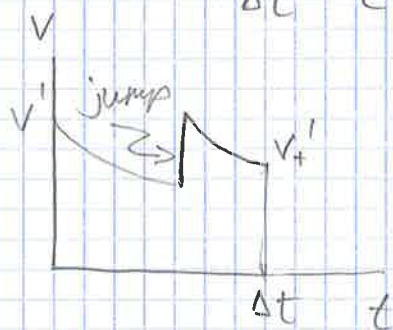


Using Eg. 1

$$V_0' = V' e^{-\Delta t / \tau_m}$$

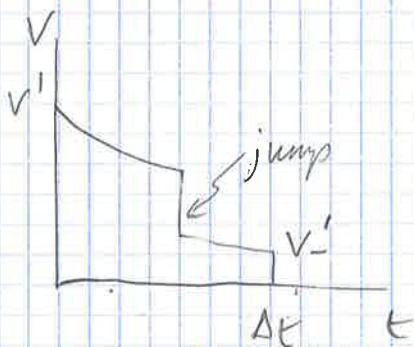
small

$$\approx V' \left(1 - \frac{\Delta t}{\tau_m} \right)$$



$$V_+' = V' \left(1 - \frac{\Delta t}{\tau_m} \right) + J_e + o(\Delta t)$$

there's a part here which is the decay of the jump J_e , but we don't need it



$$V_-' = V' \left(1 - \frac{\Delta t}{\tau_m} \right) - J_i + o(\Delta t)$$

then,

$$\int_{-\infty}^{\infty} dV' p(V, t+\Delta t | V', t) p(V', t | V_0, t_0)$$

$$= P_0 \int_{-\infty}^{\infty} dV' \delta(V - V' (1 - \frac{\Delta t}{\tau_m})) p(V', t) \quad \text{lim leaving off the } V_0, t_0 \text{ here.}$$

$$+ P_+ \int_{-\infty}^{\infty} dV' \delta(V - V' (1 - \frac{\Delta t}{\tau_m}) - J_e) p(V', t)$$

$$+ P_- \int_{-\infty}^{\infty} dV' \delta(V - V' (1 - \frac{\Delta t}{\tau_m}) + J_i) p(V', t)$$

for (A) let $x = V' (1 - \frac{\Delta t}{\tau_m})$, $dx = dV' (1 - \frac{\Delta t}{\tau_m})$

$$V' = x \left(1 + \frac{\Delta t}{\tau_m} \right) \Rightarrow dV' = dx \left(1 + \frac{\Delta t}{\tau_m} \right) + O(\Delta t^2)$$

for (B) let $x = V' (1 - \frac{\Delta t}{\tau_m}) - J_e$

for (C) let $x = V' (1 - \frac{\Delta t}{\tau_m}) + J_i$

$$\Rightarrow p(V, t+\Delta t) = P_0 \left(1 + \frac{\Delta t}{\tau_m} \right) \int_{-\infty}^{\infty} dx \delta(V - x) p\left(x \left(1 + \frac{\Delta t}{\tau_m} \right), t\right)$$

$$+ P_+ \left(1 + \frac{\Delta t}{\tau_m} \right) \int_{-\infty}^{\infty} dx \delta(V - x) p\left(x \left(1 + \frac{\Delta t}{\tau_m} \right) - J_e, t\right)$$

$$+ P_- \left(1 + \frac{\Delta t}{\tau_m} \right) \int_{-\infty}^{\infty} dx \delta(V - x) p\left(x \left(1 + \frac{\Delta t}{\tau_m} \right) + J_i, t\right)$$

We won't need these terms since $P_+, P_- \sim O(\Delta t)$ already and we will only keep terms of $O(\Delta t)$

$$p(v, t + \Delta t) = \left(1 - [C_e v_e + C_i v_i] \Delta t\right) p\left(v\left(1 + \frac{\Delta t}{\tau_m}\right), t\right) \cdot \left(1 + \frac{\Delta t}{\tau_m}\right) \\ + C_e v_e \Delta t p(v - J_e, t) + C_i v_i \Delta t p(v + J_i, t)$$

expand in Taylor Series for small Δt

$$p(v, t) + \frac{\partial p(v, t)}{\partial t} \Delta t = \left(1 - [C_e v_e + C_i v_i] \Delta t\right) \left(1 + \frac{\Delta t}{\tau_m}\right) \left(p(v, t) + \frac{\partial p(v, t)}{\partial v} \cdot \frac{v \Delta t}{\tau_m} + C_e v_e \Delta t p(v - J_e, t) + C_i v_i \Delta t p(v + J_i, t) + O(\Delta t^2)\right)$$

$$\cancel{p(v, t)} + \frac{\partial p(v, t)}{\partial t} \Delta t = \cancel{p(v, t)} + \frac{\partial p(v, t)}{\partial v} \cdot \frac{v \Delta t}{\tau_m} + \frac{\Delta t}{\tau_m} p(v, t) + C_e v_e \Delta t [p(v - J_e, t) - p(v, t)] + C_i v_i \Delta t [p(v + J_i, t) - p(v, t)] + O(\Delta t^2)$$

$$\frac{\partial p}{\partial t} = \frac{\partial}{\partial v} \left(\frac{v}{\tau_m} p \right) + C_e v_e (p(v - J_e, t) - p(v, t)) + C_i v_i (p(v + J_i, t) - p(v, t)) + O(\Delta t)$$

(2)

Then take $\Delta t \rightarrow 0$

If we assume $J_e \ll 1$ we can Taylor expand $p(v - J_e, t)$

$$p(v - J_e, t) = p(v, t) - J_e \frac{\partial p(v, t)}{\partial v} + \frac{J_e^2}{2} \frac{\partial^2 p(v, t)}{\partial v^2} + O(J_e^3)$$

If we truncate at second order we can write Eq. 2 as

$$\frac{\partial p}{\partial t} = \frac{\sigma^2}{2\tau_m} \frac{\partial^2 p}{\partial v^2} + \frac{\partial}{\partial v} \left(\left(\frac{v - \mu}{\tau_m} \right) p \right) \quad (3)$$

$$\sigma^2 = \tau_m [J_e^2 C_e v_e + J_i^2 C_i v_i]$$

$$\mu = \tau_m [J_e C_e v_e - J_i C_i v_i]$$

Eq. 3 is the Fokker-Planck equation. It arises through the diffusion approximation, i.e. $J_e, J_i \ll 1$ but $J_e C_e v_e$ and $J_i C_i v_i \sim O(1)$

B. Steady-state Solution to Fokker-Planck Equation

$$\frac{\partial \rho}{\partial t} = \frac{\omega^2}{2\tau} \frac{\partial^2 \rho}{\partial v^2} + \frac{\partial}{\partial v} \left(\left(\frac{v-\mu}{\tau} \right) \rho \right) \quad (4)$$

can also be written as

$$\frac{\partial p}{\partial t} + \frac{\partial F}{\partial V} = 0$$

$$F_z = \frac{\sigma^2}{2\pi} \frac{\partial \rho}{\partial v} - \left(\frac{v - \mu}{\tau} \right) \rho$$

Boundary Conditions:

- $p(V_t, t) = 0$ (p is zero for all voltages $> V_t$ and so must be zero at V_t , otherwise the gradient $\frac{\partial p}{\partial V}$ would be infinite \Rightarrow inf. flux \rightarrow inf. firing rate)

$$\bullet V(t) = F(v_t, t) = -\frac{\sigma^2}{2\tau} \frac{\partial^2 p(v_t, t)}{\partial v^2} - \frac{(v_t - \mu)}{\tau} p(v_t, t)$$

$$\rightarrow \frac{\partial p(v_t, t)}{\partial v} = -\frac{2\tau V(t)}{\sigma^2} \quad (\text{the flux at threshold is just the firing rate})$$

But the flux at threshold gets reinjected at the reset potential, so there is a jump in the flux there, i.e.

$$\lim F(V_r + \epsilon, t) - F(V_r - \epsilon, t) = \nu(t)$$

$$\epsilon \rightarrow 0 \quad \frac{-\omega^2}{2\tau} \frac{\partial p}{\partial V}(V_{r+\epsilon}, t) - (V_{r+\epsilon} - \mu) \frac{p(V_{r+\epsilon}, t)}{\tau}$$

$$+\frac{\sigma^2}{2\tau} \frac{\partial p}{\partial v}(v_r - \epsilon, t) + \frac{(v_r - \epsilon - \mu)}{\tau} p(v_r - \epsilon, t)$$

$$\rightarrow \lim_{\epsilon \rightarrow 0} \left[\frac{\partial p(V_r + \epsilon, t)}{\partial V} - \frac{\partial p(V_r - \epsilon, t)}{\partial V} \right] = - \frac{2\tau V(t)}{\sigma^2}$$

$$\bullet \int_{-\infty}^{V_t} dV p(V, t) = 1 \quad (p \text{ is a probability distribution or prob. density})$$

if there is a refractory time τ_{ref} then this condition changes to

$$\int_{-\infty}^{V_t} dV p(V, t) + \tau_{ref} V(t) = 1$$

• We also require that $p \rightarrow 0$ as $V \rightarrow \infty$ but this will be fulfilled trivially

Now we solve for the steady-state, i.e. $p(V, t) = p_0(V)$
 $V(t) = V_0$

From Eq. 4

$$\frac{\partial p_0}{\partial V} + 2 \frac{(V - \mu)}{\sigma^2} p_0 = G'$$

What is G' ? Since the flux has a jump at V_r , G' takes on one value for $V < V_r$ and another for $V \geq V_r$

$$\frac{\partial p_0(V_t)}{\partial V} + 2 \frac{(V_t - \mu)}{\sigma^2} p_0(V_t) = G' \quad \text{at } V_t = V_r$$

$$\Rightarrow G' = - \frac{2\tau V_0}{\sigma^2} \quad V \geq V_r$$

this implies $G' = 0 \quad V < V_r$

So we can write

$$\frac{\partial p_0}{\partial V} + 2 \frac{(V-\mu)}{\sigma^2} p_0 = -\frac{2\tau V_0}{\sigma^2} H(V-V_r),$$

$$\text{where } H(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$

Integrating,

$$\frac{\partial}{\partial V} \left(p_0 e^{\frac{(V-\mu)^2}{\sigma^2}} \right) = -\frac{2\tau V_0}{\sigma^2} e^{\frac{(V-\mu)^2}{\sigma^2}} H(V-V_r)$$

$$p_0 e^{\frac{(V-\mu)^2}{\sigma^2}} = -\frac{2\tau V_0}{\sigma^2} \int_V^{V_t} dx e^{\frac{(x-\mu)^2}{\sigma^2}} H(x-V_r)$$

$$p_0(V) = +\frac{2\tau V_0}{\sigma^2} e^{-\frac{(V-\mu)^2}{\sigma^2}} \int_V^{V_t} dx e^{\frac{(x-\mu)^2}{\sigma^2}} H(x-V_r)$$

$$\text{let } y = \frac{x-\mu}{\sigma} \quad dx = \sigma dy$$

$$p_0(V) = \frac{2\tau V_0}{\sigma} e^{-\frac{(V-\mu)^2}{\sigma^2}} \int_{\frac{V-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} dy e^{y^2} H\left(y - \left(\frac{V_r-\mu}{\sigma}\right)\right)$$

To find the firing rate we use the condition

$$\int_{-\infty}^{V_t} dV p_0(V) = 1$$

$$\frac{2\tau V_0}{\sigma} \int_{-\infty}^{V_t} dV e^{-\frac{(V-\mu)^2}{\sigma^2}} \int_{\frac{V-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} dy e^{y^2} H\left(y - \left(\frac{V_r-\mu}{\sigma}\right)\right)$$

$$\text{let } u = \frac{v - \mu}{\sigma} \quad dv = \sigma du$$

$$\Rightarrow 2\tau V_0 \int_{-\infty}^{\frac{v_t - \mu}{\sigma}} du e^{-u^2} \int_u^{\frac{v_t - \mu}{\sigma}} dy e^{y^2} H\left(y - \frac{(v_r - \mu)}{\sigma}\right)$$

$$-\infty < u \leq \frac{v_t - \mu}{\sigma}$$

$$u < y \leq \frac{v_t - \mu}{\sigma}$$

inverting order of integration

$$2\tau V_0 \int_{-\infty}^{\frac{v_t - \mu}{\sigma}} dy e^{y^2} H\left(y - \frac{(v_r - \mu)}{\sigma}\right) \int_{-\infty}^y du e^{-u^2}$$

$$= 2\tau V_0 \int_{\frac{v_r - \mu}{\sigma}}^{\frac{v_t - \mu}{\sigma}} dy e^{y^2} \cdot \frac{\sqrt{\pi}}{2} \operatorname{erfc}(-y)$$

check this with Wikipedia!

$$\Rightarrow \tau V_0 \sqrt{\pi} \int_{\frac{v_r - \mu}{\sigma}}^{\frac{v_t - \mu}{\sigma}} dy e^{y^2} \operatorname{erfc}(-y) = 1$$

$$V_0 = \frac{1}{\tau \sqrt{\pi}} \left[\int_{\frac{v_r - \mu}{\sigma}}^{\frac{v_t - \mu}{\sigma}} dy e^{y^2} \operatorname{erfc}(-y) \right]^{-1} \quad (5)$$

C. Here we calculate the CV of the ISI using the recurrence relation Eq. 15.27 from Renaut et al.

First note that I will call $\langle T_{fp}^k \rangle$ (kth moment of mean first passage time)
Just μ_k

So

$$\frac{\sigma^2}{2} \frac{d^2 \mu_1}{dx^2} - (x - \mu) \frac{d\mu_1}{dx} = -1 \quad (\text{eq. 15.27 in Renaut et al.})$$

let $z = \frac{d\mu_1}{dx}$

$$\frac{dz}{dx} - \frac{2(x-\mu)}{\sigma^2} z = -\frac{2}{\sigma^2}$$

$$\frac{d}{dx} \left(z e^{-\frac{(x-\mu)^2}{\sigma^2}} \right) = -\frac{2}{\sigma^2} e^{-\frac{(x-\mu)^2}{\sigma^2}}$$

$$z e^{-\frac{(x-\mu)^2}{\sigma^2}} = -\frac{2}{\sigma^2} \int dy e^{-\frac{(y-\mu)^2}{\sigma^2}}$$

$$z = -\frac{2}{\sigma^2} e^{\frac{(x-\mu)^2}{\sigma^2}} \int_{-\infty}^x dy e^{-\frac{(y-\mu)^2}{\sigma^2}}$$

$$\mu_1 = -\frac{2}{\sigma^2} \int_{V_t}^x du e^{\frac{(u-\mu)^2}{\sigma^2}} \int_{-\infty}^u dy e^{-\frac{(y-\mu)^2}{\sigma^2}}$$

$$= \frac{2}{\sigma^2} \int_{V_r}^{V_t} du e^{\frac{(u-\mu)^2}{\sigma^2}} \int_{-\infty}^u dy e^{-\frac{(y-\mu)^2}{\sigma^2}}$$

let $w = \frac{y-\mu}{\sigma}$

$$v = \frac{u-\mu}{\sigma}$$

$$= 2 \int_{\frac{V_r-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} dv e^{v^2} \underbrace{\int_{-\infty}^v dw e^{-w^2}}_{\frac{\sqrt{\pi}}{2} \text{erfc}(-v)}$$

$$= \sqrt{\pi} \int_{\frac{V_r-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} dv e^{v^2} \text{erfc}(-v)$$

So we have $\mu_1 = \frac{1}{V_0 \tau}$

Now

$$\frac{\sigma^2}{2} \frac{d^2 \mu_2}{dx^2} - (x - \mu) \frac{d\mu_2}{dx} = -2\mu_1(x)$$

let $z = \frac{d\mu_2}{dx}$

$$\frac{dz}{dx} - \frac{2}{\sigma^2} (x - \mu) z = -\frac{4}{\sigma^2} \mu_1(x)$$

$$\frac{d}{dx} \left(z e^{-\frac{(x-\mu)^2}{\sigma^2}} \right) = -\frac{4}{\sigma^2} e^{-\frac{(x-\mu)^2}{\sigma^2}} \mu_1(x)$$

$$z e^{-\frac{(x-\mu)^2}{\sigma^2}} = -\frac{4}{\sigma^2} \int_{-\infty}^x dy e^{-\frac{(y-\mu)^2}{\sigma^2}} \mu_1(y)$$

$$z = -\frac{4}{\sigma^2} e^{\frac{(x-\mu)^2}{\sigma^2}} \int_{-\infty}^x dy e^{-\frac{(y-\mu)^2}{\sigma^2}} \mu_1(y)$$

$$\mu_2 = -\frac{4}{\sigma^2} \int_{V_t}^x dx' e^{\frac{(x'-\mu)^2}{\sigma^2}} \int_{-\infty}^{x'} dy e^{-\frac{(y-\mu)^2}{\sigma^2}} \mu_1(y)$$

$$= \frac{4}{\sigma^2} \int_x^{V_t} dx' e^{\frac{(x'-\mu)^2}{\sigma^2}} \int_{-\infty}^{x'} dy e^{-\frac{(y-\mu)^2}{\sigma^2}} \cdot \sqrt{\pi} \int_{\frac{y-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} dv e^{v^2} \operatorname{erfc}(-v)$$

let $w = \frac{y-\mu}{\sigma}$

$u = \frac{x'-\mu}{\sigma}$

$$\mu_2 = 4\sqrt{\pi} \int_{\frac{V_t-\mu}{\sigma}}^{\frac{V_t-\mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_w^{\frac{V_t-\mu}{\sigma}} dv e^{v^2} \operatorname{erfc}(-v)$$

Note $\int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} = \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} + \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}}$

$$\Rightarrow \mu_2 = 4\pi \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} dv e^{v^2} \text{erfc}(-v)$$

$$+ 4\pi \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} dv e^{v^2} \text{erfc}(-v) \quad \begin{matrix} -\infty < v \leq u \\ -\infty < w < v \leq u \end{matrix}$$

$$= 2\pi \left[\int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \text{erf}(-u) \right]^2 + 4\pi \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} dv e^{v^2} \text{erfc}(-v) \int_{-\infty}^v dw e^{-w^2}$$

+ factor 2 should not be here

$$= (2\mu_1^2 + 2\pi \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} dv e^{v^2} \text{erfc}(-v))^2$$

$$CV = \frac{\mu_2 - \mu_1^2}{\mu_1^2} = \left(2\pi \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} du e^{u^2} \int_{-\infty}^u dw e^{-w^2} \int_{\frac{V_r - \mu}{\sigma}}^{\frac{V_t - \mu}{\sigma}} dv e^{v^2} \text{erfc}(-v) \right)^2$$

(6)

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D. All-to-all network of LIF neurons

$$\tau_E \frac{dV_i^E}{dt} = -V_i^E + \mu_{ext,E} + \tau \left(\frac{J_E}{N} \sum_{j=1}^{N_E} \sum_k \delta(t-t_j^k - 0) - \frac{J_i}{N} \sum_{j=1}^{N_i} \sum_k \delta(t-t_j^k - 0) \right) + \sigma \xi_i(t)$$

and an analogous equation for inh. neurons

For the sims I chose $\tau_E = \tau_I = 20 \text{ ms}$

$$V_t = 20 \text{ mV}, V_r = 10 \text{ mV}$$

$$J_E = J_i = 30, J_{ext,E} = 14.7 \text{ mV}$$

$\sigma = 5 \text{ mV}$, $\xi_i(t)$ is a Gaussian white noise process with zero mean and unit variance. So there is no diffusion approximation here.

E. Sparse Network of LIF neurons

The difference with the all-to-all network is twofold.

1- Recurrent inputs

The recurrent input to an excitatory neuron i is

$$I_i^E = \tau J_E \sum_{j=1}^{N_E} W_{ij}^{EE} \sum_k \delta(t-t_j^k - 0) - \tau J_i \sum_{j=1}^{N_i} W_{ij}^{EI} \sum_k \delta(t-t_j^k - 0)$$

\uparrow
 $W_{ij}^{EE} = 1$ with probability $P_{EE} = \frac{C_{EE}}{N_E}$
 otherwise it is zero

2- External inputs

I take C_{ext} independent Poisson processes with rate λ_{ext}^E and amplitude J_{ext} for E neurons. For I neurons it is the same but with λ_{ext}^I .

For the theory, now

$$\mu_E = \tau (J_{ext} C_{ext} \lambda_{ext}^E + J_{EE} C_{EE} \lambda_E - J_{EI} C_{EI} \lambda_I)$$

$$\sigma_E^2 = \tau (J_{ext}^2 C_{ext} \lambda_{ext}^E + J_{EE}^2 C_{EE} \lambda_E + J_{EI}^2 C_{EI} \lambda_I)$$

With analogous equations for I neurons.
 For the simulation I take $I_{ex} = I_{re} \cdot I_i$
 $J_{ii} = J_{ei} = J_i$

F. Linear Stability for All-to-all networks

$$\frac{\partial p_e}{\partial t} = \frac{\sigma^2}{2\tau} \frac{\partial^2 p_e}{\partial v^2} + \frac{\partial}{\partial v} \left[\frac{N - \mu_e}{\epsilon} p_e \right] \quad (7)$$

$$p_e(v_{\pm}, t) = 0$$

$$\frac{\partial p_e}{\partial v}(v_{\pm}, t) = -\frac{2\tau v_{\pm}(t)}{\sigma^2}$$

$$\frac{\partial p_e}{\partial v}(v_r^+, t) - \frac{\partial p_e}{\partial v}(v_r^-, t) = -\frac{2\tau v_r(t)}{\sigma^2}$$

$$\int_{-\infty}^{v_t} dv p_e(v, t) = 1$$

With an analogous equation and set of B.C.'s for the inhibitory population.

Now

$$p_e(v, t) = p_0(v) + \epsilon p_1(v) e^{\lambda t}$$

$$v_e(t) = v_{e0} + \epsilon v_{e1} e^{\lambda t}$$

$$\mu_e = \mu_{e0} + \epsilon \mu_{e1} e^{\lambda t}$$

$$\mu_e = \mu_{ext} + \tau (J_e v_e - J_i v_i)$$

$$= \mu_{ext} + \tau (J_e v_{e0} - J_i v_{i0}) + \epsilon (J_e v_{e1} - J_i v_{i1}) e^{\lambda t}$$

$$\Rightarrow \mu_{e1} = J_e v_{e1} - J_i v_{i1}$$

Plug into Eqs. 7 and collect terms of order ϵ
 (we've already calculated O(1) terms!)

$$\lambda p_1(V) = \frac{\omega^2}{2\tau} \frac{\partial^2 p_1}{\partial V^2} + \frac{\partial}{\partial V} \left[\frac{(V - \mu_{e0})}{\tau} p_1 \right] - \frac{\mu_{e1}}{\tau} \frac{\partial p_0}{\partial V} \quad (8)$$

$$p_1(V_t) = 0$$

$$\frac{\partial p_1(V_t)}{\partial V} = -\frac{2\tau V_{e1}}{\sigma^2}$$

$$\frac{\partial p_1(V_r+1)}{\partial V} - \frac{\partial p_1(V_r-1)}{\partial V} = -\frac{2\tau V_{e1}}{\sigma^2}$$

$$\int_{-\infty}^{V_t} dV p_1(V) = 0$$

You must solve Eqs. 8 to find $p_1(V)$, V_{e1} , V_{r1} etc. The solutions can be expressed in terms of Confluent Hypergeometric functions or in integral form if solved via Laplace transforms. See Brunel 2000 for more details. However all of this is a lot of work. It's much easier to solve (8) numerically!

G. Numerical Solution of Fokker-Planck Eq.

First, we rewrite Eq. 7 as

(Following Richardson PRE 2007)

$$\frac{\partial p}{\partial t} + \frac{\partial F}{\partial V} = \nu(t) (\delta(V - V_r) - \delta(V - V_t + 1))$$

$$F = -\frac{\sigma^2}{2\tau} \frac{\partial p}{\partial V} - \frac{(V - \mu)}{\tau} p$$

This takes care of jump conditions at V_t and V_r

We want to find the stationary state and its stability numerically.

$$p = p_0 + \epsilon p_1 e^{i\omega t}$$

$$V = V_0 + \epsilon V_1 e^{i\omega t}$$

$$\mu = \mu_0 + \epsilon \mu_1 e^{i\omega t}$$

$$F = F_0 + \epsilon F_1 e^{i\omega t}$$

I assume $\lambda = \lambda_r + i\omega$ so I'm right at the instability