

# Mean field limits for Hawkes processes in a diffusive regime

Xavier Erny <sup>1</sup>

with Eva Löcherbach <sup>2</sup> and Dasha Loukianova <sup>1</sup>

<sup>1</sup>Université d'Evry Val d'Essonne (LaMME)

<sup>2</sup>Université Paris 1 Panthéon-Sorbonne (SAMM)

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Example : 2 processes  $Z_1$  and  $Z_2$



$Z_1$  inhibits  $Z_2$

$Z_2$  self-excitation

# Modeling in neurosciences

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Each spike modifies the potential of the neurons

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# Stochastic Intensity

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$\lambda$  stochastic intensity of  $Z$  if :

$$\forall 0 \leq a < b, \mathbb{E}[Z([a, b]) | \mathcal{F}_a] = \mathbb{E}\left[\int_a^b \lambda(t) dt \mid \mathcal{F}_a\right]$$

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$Z^i([0, t]) =$  number of spikes of neuron  $i$  in  $[0, t]$

$X_t^{N,i} =$  potential of neuron  $i$  at time  $t$

$f_i =$  spike rate function

$h_{ji} =$  leakage function



# Hawkes processes in diffusive mean field

For each  $N \in \mathbb{N}^*$ , we consider  $(Z^{N,1}, \dots, Z^{N,N})$  :

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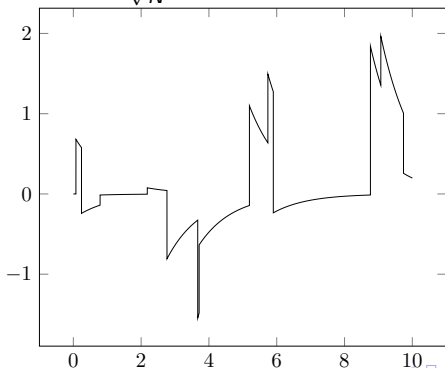
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$$\begin{cases} X_t^N = X_s^N e^{-\alpha(t-s)} & \text{if none of the } Z^{N,j} \text{ charge } [s, t] \\ X_t^N = X_{t-}^N + \frac{U_j(t)}{\sqrt{N}} & \text{if } Z^{N,j} \text{ charges } t \end{cases}$$



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A generator of  $X$  :

$$A g(x) := \frac{d}{dt} (P_t g(x)) \Big|_{t=0}$$

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$$N \rightarrow +\infty : \bar{A}g(x) = -\alpha x g'(x) + \frac{1}{2} f(x) g''(x)$$

$$d\bar{X}_t = -\alpha \bar{X}_t dt + \sqrt{f(\bar{X}_t)} dB_t$$

# Convergence of the semigroups (1)

$$\left(\bar{P}_t - P_t^N\right) g(x) = \int_0^t P_{t-s}^N \left(\bar{A} - A^N\right) \bar{P}_s g(x) ds$$

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$$\begin{aligned} \left(\bar{P}_t - P_t^N\right) g(x) &= u(t) - u(0) \\ &= \int_0^t u'(s) ds \end{aligned}$$

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$$= \int_0^t \left[ -P_{t-s}^N A^N \bar{P}_s g(x) + P_{t-s}^N \bar{A} \bar{P}_s g(x) \right] ds$$

# Convergence of the semigroups (2)

$$\left(\bar{P}_t - P_t^N\right) g(x) = \int_0^t P_{t-s}^N \left(\bar{A} - A^N\right) \bar{P}_s g(x) ds$$

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# Convergence in finite-dimensional distribution

Convergence of the semigroups :

$$\mathbb{E}_x \left[ g \left( X_t^N \right) \right] \longrightarrow \mathbb{E}_x \left[ g \left( \bar{X}_t \right) \right]$$

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Induction + classical argument of Markov theory

$\implies$  Convergence in finite-dimensional distribution :

$$\mathbb{E}_x \left[ g_1 \left( X_{t_1}^N \right) \dots g_n \left( X_{t_n}^N \right) \right] \longrightarrow \mathbb{E}_x \left[ g_1 \left( \bar{X}_{t_1} \right) \dots g_n \left( \bar{X}_{t_n} \right) \right]$$

# Convergence of the processes

- $X^N$  converges in fidi distribution to  $\bar{X}$
- $\{X^N : N \in \mathbb{N}^*\}$  tight on  $D(\mathbb{R}_+, \mathbb{R})$  (admitted)

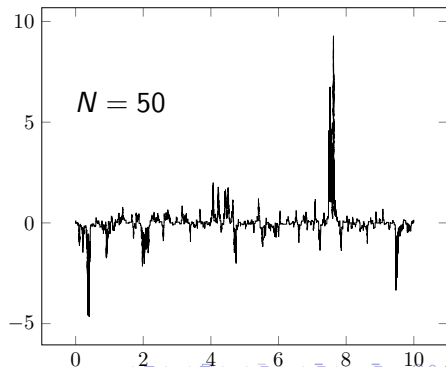
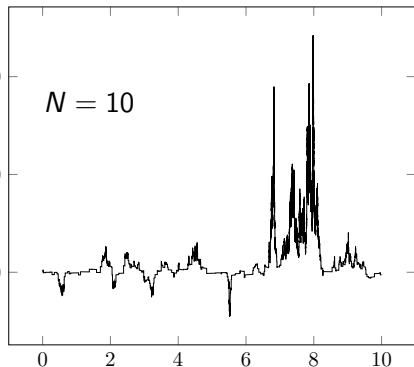


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Convergence of  $Z^{N,i}$ 

$$Z_t^{N,i} := \int_{]0,t] \times \mathbb{R}_+} \mathbf{1}_{\{z \leq f(X_{s-}^N)\}} d\pi_i^N(s, z)$$

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$\Phi$  continuous in  $(\bar{X}, \bar{\pi}_i)$  as

Skorohod's Representation Theorem :

$$\begin{array}{ccc} (X^N, \pi_i^N) & \xrightarrow{\mathcal{L}} & (\bar{X}, \bar{\pi}_i) & \quad & \Phi(X^N, \pi_i^N) & \xrightarrow{\mathcal{L}} & \Phi(\bar{X}, \bar{\pi}_i) \\ \mathcal{L} \parallel & & \parallel \mathcal{L} & \implies & \mathcal{L} \parallel & & \parallel \mathcal{L} \\ (\tilde{X}^N, \tilde{\pi}^N) & \xrightarrow{as} & (\tilde{X}, \tilde{\pi}) & & \Phi(\tilde{X}^N, \tilde{\pi}^N) & \xrightarrow{as} & \Phi(\tilde{X}, \tilde{\pi}) \end{array}$$

Result :  $(Z^{N,i})_{i \geq 1}$  converges to  $(\bar{Z}^i)_{i \geq 1}$  in distribution in  $D(\mathbb{R}_+, \mathbb{R})^{N^*}$

# Bibliography

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Thank you for your attention !

Questions ?