

Concentration phenomena for a FitzHugh-Nagumo neural network

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Challenges in the Kinetic Modelling of Complex Systems - Part I of III

Introduction

Microscopic description

FitzHugh-Nagumo neural network of size n

For i between 1 and n :

$$\left\{ \begin{array}{l} dv_t^i = \left(N(v_t^i) - w_t^i - \frac{1}{n} \sum_{j=1}^n \Phi(x_i, x_j)(v_t^i - v_t^j) \right) dt + \sqrt{2} dB_t^i, \\ dw_t^i = A(v_t^i, w_t^i) dt. \end{array} \right.$$

- v_t^i represents the tension of neuron i at time t .
- w_t^i is an adaptation variable added for modeling reasons.
- N (resp. A) is a non-linear drift (resp. an affinity) with confining property.
- Brownian motion B_t^i takes into account random fluctuation of the voltage.
- Neurons interact following Ohm's law. The conductance $\Phi(x_i, x_j)$ between neuron i and j depend on their spatial location x_i and x_j .

FitzHugh-Nagumo's mean-field equation

$$\partial_t f + \operatorname{div}_{(v, w)} \left[\begin{pmatrix} N(v) - w - \mathcal{K}_\Phi(f) \\ A(v, w) \end{pmatrix} f \right] - \partial_v^2 f = 0,$$

- $f(t, x, v, w)$ is the probability of finding neurons at time $t \geq 0$ and position $x \in K$, with potential $v \in \mathbb{R}$ and adaptation variable $w \in \mathbb{R}$.
- $\mathcal{K}_\Phi(f)$ is the **non-local term** due to **interactions** between neurons

$$\mathcal{K}_\Phi(f)(x, v) = \int_{K \times \mathbb{R}^2} \Phi(x, x')(v - v')f(x', v', w')dx' dv' dw'.$$

We decompose the interaction kernel Φ as follows

$$\Phi(x, x') = \underbrace{\Psi(x, x')}_{\text{weak-long range interactions}} + \underbrace{\frac{1}{\epsilon} \delta_0(x - x')}_{\text{strong-short range interactions}}.$$

Regime of strong interactions

Weak-Long / Strong-Short decomposition

$$\Phi(x, x') = \Psi(x, x') + \frac{1}{\epsilon} \delta_0(x - x').$$

The mean-field equation rewrites

$$\partial_t f^\epsilon + \operatorname{div}_{(v, w)} \left[\begin{pmatrix} N(v) - w - \mathcal{K}_\Psi(f^\epsilon) \\ A(v, w) \end{pmatrix} f^\epsilon \right] - \partial_v^2 f^\epsilon = \frac{\rho_0^\epsilon}{\epsilon} \partial_v [(v - \mathcal{V}^\epsilon) f^\epsilon],$$

where

$$\rho_0^\epsilon(x) = \int_{\mathbb{R}^2} f^\epsilon dv dw \quad \text{and} \quad \mathcal{V}^\epsilon(t, x) = \frac{1}{\rho_0^\epsilon} \int_{\mathbb{R}^2} v f^\epsilon dv dw.$$

Main goal

Analysis of the regime of **Strong/Local interactions**, that is when $\epsilon \rightarrow 0$.

Formal derivation

Strong interactions and concentration phenomenon

$$\partial_t f^\epsilon + \operatorname{div}_{(v,w)} \left[\begin{pmatrix} N(v) - w - \mathcal{K}_\Psi(f^\epsilon) \\ A(v,w) \end{pmatrix} f^\epsilon \right] - \partial_v^2 f^\epsilon = \frac{\rho_0^\epsilon}{\epsilon} \partial_v [(v - \mathcal{V}^\epsilon) f^\epsilon],$$

- Multiplying the equation by $|v - \mathcal{V}^\epsilon|^2$ and integrating yields

$$W_2^2 \left(\frac{1}{\rho_0^\epsilon} f^\epsilon, \frac{1}{\rho_0^\epsilon} \delta_{\mathcal{V}^\epsilon} \otimes \int_{\mathbb{R}} f^\epsilon dv \right) = \int_{\mathbb{R}^2} |v - \mathcal{V}^\epsilon|^2 f^\epsilon dv dw \underset{\epsilon \rightarrow 0}{=} O(\epsilon),$$

where W_2 is the Wasserstein distance of order 2. In the end, we obtain¹

$$f^\epsilon(t, x, v, w) \underset{\epsilon \rightarrow 0}{=} \delta_{\mathcal{V}(t,x)}(v) \otimes F(t, x, w) + O(\sqrt{\epsilon}),$$

where (\mathcal{V}, F) satisfies

$$\begin{cases} \partial_t \mathcal{V} = N(\mathcal{V}) - \mathcal{W} - (\Psi *_r \rho_0(x) \mathcal{V} - \Psi *_r (\rho_0 \mathcal{V})(x)), \\ \partial_t F + \partial_w (A(\mathcal{V}, w) F) = 0, \\ \rho_0(x) \mathcal{W} = \int_{\mathbb{R}} w F dw. \end{cases} \quad (1)$$

1. Crevat, Faye, Filbet (19), Jabin, Rey (17), Kang, Vasseur (15)

- To refine our description, we consider a re-scaled version g^ϵ of f^ϵ ² :

$$f^\epsilon(t, x, v, w) = \frac{1}{\theta^\epsilon} g^\epsilon \left(t, x, \frac{v - \mathcal{V}^\epsilon}{\theta^\epsilon}, w - \mathcal{W}^\epsilon \right),$$

where $\theta^\epsilon > 0$ needs to be determined and

$$\mathcal{W}^\epsilon = \frac{1}{\rho_0^\epsilon} \int_{\mathbb{R}^2} w f^\epsilon dv dw.$$

Main goal

proving that g^ϵ converges and compute the limit.

2. Mouhot, Mischler (06), Rey (12)

Formal derivation

Suppose $\theta^\epsilon = \epsilon^\alpha$. Changing variables in the equation of f^ϵ it yields

$$\partial_t g^\epsilon + \operatorname{div}_{(v,w)} [b_0^\epsilon g^\epsilon] = \frac{1}{\epsilon^{2\alpha}} \partial_v (\rho_0^\epsilon \epsilon^{2\alpha-1} v g^\epsilon + \partial_v g^\epsilon),$$

Therefore $\theta^\epsilon = \sqrt{\epsilon}$ (i.e. $\alpha = 1/2$) is the only suitable choice

Equation on the profile

$$\partial_t g^\epsilon + \operatorname{div}_{(v,w)} [b_0^\epsilon g^\epsilon] = \frac{1}{\epsilon} \partial_v [\rho_0^\epsilon v g^\epsilon + \partial_v g^\epsilon],$$

where b_0^ϵ depends on $1/\sqrt{\epsilon}$ and f^ϵ . Therefore, we expect

$$g^\epsilon(t, x, v, w) \underset{\epsilon \rightarrow 0}{=} \mathcal{M}_{\rho_0^\epsilon}(v) \otimes G^\epsilon(t, x, w) + O(\sqrt{\epsilon}).$$

Weak convergence result

Weak convergence result

Theorem (F. Filbet and A.B. ³)

Under suitable assumptions on f_0^ϵ , there exists $C > 0$ such that

$$\sup_{x \in K} W_2 \left(\frac{1}{\rho_0^\epsilon} f^\epsilon, \frac{1}{\rho_0} \mathcal{M}_{\rho_0/\epsilon}(\cdot - \mathcal{V}) \otimes F \right) \leq C \left(e^{Ct} \epsilon + e^{-\rho_0^\epsilon t/\epsilon} \right),$$

for all $\epsilon > 0$ and $t \geq 0$.

Here, W_2 stands for the Wasserstein distance of order 2.

Key arguments of the proof :

- Uniform moment estimates.
- Analytic **coupling method** ⁴ in order to estimate the Wasserstein distance between g^ϵ and $\mathcal{M} \otimes G$ (G satisfies (1) after changing variables) .

3. Concentration phenomena in Fitzhugh-Nagumo's equations : A mesoscopic approach, arXiv :2201.02363

4. Fournier, Perthame (20).

Strong convergence results

Towards strong convergence : time dependent θ^ϵ

- $\theta^\epsilon = \sqrt{\epsilon}$ induces that at time $t = 0$, it holds

$$f_0^\epsilon(x, v, w) = \frac{1}{\sqrt{\epsilon}} g_0^\epsilon \left(x, \frac{v - \mathcal{V}_0^\epsilon}{\sqrt{\epsilon}}, w - \mathcal{W}_0^\epsilon \right).$$

- Therefore, we impose $\theta^\epsilon(t = 0) = 1$. The only suitable choice is

$$\theta^\epsilon(t, x) = \sqrt{\epsilon} \left[1 + \underbrace{e^{-2\rho_0^\epsilon(x)t/\epsilon} (\epsilon^{-1} - 1)}_{\text{exponentially decaying remainder}} \right]^{\frac{1}{2}}.$$

The equation on g^ϵ rewrites

$$\partial_t g^\epsilon + \operatorname{div}_{(v, w)} [b_0^\epsilon g^\epsilon] = \frac{1}{|\theta^\epsilon|^2} \partial_v [\rho_0^\epsilon v g^\epsilon + \partial_v g^\epsilon],$$

Strong convergence result

Theorem (A.B.⁵)

Under suitable assumptions on f_0^ϵ , there exists $C > 0$ such that

$$\int_0^t \|f^\epsilon - f\|_{L_x^\infty L_{(v,w)}^1}(s) ds \leq C e^{Ct} \sqrt{\epsilon},$$

for all $\epsilon > 0$ and $t \geq 0$, where the limit f is given by

$$f(t, x, v, w) = \mathcal{M}_{\rho_0 |\theta^\epsilon|^{-2}}(v - \mathcal{V}) \otimes F,$$

where (\mathcal{V}, F) solves (1).

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5. Large coupling in a FitzHug-Nagumo neural network : quantitative and strong convergence results (in preparation).

Key arguments

- Relative entropy estimate yields $g^\epsilon \underset{\epsilon \rightarrow 0}{\sim} \mathcal{M}_{\rho_0^\epsilon} \otimes G^\epsilon + O(\sqrt{\epsilon})$ in L^1 .
- Proving that G^ϵ converges towards G :

(i) We work on the re-scaled version H^ϵ

$$H^\epsilon = \int_{\mathbb{R}} g^\epsilon \left(t, x, v, w - \epsilon^{3/2} v \right) dv.$$

- (ii) L^1 -equicontinuity estimates for g^ϵ yield $H^\epsilon \underset{\epsilon \rightarrow 0}{=} G^\epsilon + O(\sqrt{\epsilon})$ in L^1 .
- (iii) Then we prove $H^\epsilon \underset{\epsilon \rightarrow 0}{=} G + O(\sqrt{\epsilon})$ in L^1 .

H^ϵ converges towards G : simplified example

- We consider the diffusive scaling for Fokker-Planck equation

$$\partial_t g^\epsilon + \frac{1}{\epsilon} v \cdot \nabla_x g^\epsilon = \frac{1}{\epsilon^2} \operatorname{div}_v [v g^\epsilon + \nabla_v g^\epsilon],$$

and we prove

$$G^\epsilon = \int_{\mathbb{R}} g^\epsilon dv \xrightarrow{\epsilon \rightarrow 0} G, \quad \text{where} \quad \partial_t G = \Delta_x G.$$

- Define

$$H^\epsilon(t, x) = \int_{\mathbb{R}} g^\epsilon(t, x + \epsilon v, v) dv.$$

- H^ϵ SOLVES the limiting equation

$$\partial_t H^\epsilon = \Delta_x H^\epsilon.$$

- Therefore, it is sufficient to prove $H^\epsilon \sim G^\epsilon$ (i.e. equicontinuity for g^ϵ).

Conclusion and perspectives

Conclusion :

- Weak convergence result with the (formal) optimal convergence rate.
- L^1 convergence result with deteriorated convergence rate.
- We also prove a convergence result in (inverse Gaussian) weighted L^2 spaces and recover the optimal rate by propagating regularity.

Perspective :

- Obtaining similar results (explicit convergence rates) following a Hamilton-Jacobi approach⁶.

Thank you for your attention !

6. Quininao, Touboul (20)