

Concentration phenomena for a FitzHugh-Nagumo neural network

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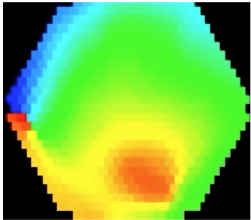
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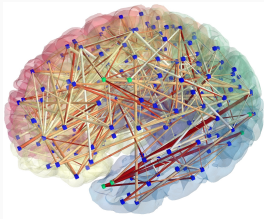
Modèles et méthodes pour les équations cinétiques

Overview

We consider that **neural networks** are **particle systems**. It is possible to describe them through **3 distinct scales** :



Macroscopic scale



Mesoscopic scale



Microscopic scale

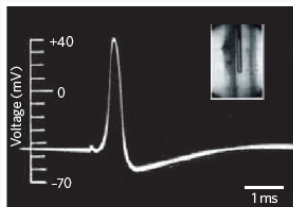
GOAL :

Derive information at the **macroscopic scale** from the **mesoscopic scale**

Microscopic scale and mean field limit

Behavior of a neuron

- We focus on the dynamics of the **voltage** through the membrane of a neuron. Experiments showcase 2 main features



(i) **Delay** when responding to an external input.

(ii) **Self-regulation**.

Hodgkin & Huxley, '52.

- **Hodgkin & Huxley** obtained a precise but mathematically complicated model.
- We will use a **simplified** version that captures its **main features** : FitzHugh-Nagumo's model for a neuron

FitzHugh-Nagumo's neuron

A neuron i is modeled by $(v_t^i, w_t^i) \in \mathbb{R}^2$ (voltage & delay)

$$\begin{cases} dv_t^i = (N(v_t^i) - w_t^i + I_{\text{ext}}) dt + \sqrt{2}dB_t, \\ dw_t^i = A(v_t^i, w_t^i) dt, \end{cases}$$

- A is an affinity while N is non-linear, typically $N(v) = v - v^3$.
- I_{ext} is the current resulting from interaction with other neurons

$$I_{\text{ext}} = -\frac{1}{n} \sum_{j=1}^n \Phi(x_i, x_j)(v^i - v^j),$$

where $x_i \in K \subset \mathbb{R}^d$ is the location of neuron i .

- We obtain the following **microscopic model**, where $1 \leq i \leq N$

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

(1)

Mean field limit

It was proved that in the **mean field limit** $N \rightarrow +\infty$ the microscopic system is described by

FitzHugh-Nagumo's mean field equation

$$\begin{cases} \partial_t f = -\partial_v ((N(v) - w - \mathcal{K}_\Phi(f)) f) - \partial_w (A(v, w) f) + \partial_v^2 f, \\ f(0, \cdot) = f_0, \end{cases}$$

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

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

References

- J. Baladron, D. Fasoli, O. Faugeras and J. Touboul. *Mean-field description and propagation of chaos in networks of Hodgkin-Huxley and FitzHugh-Nagumo neurons* (2012).
- M. Bossy, O. Faugeras and D. Talay. *Clarification and complement to "mean-field description and propagation of chaos in networks of Hodgkin-Huxley and FitzHugh-Nagumo neurons"* (2015).
- S. Mischler, C. Quiñinao and J. Touboul. *On a kinetic FitzHugh-Nagumo model of neuronal network* (2015).
- E. Luçon and W. Stannat. *Mean-field limit for disordered diffusions with singular interactions* (2014).
- J. Crevat. *Mean-field limit of a spatially-extended FitzHugh-Nagumo neural network*(2019).

Concentration phenomena

Modeling assumptions

We decompose the interaction kernel Φ as follows

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

where

- Ψ is "more regular" and accounts for **Weak/Long range interactions**,
- dirac mass δ_0 accounts for **Strong/Local interactions** with strength $\epsilon > 0$.

GOAL :

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

References :

Bressloff (03), Luçon/Stannat (14).

Strong interactions & concentration phenomenon

- With this ansatz the equation rewrites

$$\begin{aligned} \partial_t f^\epsilon = - \partial_v \left(\left(N(v) - w - \frac{\rho_0^\epsilon}{\epsilon} (v - \mathcal{V}^\epsilon) - \mathcal{K}_\Psi(f^\epsilon) \right) f^\epsilon \right) \\ - \partial_w (A(v, w) f^\epsilon) + \partial_v^2 f^\epsilon, \end{aligned} \quad (2)$$

where

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

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

$$\int_{\mathbb{R}^2} |v - \mathcal{V}^\epsilon(t)|^2 f^\epsilon dv dw \underset{\epsilon \rightarrow 0}{=} O(\epsilon).$$

Hence, f^ϵ is expected to **concentrate around** \mathcal{V}^ϵ

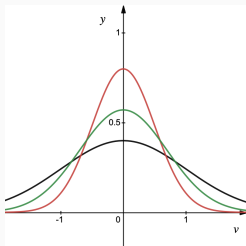
$$f^\epsilon(t, v, w) \underset{\epsilon \rightarrow 0}{=} \delta_{\mathcal{V}^\epsilon(t)}(v) \otimes F^\epsilon(t, w) + \sqrt{\epsilon},$$

where $F^\epsilon(t, w) = \int_{\mathbb{R}} f^\epsilon(t, v, w) dv$, in some probability space.

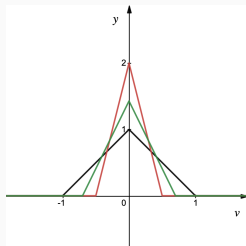
Concentration's profile

Concentration's profile

Our goal is to **determine the profile** of this concentration.



Concentration with **Gaussian profile**



Concentration with **triangular profile**

Here are plots of

$$y = \frac{1}{\sqrt{\epsilon}} g\left(\frac{v}{\sqrt{\epsilon}}\right),$$

for $\sqrt{\epsilon} = 1; 0.7; 0.5$ and g a gaussian profile (fig. 1) and triangular profile (fig. 2).

Goal of the presentation

- Consider the following re-scaled version g^ϵ of f^ϵ :

$$f^\epsilon(t, v, w) = \frac{1}{\epsilon^\alpha} g^\epsilon \left(t, \frac{v - \mathcal{V}^\epsilon(t)}{\epsilon^\alpha}, w - \mathcal{W}^\epsilon(t) \right), \quad (3)$$

where $\alpha > 0$ needs to be determined, and

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

- GOAL :

proving that g^ϵ converges and compute the limit.

Formal derivation of the concentration's profile

We obtain the equation on g^ϵ changing variables in the equation on f^ϵ

$$\begin{aligned}\partial_t g^\epsilon = & -\frac{1}{\epsilon^\alpha} \partial_u [(N(\mathcal{V}^\epsilon + \epsilon^\alpha u) - N(\mathcal{V}^\epsilon) - (\Psi *_r \rho_0^\epsilon) \epsilon^\alpha u - \omega - \mathcal{E}) g^\epsilon] \\ & - \partial_\omega [A_0(\epsilon^\alpha u, \omega) g^\epsilon] + \frac{1}{\epsilon^{2\alpha}} \partial_u (\rho_0^\epsilon \epsilon^{2\alpha-1} u g^\epsilon + \partial_u g^\epsilon),\end{aligned}$$

where $A_0 = A - c$ and \mathcal{E} is an error term.

- It is natural to take $\alpha = 1/2$. Indeed, we obtain

$$\begin{aligned}\partial_t g^\epsilon = & -\frac{1}{\sqrt{\epsilon}} \partial_u [(N(\mathcal{V}^\epsilon + \sqrt{\epsilon} u) - N(\mathcal{V}^\epsilon) - (\Psi *_r \rho_0^\epsilon) \sqrt{\epsilon} u - \omega - \mathcal{E}) g^\epsilon] \\ & - \partial_\omega [A_0(\sqrt{\epsilon} u, \omega) g^\epsilon] + \frac{1}{\epsilon} \partial_u (\rho_0^\epsilon u g^\epsilon + \partial_u g^\epsilon).\end{aligned}\tag{4}$$

Formal derivation

- Considering the stiffer term in the former equation, we expect

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

where $\mathcal{M}_{\rho_0^\epsilon}(u) = \sqrt{\frac{\rho_0^\epsilon}{2\pi}} \exp\left(-\frac{\rho_0^\epsilon u^2}{2}\right)$ and $G^\epsilon(t, x, \omega) = \int_{\mathbb{R}} g^\epsilon du$.

- Furthermore, since G^ϵ solves

$$\partial_t G^\epsilon - b\partial_\omega(xG^\epsilon) = -a\sqrt{\epsilon}\partial_\omega\left(\int_{\mathbb{R}} ug^\epsilon du\right), \quad (5)$$

and

$$\int_{\mathbb{R}} ug^\epsilon du \underset{\epsilon \rightarrow 0}{=} \int_{\mathbb{R}} u\mathcal{M}_{\rho_0^\epsilon} \otimes G^\epsilon du + O(\sqrt{\epsilon}) = O(\sqrt{\epsilon}),$$

it is expected that

$$G^\epsilon(t, x, \omega) \underset{\epsilon \rightarrow 0}{=} G(t, x, \omega) + O(\epsilon),$$

where G solves

$$\partial_t G - b\partial_\omega(\omega G) = 0.$$

Main result

Main result

Coming back to the solution f^ϵ to the mean field equation, this yields

Théorème

Under some assumptions on f_0^ϵ , there exists a constant $C > 0$ such that

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

for all $\epsilon > 0$ and $t \geq 0$, where the macroscopic system (\mathcal{V}, F) solves the following coupled reaction-diffusion/transport equation

$$\left\{ \begin{array}{l} \partial_t \mathcal{V} = N(\mathcal{V}) - \mathcal{W} - (\Psi *_r \rho_0(x) \mathcal{V} - \Psi *_r (\rho_0 \mathcal{V}))(x), \\ \partial_t F + \partial_w (A(\mathcal{V}, w) F) = 0, \\ \rho_0(x) \mathcal{W} = \int_{\mathbb{R}} w F dw. \end{array} \right. \quad (6)$$

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

Key arguments & Comments

- Since $W_2(\mathcal{M}_{\rho_0/\epsilon}, \delta_0) = \sqrt{\epsilon/\rho_0}$, we recover

$$\sup_{x \in K} W_2 \left(\frac{1}{\rho_0^\epsilon} f^\epsilon, \frac{1}{\rho_0} \delta_{\mathcal{V}} \otimes F \right) \underset{\epsilon \rightarrow 0}{\sim} \sqrt{\epsilon},$$

Key argument for the proof :

- Uniform estimates (in time and ϵ) for moments and some relative energy using **confining properties** of A and N (uniformity in time) and **concentrating properties** of the stiffer term $-\frac{1}{\epsilon}(\nu - \mathcal{V}^\epsilon)$ (uniformity in ϵ).
- **Coupling method** in order to estimate the Wasserstein distance between g^ϵ and $M \otimes G$

References :

Fournier/Perthame (2019).

- Obtaining similar results following a Hamilton-Jacobi method.
- Obtaining a strong convergence result.
- Numerical analysis of the model.

Thank you for your attention !