

# Modèles de neurones en champ-moyen avec interactions spatiales

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Travail en commun avec Wilhelm Stannat (TU Berlin)

[L., Stannat, 2014]

[L., Stannat, 2015, arXiv :1502.00532]

# Dynamics of large population of interacting neurons

- Consider a population a  $N$  interacting neurons, each of them described by its action potential  $V_i$ , and possibly by others variables (eg. gating variables, etc.)  
→ possibly very complex dynamics
- The graph of interaction is unknown, random and a priori very complex too.  
Common assumption : interaction on the complete graph  $\leftrightarrow$  mean-field models

Two main types of models :

- Models of interacting diffusions (Hodgkin-Huxley, FitzHugh-Nagumo [Baladron, Fasoli, Faugeras, Touboul, '12, Faugeras, MacLaurin, '14, etc.]),
- Jump processes (integrate-and-fire [Delarue, Inglis, Rubenthaler, Tanré, '15, Inglis, Talay, '15.], Hawkes processes [Brémaud-Massoulié, '96, Reynaud-Bouret, Rivoirard, Grammont, Tuleau-Malot, '14., Chevallier, '15])

## Mean-field diffusions within random environment

We consider the system of  $N$  stochastic differential equations

$$d\theta_{i,t} = c(\theta_{i,t}, \omega_i) dt + \frac{1}{N} \sum_{j=1}^N \Gamma(\theta_{i,t}, \omega_i, \theta_{j,t}, \omega_j) dt + dB_{i,t}, \quad i = 1, \dots, N,$$

- $c(\cdot)$  : local dynamics of the particle  $\theta_i$
- $\Gamma(\cdot)$  : interaction kernel between two particles,
- $\{B_i\}_i$  : IID standard Brownian motions (thermal noise).
- $\{\omega_i\}_i$  : IID random variables with law  $\mu$  (local inhomogeneity of the particles, **random environment**), independent with the thermal noise.

## Example 1 : The Kuramoto model

- $\theta_i \in \mathbb{S} := \mathbb{R}/2\pi$  (phase oscillators),
- $c(\theta, \omega) = \omega$ ,
- $\Gamma(\theta, \omega, \tilde{\theta}, \tilde{\omega}) = K \sin(\tilde{\theta} - \theta)$

$$d\theta_{i,t} = \omega_i dt + \frac{K}{N} \sum_{j=1}^N \sin(\theta_{j,t} - \theta_{i,t}) dt + \sigma dB_{i,t}, \quad i = 1, \dots, N,$$

### Remarks

- The real parameter of the model is  $K/\sigma^2$ . In the following we take  $\sigma = 1$ .
- The model is invariant by rotation : if  $\{\theta_j(t)\}_{j=1\dots N}$  is solution, so is  $\{\theta_j(t) + \psi\}_{j=1\dots N}$ , for all  $\psi \in \mathbb{S}$ .
- When  $\omega_i \equiv 0$ , the process is reversible under the invariant measure  $\pi_{N,K}$  (**Hamiltonian Mean-Field model, HMF or XY model**)

$$\pi_{N,K}(d\theta) \propto \exp\left(\frac{K}{N} \sum_{i,j=1}^N \cos(\theta_i - \theta_j)\right) d\theta$$

## Example 2 : FitzHugh-Nagumo oscillators for neuronal dynamics

- Dynamics of one neuron  $\theta = (V, W) \in \mathbf{R}^2$  :

$$c(V, W) = \begin{pmatrix} \frac{1}{\varepsilon} (V - V^3/3 + W + I) \\ aW + bV, \end{pmatrix}$$

$V$  : membrane potential,  $W$  *recovery variable*. [📖 Berglund, Gentz - 2007, Berglund, Landon 2012]

- Random environment  $\omega = (a, b)$  : inhomogeneous discrimination between inhibited/excited neurons.
- $\Gamma$  : synaptic coupling between neurons [📖 Baladron, Fasoli, Faugeras, Touboul - 2012]

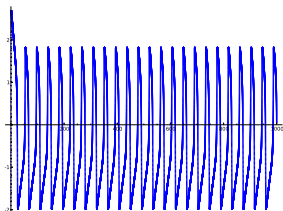
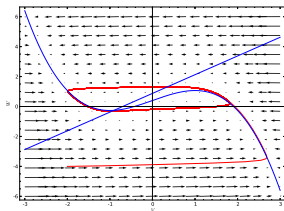
### Main features

- The dynamics  $c$  is not globally-Lipschitz
- Dynamics with polynomial bounds
- But  $c$  is monotone :

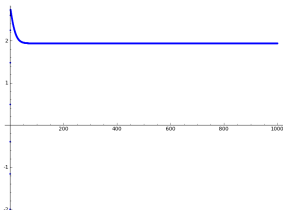
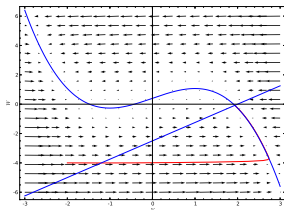
$$\langle c(\theta, \omega) - c(\tilde{\theta}, \omega), \theta - \tilde{\theta} \rangle \leq -C \|\theta - \tilde{\theta}\|^2.$$

Depending on the values of the local disorder  $\omega = (a, b)$ , one obtains two different behaviors

- Excitation :



- Inhibition :



## Propagation of chaos and large population behavior

All the statistical information is contained in the disordered empirical measure of the system

$$\mathbf{v}_{N,t} := \frac{1}{N} \sum_{j=1}^N \delta_{(\theta_{j,t}, \omega_j)}, \quad t \in [0, T]$$

The system can be rewritten as

$$d\theta_{i,t} = c(\theta_{i,t}, \omega_i) dt + \int \Gamma(\theta_{i,t}, \omega_i, \theta, \omega) \mathbf{v}_{N,t}(d\theta, d\omega) dt + dB_{i,t}, \quad i = 1, \dots, N.$$

Taking formally the limit as  $N \rightarrow \infty$  (i.e. suppose that  $\theta_i \rightarrow \bar{\theta}$  and  $\mathbf{v}_N \rightarrow \mathbf{v}$ ), one obtains (propagation of chaos) **the nonlinear process** [Sznitman '91]:






$$d\bar{\theta}_t = c(\bar{\theta}_t, \omega) dt + \int \Gamma(\bar{\theta}_t, \omega, \tilde{\theta}, \tilde{\omega}) \mathbf{v}_t(d\tilde{\theta}, d\tilde{\omega}) dt + dB_t.$$

with  $\mathbf{v}_t =$  law of  $(\bar{\theta}_t, \omega)$ .

Writing  $\mathbf{v}_t(d\theta, d\omega) = q_t(\theta, \omega) d\theta \mu(d\omega)$ , it is the weak solution of the McKean-Vlasov equation

$$\partial_t q_t(\theta, \omega) = \frac{1}{2} \Delta q_t(\theta, \omega) - \operatorname{div}_{\theta} \left( q_t(\theta, \omega) \left( c(\theta, \omega) + \int \Gamma(\theta, \omega, \tilde{\theta}, \tilde{\omega}) q_t(\tilde{\theta}, \tilde{\omega}) d\tilde{\theta} \mu(d\tilde{\omega}) \right) \right).$$

## Known results

- 1 Quenched and averaged law of large numbers [ Gärtner, Oelschläger '87, Dai Pra, den Hollander '96, L. '11],
- 2 Quenched and averaged fluctuations of  $v_N$  around its mean-field limit : [ Fernandez, Méléard '97], [L. '11]
- 3 Averaged large deviations as  $N \rightarrow \infty$  : [ Dai Pra, den Hollander '96]
- 4 Linear and nonlinear stability results of the Kuramoto model [ Bertini Giacomini Pakdaman '10], [Giacomini, Pakdaman, Pellegrin, Poquet, '12], [Giacomini, Pakdaman, Pellegrin, '12], [Giacomini, L. Poquet, '14]
- 5 Long-time dynamics for the Kuramoto model and effect of quenched disorder [ Bertini, Giacomini, Poquet, '13], [L. Poquet, '15]



# Generalization to spatially-extended mean-field systems

Motivation : understand the behavior as  $N \rightarrow \infty$  of similar systems where :

- the interaction no longer follows the complete graph (random graphs), or
- the interaction is still mean-field but the strength of interaction is not uniform along the population

## Intuition

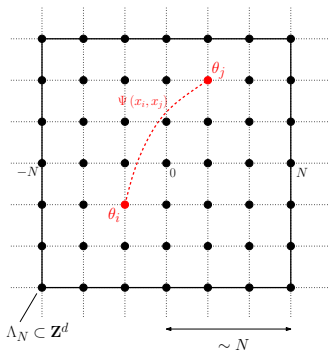
If each particle has enough connections, one should keep the same mean-field properties (law of large numbers, fluctuations, large deviations).

## Mean-field diffusions with spatial interaction

Each particle  $\theta_i$  has a fixed position  $x_i$  in  $\mathbb{R}^d$ ,

$$x_i := \frac{i}{2N}, \quad i \in \Lambda_N,$$
$$\Lambda_N := \{-N, \dots, N\}^d.$$

The interaction between  $\theta_i$  and  $\theta_j$  depends on a spatial kernel  $\Psi(x_i, x_j)$ .



$$d\theta_{i,t} = c(\theta_{i,t}, \omega_i) dt + \frac{1}{|\Lambda_N|} \sum_{\substack{j \in \Lambda_N \\ j \neq i}} \Gamma(\theta_{i,t}, \omega_i, \theta_{j,t}, \omega_j) \Psi(x_i, x_j) dt + dB_{i,t},$$

with a (possibly irregular) spatial kernel  $\Psi(\cdot, \cdot)$ .

### Remark

It is possible to impose periodic boundary conditions on  $\Lambda_N$ .

## Existing models

- 1 Kuramoto-type oscillators with phase-lag and space :  $\Gamma(\theta, \tilde{\theta}) = \sin(\tilde{\theta} - \theta + \alpha)$   
with regular kernels [📖 Kuramoto, Battogtokh, '02]

$$\Psi(x, y) \propto \exp(-\kappa|x-y|),$$

$$\Psi(x, y) \propto 1 + A \cos(x-y).$$

[📖 Abrams, Strogatz - Int. Journ. Bifurc. Chaos, '06] or general kernels [📖 O. Omel'chenko - Nonlin., '13], [M. Wolfrum, O. Omel'chenko, S. Yanchuk, Y. Maistrenko, Chaos, '11], [O. Omel'chenko, M. Wolfrum, C. Laing, Chaos '14], etc.

- 2  $P$ -nearest neighbor model : each particle interacts only with a fixed-proportion  $RN$  (with  $R \in ]0, 1]$ ) of its nearest neighbors :

$$\Psi(x, y) := \frac{1}{R^d} \mathbf{1}_{[-R, R]^d}(|x-y|).$$

[📖 I. Omelchenko, B. Riemenschneider, P. Hövel, Y. Maistrenko, E. Schöll, Phys. Rev E, '12] (Rössler system), [📖 A. Vüllings, J. Hizanidis, I. Omelchenko, P. Hövel, '14] (FitzHughNagumo oscillators), and many others. . .

- 3 Kuramoto oscillators with polynomial decay ( $\alpha$ -Hamiltonian Mean-Field model,  $\alpha$ -HMF)

$$\Psi(x, y) := \frac{1}{|x-y|^\alpha}$$

[📖 Gupta, Potters, Ruffo - Phys. Rev. E '12], [Gupta, Campa, Ruffo - '12], [Gupta, Campa, Ruffo - Kuramoto model of synchronization : equilibrium and nonequilibrium aspects, '14]

# Large population behavior of spatially-extended systems

We focus here on the rigorous derivation of the large population behavior for spatially extended models, with special attention given to the **singular spatial kernels**, namely the  **$P$ -nearest neighbor model** and the model with **polynomial decay**.

- 1 Law of large numbers for the empirical measure and propagation of chaos ?
- 2 Well-posedness of the McKean-Vlasov equation ?
- 3 Fluctuations around its limit ?

## The empirical measure and the McKean-Vlasov equation

Look at

$$d\theta_{i,t} = c(\theta_{i,t}, \omega_i) dt + \frac{1}{|\Lambda_N|} \sum_{\substack{j \in \Lambda_N \\ j \neq i}} \Gamma(\theta_{i,t}, \omega_i, \theta_{j,t}, \omega_j) \Psi(x_i, x_j) dt + dB_{i,t}.$$

The empirical measure of the system is given by

$$\nu_{N,t} = \frac{1}{|\Lambda_N|} \sum_{i \in \Lambda_N} \delta_{(\theta_{i,t}, \omega_i, x_i)}.$$

Its natural limit is given by  $\nu_t(d\theta, d\omega, dx) = q_t(\theta, \omega, x) d\theta \mu(d\omega) dx$  where  $q$  solves

$$\partial_t q_t = \frac{1}{2} \Delta_{\theta} q_t - \operatorname{div}_{\theta} \left( q_t \left\{ c(\cdot) + \int \Gamma(\cdot, \bar{\theta}, \bar{\omega}) \Psi(\cdot, \bar{x}) q_t(\bar{\theta}, \bar{\omega}, \bar{x}) d\bar{\theta} d\mu(\bar{\omega}) d\bar{x} \right\} \right)$$

Problem : is (the weak formulation of) this equation well-posed ( $c$  unbounded,  $\Psi$  possibly singular) ?

Law of large numbers in  $P$ -nearest neighbor case :  $\Psi(x, y) = \frac{1}{R^d} \mathbf{1}_{|x-y| \leq R}$

### Theorem (L., Stannat - 2014)

For all  $d \geq 1$ ,  $R \in ]0, 1]$ , under some regularity assumptions on  $c$  and  $\Gamma$  and some moments hypothesis on the initial condition and the disorder,

- The weak formulation of the McKean-Vlasov equation with space is well-posed,
- The empirical measure  $\nu_N$  converges in law as  $N \rightarrow \infty$ , in  $C([0, T], \mathcal{M}_1(\mathbb{R}^m \times \mathbb{R}^n \times [-1/2, 1/2]^d))$  to the unique solution  $\nu$  to the McKean-Vlasov equation.

Moreover, there exists a constant  $C$  (depending only on  $T$ ,  $R$ ,  $c$  and  $\Gamma$ ) such that

$$\sup_{0 \leq t \leq T} d(\nu_{N,t}, \nu_t) \leq \frac{C}{N^{1 \wedge \frac{d}{2}}},$$

for some distance  $d$  of Wasserstein type.

### Remarks

- The same estimates holds for any regular spatial weight  $\Psi$ ,
- In small dimension ( $1 \leq d \leq 2$ ), we obtain the Gaussian scaling  $N^{\frac{d}{2}}$ ,
- In higher dimension ( $d \geq 3$ ), the speed of convergence is governed by the spatial constraints.

## Law of large numbers in the polynomial case : $\Psi(x, y) = \frac{1}{|x-y|^\alpha}$ , ( $\alpha < d$ )

### Theorem (L., Stannat - 2014)

For all  $d \geq 1$ ,  $\alpha < d$ , under some regularity assumptions on  $c$  and  $\Gamma$  and some moments hypothesis on the initial condition and the disorder, one has a similar result concerning the well-posedness of the McKean-Vlasov equation and the convergence of the empirical measure.

Moreover, one has the following estimate : for all  $\alpha \leq \gamma < \frac{d}{2}$ ,

$$\sup_{0 \leq t \leq T} d(\mathbf{v}_{N,t}, \mathbf{v}_t) \leq C \begin{cases} \frac{1}{N^\gamma}, & \text{if } \alpha \in [0, \frac{d}{2}), \\ \frac{\ln N}{N^{\frac{d}{2}}}, & \text{if } \alpha = \frac{d}{2}, \\ \frac{\ln N}{N^{d-\alpha}}, & \text{if } \alpha \in (\frac{d}{2}, d). \end{cases}$$

for some distance  $d$  of Wasserstein type.

### Remark

For small spatial constraints ( $\alpha < \frac{d}{2}$ ), we retrieve (approximately) the Gaussian scaling, whereas for strong spatial constraints, a nontrivial scaling appears.

**Question :** Are the scaling found the correct ones ?

## Idea of proof

Introduce the nonlinear process associated to the particle system

$$d\theta_t = c(\theta_t, \omega) dt + \int \Gamma(\theta_t, \omega, \bar{\theta}, \bar{\omega}) \Psi(x, y) \mathbf{v}_t(d\bar{\theta}, d\bar{\omega}, dy) + dB_t.$$

and define its propagator

$$P_{s,t} f(\theta, \omega, x) := \mathbf{E}_{\theta_s = \theta}(f(\theta_t, \omega, x)).$$

$P_{s,t}$  satisfies a **Backward Kolmogorov equation** which enables to get rid of the second derivative :

$$\begin{aligned} \langle f, \mathbf{v}_{N,t} - \mathbf{v}_T \rangle &= \langle P_{0,t} f, \mathbf{v}_{N,0} - \mathbf{v}_0 \rangle + \frac{1}{|\Lambda_N|} \sum_k \int_0^t \partial_{\theta}(P_{s,t} f)(\theta_{k,s}, \omega_k, x_k) dB_{k,s} \\ &\quad + \frac{1}{|\Lambda_N|} \sum_k \int_0^t \partial_{\theta}(P_{s,t} f)(\theta_{k,s}, \omega_k, x_k) \langle \Gamma(\theta_{k,s}, \omega_k, \cdot) \Psi(x_k, \cdot), \mathbf{v}_{N,s} - \mathbf{v}_s \rangle ds. \end{aligned}$$

**Key point** : find a space  $C$  of regular (Lipschitz) functions that is stable under  $P_{s,t}$  + Gronwall Lemma.



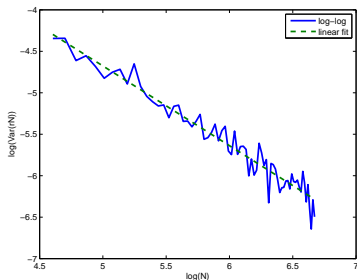
# Fluctuations of the order parameter in the Kuramoto model

The correct order parameter in the Kuramoto model is

$$r_{N,t} := \frac{1}{N} \sum_{j=1}^N e^{i\theta_{j,t}} \rightarrow_{N \rightarrow \infty} r_t := \int e^{i\theta} q_t(d\theta, d\omega).$$

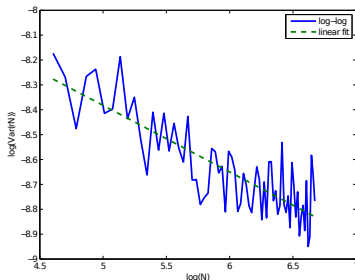
Log-log computation of  $(\mathbb{E}(r_{N,t} - r_t)^2)^{1/2}$  w.r.t.  $N$  :

$\alpha = 0$



Result : decay as  $\approx 0.481$

$\alpha = 0.9$



Result : decay as  $\approx 0.134$

## Fluctuations around the limit I

Suppose for simplicity  $d = 1$ . Looking at the fluctuations of the empirical measure  $\nu_N$  around its limit  $\nu$  is to look at the (distribution-valued) process

$$\eta_N = a_N(\nu_N - \nu).$$

For which renormalization  $a_N$  ?

One wants to compute the speed of convergence of the particle system

$$d\theta_{i,t} = c(\theta_{i,t}, \omega_i) dt + \frac{1}{|\Lambda_N|} \sum_{\substack{j \in \Lambda_N \\ j \neq i}} \Gamma(\theta_{i,t}, \omega_i, \theta_{j,t}, \omega_j) \Psi(x_i, x_j) dt + dB_{i,t}$$

to the nonlinear process  $\bar{\theta}_t = \bar{\theta}_t(\omega, x)$

$$d\bar{\theta}_t = c(\bar{\theta}_t, \omega) dt + \int \Gamma(\bar{\theta}_t, \omega, \tilde{\theta}, \tilde{\omega}) \Psi(x, \tilde{x}) \nu_t(d\tilde{\theta}, d\tilde{\omega}, d\tilde{x}) dt + dB_t,$$

## Fluctuations around the limit II

One has two scales of convergence :

- The convergence of the empirical distribution of the empirical measure of the Brownian motion and the initial condition : **typical scale is  $\sqrt{N}$**
- The convergence with respect to the space variable :

$$\frac{1}{|\Lambda_N|} \sum_{k \in \Lambda_N} \frac{1}{\left| \frac{k}{2N} \right|^\alpha} \xrightarrow{N \rightarrow \infty} \int_{-1/2}^{1/2} \frac{1}{|x|^\alpha} dx.$$

It is related to the speed of convergence of the Riemann sum related to  $x \mapsto \frac{1}{|x|^\alpha}$  towards the corresponding integral : **typical scale is  $N^{1-\alpha}$** .

## Phase transition ( $d = 1$ )

One should take the smallest renormalization in all cases

- For  $0 \leq \alpha < \frac{1}{2}$ , the proper renormalization is  $a_N := \sqrt{N}$ ,
- For  $\frac{1}{2} < \alpha < 1$ , the proper renormalization is  $a_N := N^{1-\alpha}$ .

### Intuition

- When  $\alpha < \frac{1}{2}$ , the spatial constraints are too weak, the randomness should prevail and everything should happen like we were in the full mean-field case ( $\alpha = 0$ ) : the limit of  $\eta_N$  should be random (Gaussian),
- The case  $\alpha > \frac{1}{2}$  corresponds to a strong spatial attenuation. What we see at first order should not be the Gaussian fluctuations but the constraints due to the spatial kernel : the limit of  $\eta_N$  should be deterministic.

### Remark

This scaling was first noticed by [\[Firpo, Ruffo - J Phys A, 2001\]](#) in the case of the  $\alpha$ -Hamiltonian Mean-Field model.

## Fluctuation process

Question : is it the right scaling ? Do we have an actual Central Limit Theorem ? Let us go back to  $\eta_N = a_N(v_N - v)$ . Strategy to study  $\eta_N$  :

- 1 Write a semi-martingale decomposition for  $\eta_N$ ,
- 2 Prove tightness of the process  $\eta_N$  in a certain Sobolev space of regularity of type  $H^{-k}$ ,
- 3 Identify and prove uniqueness of the limit.

[Fernandez, Méléard - SPA, 1997], [Meleard, Roelly - SPA, 1987], [Jourdain, Méléard - Ann. IHP 1998], [Oelschläger, 1987], [L. 2011]

Problem : here, if one looks at  $\eta_N$  only, we need to consider test functions  $(\theta, \omega, x) \mapsto f(\theta, \omega, x)$  that are **singular w.r.t. the space variable  $x$**  (in order to capture singularities such as  $x \mapsto \frac{1}{|x|^\alpha}$ ).

This does not work w.r.t the usual tightness criteria.

## Two-particle fluctuation process

Idea : build an auxiliary fluctuation process that carries the singularity in space. This process takes into account the mutual fluctuations of **two** particles  $(\theta_i, \theta_j)$  instead of one.

$$\langle \mathcal{H}_{N,t}, g \rangle := a_N \left( \frac{1}{|\Lambda_N|^2} \sum_{i,j \in \Lambda_N} \Psi(x_i, x_j) g(\theta_{i,t}, \theta_{j,t}) - \frac{1}{|\Lambda_N|} \sum_{i \in \Lambda_N} \int \Psi(x_i, \tilde{x}) g(\theta_{i,t}, \tilde{\theta}) d\nu_t \right)$$

Now, it is the process itself that carries the singularity in space, not the test functions.

### Remark

This process gives the correct renormalization in space : if  $g \equiv 1$ , it boils down to

$$a_N \left( \frac{1}{|\Lambda_N|^2} \sum_{i,j \in \Lambda_N} \Psi(x_i, x_j) - \frac{1}{|\Lambda_N|} \sum_{i \in \Lambda_N} \int \Psi(x_i, \tilde{x}) d\tilde{x} \right)$$

which is exactly of order 1, when  $\alpha > \frac{1}{2}$ .

## Semi-martingale decomposition of $(\eta_N, \mathcal{H}_N)$

### Proposition (L., Stannat - 2015)

One has the following decomposition :

$$\langle \eta_{N,t}, f \rangle = \langle \eta_{N,0}, f \rangle + \int_0^t \langle \eta_{N,s}, L_s^{(1)} f \rangle ds + \int_0^t \langle \mathcal{H}_{N,s}, \Phi[f] \rangle ds + \mathcal{M}_{N,t}^{(\eta)} f,$$

$$\langle \mathcal{H}_{N,t}, g \rangle = \langle \mathcal{H}_{N,0}, g \rangle + \int_0^t \langle \mathcal{H}_{N,s}, L_s^{(2)} g \rangle ds + \int_0^t F_{N,s} g ds + \mathcal{M}_{N,t}^{(\mathcal{H})} g,$$

where  $L^{(1)}$  and  $L^{(2)}$  are explicit linear operators, with smooth coefficients.

## Sub-critical fluctuations ( $\alpha < \frac{1}{2}$ )

### Theorem (L., Stannat - 2015)

Suppose  $\alpha < \frac{1}{2}$ . Under regularity assumptions  $c$  and  $\Gamma$  and moment conditions on the disorder and initial condition, there exist a Sobolev space  $\mathbf{H}$  such that  $\eta_N$  converges as  $N \rightarrow \infty$  in  $C([0, T], \mathbf{H})$  to the unique solution of the following linear stochastic partial differential equation

$$\eta_t = \eta_0 + \int_0^t L_s^{(1)*} \eta_s ds + \mathcal{M}_t^{(\eta)}, \quad t \in [0, T],$$

where  $\theta_0$  and  $\mathcal{M}^{(\eta)}$  are independent Gaussian processes.

### Remarks

- In the case  $\alpha < \frac{1}{2}$ , the process  $\mathcal{H}_N$  vanishes as  $N \rightarrow \infty$ .
- This generalizes the result of [Fernandez, Méléard - SPA, 1997], [L. 2011] where the case  $\alpha = 0$  was considered.
- This is an **averaged** result. A **quenched** result is also true (see [L. 2011]).



## Supercritical fluctuations ( $\alpha > \frac{1}{2}$ )

### Theorem (L. Stannat - 2015)

Suppose  $\alpha > \frac{1}{2}$ . Under regularity assumptions on  $c$  and  $\Gamma$  and moment conditions on the disorder and initial condition, there exist Sobolev spaces  $(\mathbf{H}_1, \mathbf{H}_2)$  such that the couple  $(\eta_N, \mathcal{H}_N)$  converges in law in  $C([0, T], \mathbf{H}_1 \oplus \mathbf{H}_2)$  to the unique solution of the system of coupled PDEs

$$\begin{cases} \eta_t = \int_0^t L_s^{(1),*} \eta_s ds + \int_0^t \Phi^* \mathcal{H}_s ds, \\ \mathcal{H}_t = \mathcal{H}_0 + \int_0^t L_s^{(2),*} \mathcal{H}_s ds, \end{cases} \quad t \in [0, T],$$

where  $\mathcal{H}_0$  is explicit.

### Remark

Unfortunately, our assumptions do not cover the FitzHugh-Nagumo case.

## On the dynamical properties of the continuous Kuramoto model with space

Look at the stationary states of the Kuramoto equation with space

$$0 = \frac{1}{2} \partial_{\theta}^2 q(\theta, x) - K \partial_{\theta} \left( q(\theta, x) \int \sin(\tilde{\theta} - \theta) \Psi(x, \tilde{x}) q(\tilde{\theta}, \tilde{x}) d\tilde{\theta} d\tilde{x} \right).$$

[[Gupta, Campa, Ruffo - 2014](#)] : introduce magnetization parameters at position  $x$

$$m_X(x) := \int \cos(\theta) q(\theta, x) d\theta, \quad m_Y(x) := \int \sin(\theta) q(\theta, x) d\theta,$$

and their spatial transform

$$\hat{m}_X(x) := \int \Psi(x, y) m_X(y) dy, \quad \hat{m}_Y(x) := \int \Psi(x, y) m_Y(y) dy.$$

Then, any stationary solution can be written as

$$q(\theta, x) = \frac{1}{Z(x)} \exp(e^{2K \cos(\theta) \hat{m}_X(x) + 2K \sin(\theta) \hat{m}_Y(x)})$$

where the magnetization parameters satisfy the consistency relation

$$\sqrt{m_X(x)^2 + m_Y(x)^2} = \frac{I_1}{I_0} \left( 2K \sqrt{\hat{m}_X(x)^2 + \hat{m}_Y(x)^2} \right).$$

## Conclusion

- For  $\alpha > \frac{1}{2}$ , what is the next term in the asymptotic development of  $v_N$  ?
- What if one chooses positions  $x$  randomly ? Positions depending on time ?
- What are the dynamical properties of the Kuramoto model with space ? Synchronisation ? Stability of synchronized profiles ? Chimera states ?
- What about random graphs ? Is there a mean-field limit for the empirical measure too ?

Merci de votre attention !

- E. Luçon and W. Stannat. *Mean field limit for disordered diffusions with singular interactions*. Ann. Appl. Probab., 24(5) :1946–1993, 2014.
- E. Luçon and W. Stannat. *Transition from Gaussian to non-Gaussian fluctuations for mean-field diffusions in spatial interaction*, arXiv : 1502.00532, 2015.