

Quantitative, qualitative and uniform in time propagation of chaos

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*A survey on recent papers by the “Kinetic French school”:
Carrapatoso, Fournier, Guillin, Hauray, M., Mouhot and co-authors*

Stochastic Limit Analysis for
Reacting Particle System workshop
Berlin, December 16-18, 2015

- General discussion about propagation of chaos
 - ▶ introduction to mean field limit / propagation of chaos
 - ▶ short discussion about chaos
 - ▶ quantitative/qualitative and uniform in time chaos
 - ▶ four different methods
- 2D viscous Vortex model and the nonlinear Martingale method
 - ▶ An example of “singular” McKean-Vlasov model
 - ▶ sketch of the proof
 - ▶ nonlinear Martingale method
 - ▶ Estimates thanks to Fisher information
 - ▶ Qualitative (entropic) chaos

Outlines of the talk

- 1 Introduction
- 2 Short discussion about chaos
- 3 Short discussion about methods
- 4 Nonlinear Martingale method and the vortex model
- 5 sketch of the proof - a priori estimates
- 6 sketch of the proof - probability argument
- 7 Sketch of the proof - functional analysis argument
- 8 Sketch of the proof - PDE/SDE argument
- 9 Sketch of the proof - entropy argument

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Micro to macro

- How to go **rigorously** from a microscopic description to a **statistical** description?
How to derive (justify) the equation at the **mesoscopic/macroscopic** level ?
How to get something (simpler) from a microscopic description with a huge number of particles ?
- (Kac's) mean field limit (\neq Boltzmann-Grad limit) in the sense that each particle interacts with all the other particles with an intensity of order $\mathcal{O}(1/N)$
 \Rightarrow statistical description = *law of large numbers limit* of a N -particle system
- at the formal level the identification of the limit is quite clear when one assumes the molecular chaos for the limit model
- main difficulty : propagation of chaos
 - ▷ chaos for ∞ particles = Boltzmann's molecular chaos (stochastic independence)
 - ▷ chaos for $N \rightarrow \infty$ particles = Kac's chaos (asymptotic stochastic independence)
 - ▷ propagation of chaos: holds at time $t = 0$ implies holds for any $t > 0$
 - ▷ propagation of chaos is necessary in order to identify the limit as $N \rightarrow \infty$

The Kac's approach (1956) for Boltzmann model and others - trajectories

Consider a system of N indistinguishable (exchangeable) particles, each particle being described by its state (position, velocity) $\mathcal{Z}_1^N, \dots, \mathcal{Z}_N^N \in E$, $E = \mathbb{R}^d$, which evolves according to

$$d\mathcal{Z}_i = \frac{1}{N} \sum_{j=1}^N K(\mathcal{Z}_i - \mathcal{Z}_j) dt + \sqrt{2\nu} d\mathcal{B}_i \quad (\text{ODE or Brownian SDE})$$

$$d\mathcal{Z} = \frac{1}{N} \sum_{i,j=1}^N \int_{S^{d-1}} (\mathcal{Z}'_{ij} - \mathcal{Z}) B d\mathcal{N}_{i,j}(d\sigma) \quad (\text{Boltzmann-Kac})$$

$$d\mathcal{Z}_i = \frac{1}{N} \sum_{j=1}^N b(\mathcal{Z}_i - \mathcal{Z}_j) dt + \frac{1}{\sqrt{N}} \sum_{j=1}^N a^{1/2}(\mathcal{Z}_i - \mathcal{Z}_j) d\mathcal{B}_{i,j} \quad (\text{Landau-Kac})$$

where K is a pairwise interaction force field, \mathcal{B}_i independent Brownian motions, $\nu \geq 0$, \mathcal{N} Poisson measure, $\mathcal{Z}'_{ij} = (\mathcal{Z}_1, \dots, \mathcal{Z}'_i, \dots, \mathcal{Z}'_j, \dots, \mathcal{Z}_N)$ represents the system after collision of the pair $(\mathcal{Z}_i, \mathcal{Z}_j)$, B cross-section, a Landau matrix kernel, $b = \text{div}_a$, $\mathcal{B}_{i,j}$ Brownian motions such that $\mathcal{B}_{j,i} = -\mathcal{B}_{i,j}$ and independent if $i < j$.

The Kac's approach (1956) for Boltzmann and others - Markov semigroup

The law $G^N(t) := \mathcal{L}(Z_t^N)$ satisfies the Master (Liouville or backward Kolmogorov) equation

$$\partial_t \langle G^N, \varphi \rangle = \langle G^N, \Lambda^N \varphi \rangle \quad \forall \varphi \in C_b(E^N)$$

where the generator Λ^N writes

$$(\Lambda^N \varphi)(Z) := \frac{1}{N} \sum_{i,j=1}^N K(z_i - z_j) \cdot \nabla_i \varphi + \nu \sum_{i=1}^N \Delta_i \varphi \quad (\text{ODE/SDE})$$

$$(\Lambda^N \varphi)(Z) = \frac{1}{N} \sum_{1 \leq i < j \leq N} \int_{\mathbb{S}^{d-1}} [\varphi(Z'_{ij}) - \varphi(Z)] B(z_i - z_j, \sigma) d\sigma \quad (\text{Boltzmann-Kac})$$

$$\begin{aligned} (\Lambda^N \varphi)(Z) &= \frac{1}{N} \sum_{1 \leq i, j \leq N} b(z_i - z_j) \cdot (\nabla_i \varphi - \nabla_j \varphi) \quad (\text{Landau-Kac}) \\ &+ \frac{1}{2N} \sum_{1 \leq i, j \leq N} a(z_i - z_j) : (\nabla_{ii}^2 \varphi + \nabla_{jj}^2 \varphi - \nabla_{ij}^2 \varphi - \nabla_{ji}^2 \varphi) \end{aligned}$$

What is the limit as $N \rightarrow \infty$

Is it possible to identify the limit of the law $\mathcal{L}(\mathcal{Z}_1^N)$ of one typical particle?

More precisely, we want to show that $\mathcal{L}(\mathcal{Z}_1^N) \rightarrow f = f(t, dz)$ and that $f \in C([0, \infty); P(E))$ is a solution to

$$\partial_t f = \operatorname{div}_z [(K * f)f] + \nu \Delta f \quad (\text{McKean} - \text{Vlasov})$$

$$\partial_t f = \int_{\mathbb{R}^d \times S^{d-1}} [f(z')f(v') - f(z)f(v)] B \, dz d\sigma \quad (\text{Boltzmann})$$

$$\partial_t f = \nabla_z \int_{\mathbb{R}^d} a(z - z_*) [f(z_*) \nabla f(z) - f(z) \nabla f(z_*)] \, dz_* \quad (\text{Landau}),$$

depending of the N -particle dynamics

Why those equations are the right limits ?

Assuming that

$$\mathcal{L}(\mathcal{Z}_1^N) \rightarrow f = f(t, dz), \quad \mathcal{L}(\mathcal{Z}_1^N, \mathcal{Z}_2^N) \rightarrow g = g(t, dz, dv),$$

we easily (formally) show by taking $\varphi(Z) = \varphi(z_1)$ in the Master equation

$$\partial_t f = \operatorname{div}_z \left[\int a(z - v) g(dz, dv) \right] + \nu \Delta f$$

$$\partial_t f = \int_{\mathbb{R}^d \times S^{d-1}} [g(z', v') - g(z, v)] B \, dv d\sigma$$

$$\partial_t f = \nabla_z \int_{\mathbb{R}^d} a(z - v) [\nabla_z g(z, v) - \nabla_v g(z, v)] \, dv.$$

We obtain the McKean-Vlasov equation, the Boltzmann equation and the Landau equation if we make the additional

independence / molecular chaos assumption $g(v, z) = f(v) f(z)$.

- for a infinite system of **indistinguishable** particles: Boltzmann's (molecular) chaos means

$$\mathcal{L}(\mathcal{Z}_i, \mathcal{Z}_j) = f \otimes f$$

That is the stochastic independence (for a sequence of **exchangeable** random variables)

- for a system of N **indistinguishable** particles with $N \rightarrow \infty$: Kac's chaos means

$$\mathcal{L}(\mathcal{Z}_i^N, \mathcal{Z}_j^N) \rightarrow f \otimes f \quad \text{as } N \rightarrow \infty$$

That is an asymptotically stochastic independence (of the coordinates of a sequence of random vectors with **exchangeable** coordinates)

Difficulty

- The above picture is not that easy because for N fixed particles the states $\mathcal{Z}_1(t), \dots, \mathcal{Z}_N(t)$ are **never independent** for positive time $t > 0$ even if the initial states $\mathcal{Z}_1(0), \dots, \mathcal{Z}_N(0)$ are assumed to be independent : that is an inherent consequence of the fact that **particles do interact!**
- Equations are written in spaces with increasing dimension $N \rightarrow \infty$.
To overcome that difficulty we **will** work in **fixed spaces** using:
empirical probability measure

$$X \in E^N \mapsto \mu_X^N := \frac{1}{N} \sum_{i=1}^N \delta_{x_i} \in \mathbb{P}(E)$$

and/or marginal densities

$$F^N \in \mathbb{P}_{sym}(E^N) \mapsto F_j^N := \int_{E^{N-j}} F^N dz_{j+1} \dots dz_N \in \mathbb{P}_{sym}(E^j)$$

- The nonlinear PDE can be obtained as a “*law of large numbers*” for a **not independent array of exchangeable random variables** in the mean-field limit.
- That is more demanding than the usual LLN. We need to **propagate** some asymptotic independence = Kac’s stochastic chaos.

- We need at least
 - ▷ a priori estimates on the N -particle system
 - ▷ uniqueness for the limit nonlinear PDE
- Most of the works has been done in a **probability measures framework**. In order that everything make sense, it is then needed that coefficients are not singular (they must be smooth enough, say C^0).
 - ▷ propagation of chaos is even more difficult for singular models
- For numerical simulation purpose, one can introduce Nanbu-like stochastic dynamic

$$\begin{aligned} dZ_i &= \frac{1}{N} \sum_{j=1}^N \int_{S^{d-1}} \tilde{B}(Z_i, Z_j, \sigma) d\mathcal{N}_i(d\sigma) \\ &= \frac{1}{N} \sum_{j=1}^N b(Z_i - Z_j) dt + \frac{1}{N} \sum_{j=1}^N a^{1/2}(Z_i - Z_j) dB_i \end{aligned}$$

They are simpler to analyze but they are not energy conservative

Kac's contribution and Kac's program

- Kac (1956) defined the notion of chaos for sequences of random vectors. He proved the propagation of chaos for the “*Kac's caricature*” of Boltzmann model. He showed that the stochastic dynamic leaves invariant the Kac's sphere

$$\mathcal{KS}^N := \{V \in \mathbb{R}^N; |v_1|^2 + \dots + |v_N|^2 = N\},$$

and, for any fixed $N \geq 2$, convergence to the equilibrium (stationary measure)

$$G_t^N = \mathcal{L}(\mathcal{V}_{1t}^N, \dots, \mathcal{V}_{Nt}^N) \xrightarrow[t \rightarrow \infty]{} \gamma^N = \text{uniform measure on } \mathcal{KS}^N.$$

Kac's Program:

- (Pb1) Establish propagation of chaos for more realistic (**singular**) models
- (Pb2) Establish the convergence to the equilibrium as $t \rightarrow \infty$ with a **uniform chaos** rate with respect to the number N of particles
- (Pb2') Establish **quantitative chaos** estimate (rate) for Kac's chaos
- (Pb3) Establish Boltzmann's H-theorem from a microscopic description (seems to be Kac's motivation)

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Definition of chaos

Chaos is the **asymptotic independence** as $N \rightarrow \infty$ for a sequence (\mathcal{Z}^N) of exchangeable random variables with values in E^N

$$\begin{array}{ccc} \mathcal{Z}^N = (\mathcal{Z}_1^N, \dots, \mathcal{Z}_N^N) \in E^N & \rightarrow & F^N := \mathcal{L}(\mathcal{Z}^N) \in \mathbb{P}_{\text{sym}}(E^N) \\ \updownarrow & & \updownarrow \\ \mu_{\mathcal{Z}^N}^N := \frac{1}{N} \sum_{i=1}^N \delta_{\mathcal{Z}_i^N} \in \mathbb{P}(E) & \rightarrow & \hat{F}^N := \mathcal{L}(\mu_{\mathcal{Z}^N}^N) \in \mathbb{P}(\mathbb{P}(E)) \end{array}$$

For a random variable \mathcal{Y} taking values in E with law $\mathcal{L}(\mathcal{Y}) = f \in \mathbb{P}(E)$ we say that (\mathcal{Z}^N) is \mathcal{Y} -Kac's chaotic if

- $\mathcal{L}(\mathcal{Z}_1^N, \dots, \mathcal{Z}_j^N) \rightharpoonup f^{\otimes j}$ weakly in $\mathbb{P}(E^j)$ as $N \rightarrow \infty$;
- $\mu_{\mathcal{Z}^N}^N \Rightarrow f$ in law as $N \rightarrow \infty$,
meaning $\mathcal{L}(\mu_{\mathcal{Z}^N}^N) \rightarrow \delta_f$ in $\mathbb{P}(\mathbb{P}(E))$ as $N \rightarrow \infty$;
- $\mathbb{E}(|\mathcal{X}^N - \mathcal{Y}^N|) \rightarrow 0$ as $N \rightarrow \infty$ for a sequence \mathcal{Y}^N of i.i.d.r.v with law f

Exchangeable means: $\mathcal{L}(\mathcal{Z}_{\sigma(1)}^N, \dots, \mathcal{Z}_{\sigma(N)}^N) = \mathcal{L}(\mathcal{Z}_1^N, \dots, \mathcal{Z}_N^N)$ for any permutation σ of the coordinates

Definition of chaos = not about random variables but their laws !

For a given sequence (F^N) in $\mathbb{P}_{\text{sym}}(E^N)$ we define

- the marginals $F_j^N \in \mathbb{P}_{\text{sym}}(E^j)$ by

$$F_j^N = \int_{E^{N-j}} F^N dz_{j+1} \dots dz_N$$

- the projection $\hat{F}^N \in \mathbb{P}(\mathbb{P}(E))$ by

$$\langle \hat{F}^N, \Phi \rangle = \int_{E^N} \Phi(\mu_X^N) F^N(dX) \quad \forall \Phi \in C_b(\mathbb{P}(E))$$

- the **normalized** MKW distance on $\mathbb{P}(E^j)$ by

$$W_1(F, G) := \inf_{\pi \in \Pi(F, G)} \int_{E^j \times E^j} \left(\frac{1}{j} \sum_{i=1}^j |x_i - x_j| \wedge 1 \right) \pi(dX, dY).$$

- the MKW distance \mathcal{W}_1 on $\mathbb{P}(\mathbb{P}(E))$ by

$$\mathcal{W}_1(\alpha, \beta) := \inf_{\pi \in \Pi(\alpha, \beta)} \int_{\mathbb{P}(E) \times \mathbb{P}(E)} W_1(\rho, \eta) \pi(d\rho, d\eta).$$

Quantitative comparison of the several Definitions of chaos

For a given sequence (F^N) in $\mathbb{P}_{\text{sym}}(E^N)$ we define

- the marginals $F_j^N \in \mathbb{P}_{\text{sym}}(E^j)$,
- the projection $\hat{F}^N \in \mathbb{P}(\mathbb{P}(E))$,
- the normalized MKW distance W_1 on $\mathbb{P}(E^j)$,
- the MKW distance \mathcal{W}_1 on $\mathbb{P}(\mathbb{P}(E))$,

and for $f \in \mathbb{P}(E)$ we say that (F^N) is f -Kac's chaotic if (equivalently)

- $\mathcal{D}_j(F^N; f) := W_1(F_j^N, f^{\otimes j}) = \mathbb{E}(|(\mathcal{X}_1^N, \dots, \mathcal{X}_j^N) - (\mathcal{X}_1^N, \dots, \mathcal{X}_j^N)|) \rightarrow 0$
- $\mathcal{D}_\infty(F^N; f) := \mathcal{W}_1(\hat{F}^N, \delta_f) = \mathbb{E}(W_1(\mu_{\mathbb{Z}^N}^N, f)) \rightarrow 0$

More precisely, for $E = \mathbb{R}^d$

Lemma (Hauray, M.)

For given M and $k > 1$ there exist some constants $\alpha_i, C > 0$ such that

$\forall f \in \mathbb{P}(E), \forall F^N \in \mathbb{P}_{\text{sym}}(E^N)$ with $M_k(F_1^N), M_k(f) \leq M$

$$\forall j, \ell \in \{1, \dots, N, \infty\}, \ell \neq 1 \quad \mathcal{D}_j(F^N; f) \leq C (\mathcal{D}_\ell(F^N; f))^{\alpha_1} + \frac{1}{N^{\alpha_2}}.$$

Stronger chaos: entropy and Fisher's chaos

For $F^N \in \mathbb{P}_{\text{sym}}(E^N)$, $E = \mathbb{R}^d$, we define the **normalized** functionals

$$H(F^N) := \frac{1}{N} \int_{E^N} F^N \log F^N, \quad I(F^N) := \frac{1}{N} \int_{E^N} \frac{|\nabla F^N|^2}{F^N}.$$

Definition

Consider a sequence $F^N \in \mathbb{P}_{\text{sym}}(E^N)$ and $f \in \mathbb{P}(E)$

(F^N) is f -entropy chaotic if $F_1^N \rightharpoonup f$ weakly in $\mathbb{P}(E)$ and $H(F^N) \rightarrow H(f)$

(F^N) is f -Fisher's chaotic if $F_1^N \rightharpoonup f$ weakly in $\mathbb{P}(E)$ and $I(F^N) \rightarrow I(f)$

Theorem (Hauray, M.)

In the list below, each assertion implies the one which follows

- (i) (F^N) is Fisher's chaotic;
- (ii) (F^N) is Kac's chaotic and $I(F^N)$ is bounded;
- (iii) (F^N) is entropy chaotic;
- (iv) (F_j^N) converges in L^1 for any $j \geq 1$;
- (v) (F^N) is Kac's chaotic.

Extensions by Carrapatoso, Fournier, Guillin, Hauray, M.

- Kac's chaos, entropic chaos and Fisher's chaos on Kac's spheres and on Boltzmann's spheres
- For a mixture of probability measures = without chaos hypothesis
- Optimal rate of convergence of $\mathcal{D}_\infty(f^{\otimes N}, f) \sim N^{1/d}$ for $f \in \mathbb{P}_q(\mathbb{R}^d)$, $d \geq 2$

Based on many previous works from Funct Analysis, Proba, Stat, Geo, ...

- Mixture: de Finetti (1937), Hewitt-Savage (1955), Robinson-Ruelle (1967)
 - Functional and quantified LLN (Glivenko-Cantelli ... Rachev-Rüschendorf ... Barthe-Bordenave)
 - local central limit theorem of Berry-Esseen
 - HWI inequality of Otto and Villani
 - Entropy inequalities: Carlen-Lieb-Loss (2004), Arstein-Ball-Barthe-Naor (2004)
 - previous comparison, quantitative and qualitative results on chaos
- Kac: Foundations of kinetic theory. (1956)
- Sznitman: Topics in propagation of chaos. Saint-Flour -1989 (1991)
- Carlen, Carvalho, Le Roux, Loss, Villani: Entropy and chaos ... (2010)

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At least four strategies

- BBGKY method (introduced by BBGKY!)
 - ⊕ Quite easy to handle with and robust
 - ⊖ not quantitative, nor qualitative (?), two body interaction only
- semigroups method (introduced by Kac, McKean, Grünbaum)
 - ⊕ quantitative and uniform in time
 - ⊖ rates are not sharp (at all!), maybe not well adapted to singular model
- coupling method (introduced by McKean, popularized by Sznitman)
 - ⊕ best (and sharp) rate, uniform in time for some (“bounded”) model
 - ⊖ non uniform in time for most models, not well adapted to singular model
- nonlinear Martingale method (introduced by Sznitman)
 - ⊕ propagation of chaos results for singular models
 - ⊖ no rate of convergence

For B bounded the Boltzmann-Kac operator is bounded and the semigroup writes

$$F^N(t) = e^{t\Lambda_N} F^N(0) = \sum_{k=1}^{\infty} \frac{t^k}{k!} \Lambda_N^k F^N(0).$$

Because Λ_N is self-adjoint, for any $\varphi \in C_b(E^j)$, $j \leq N$, we have

$$\langle F^N(t), \varphi \rangle = \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle F^N(0), \Lambda_N^k \varphi \rangle.$$

We first consider $\varphi \in C_b(E)$, so that

$$\langle F^N(0), \Lambda_N^k \varphi \rangle \rightarrow \langle f_0^{\otimes k+1}, \varphi_{k+1} \rangle, \quad \varphi_{k+1} = Z_k \varphi \in C(E^{k+1}),$$

and, then assuming $F^N \rightarrow \pi$ as $N \rightarrow \infty$, we get

$$\langle \pi_1(t), \varphi \rangle = \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle f_0^{\otimes k+1}, \varphi_{k+1} \rangle.$$

BBGKY method - Kac's idea

Because Λ_N is self-adjoint, for any $\varphi \in C_b(E^j)$, $j \leq N$, we have

$$\langle F^N(t), \varphi \rangle = \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle F^N(0), \Lambda_N^k \varphi \rangle.$$

We first consider $\varphi \in C_b(E)$, so that

$$\langle F^N(0), \Lambda_N^k \varphi \rangle \rightarrow \langle f_0^{\otimes k+1}, \varphi_{k+1} \rangle, \quad \varphi_{k+1} = Z_k \varphi \in C(E^{k+1}),$$

and, then assuming $F^N \rightarrow \pi$ as $N \rightarrow \infty$, we get

$$\langle \pi_1(t), \varphi \rangle = \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle f_0^{\otimes k+1}, \varphi_{k+1} \rangle.$$

We next consider $\gamma := \varphi \otimes \psi \in C_b(E^2)$, and in the infinite particles limit, we get

$$\langle \pi_2(t), \gamma \rangle = \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle f_0^{\otimes k+1}, \gamma_{k+1} \rangle,$$

with

$$\gamma_{k+1} = \sum_{i=1}^k \varphi_i \psi_{k+1-i} \frac{k!}{i!(k+1-i)!}.$$

We recognize

$$\begin{aligned}\langle \pi_2(t), \gamma \rangle &= \sum_{k,i}^{\infty} \frac{t^{k-i} t^i}{i!(k+1-i)!} \langle f_0^{\otimes i}, \varphi_i \rangle \langle f_0^{\otimes k+1-i}, \psi_{k+1-i} \rangle, \\ &= \langle \pi_1(t) \otimes \pi_1(t), \varphi \otimes \psi \rangle\end{aligned}$$

We conclude that $\pi_2 = \pi_1 \otimes \pi_1$ and

$$\pi_1(t) = \sum_{k=1}^{\infty} \frac{t^k}{k!} Z_k f_0^{\otimes k+1} \quad (\text{Wild Sum} = f(t)).$$

▷ In general we cannot write such an explicit formula and we have to write the all family of equations (for a two body problem)

$$\partial_t F_j^N = (\Lambda_N F^N)_j = \Lambda_{N,j+1} F_{j+1}^N \xrightarrow{N \rightarrow \infty} \partial_t \pi_j = \bar{\Lambda}_{j+1} \pi_{j+1} \quad \forall j \geq 1.$$

▷ biblio: Bogolioubov (?), Born & Green (1946), Kirkwood (1935), Yvon (1935)

Lanford: Time evolution of large classical systems. (1974)

Spohn: On the Vlasov hierarchy (1981)

Arkeryd-Caprino-Ianiri: The homogeneous Boltzmann hierarchy ... (1991)

Gallagher-Saint-Raymond-Texier (2013), Bodineau-Gallagher-Saint-Raymond (2015)

Semigroup method - idea 1 : splitting

We split

$$\begin{aligned}\langle F_{kt}^N - f_t^{\otimes k}, \varphi \rangle &= \langle F_t^N - f_t^{\otimes N}, \varphi \otimes 1^{\otimes N-k} \rangle = \\ &= \langle F_t^N, \varphi \otimes 1^{\otimes N-k} - R_\varphi(\mu_V^N) \rangle \quad (= T_1) \\ &\quad + \langle F_t^N, R_\varphi(\mu_V^N) \rangle - \langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) \rangle \quad (= T_2) \\ &\quad + \langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) \rangle - \langle f_t^{\otimes k}, \varphi \rangle \quad (= T_3)\end{aligned}$$

where R_φ is the “polynomial function” on $\mathbb{P}(\mathbb{R}^3)$ defined by

$$R_\varphi(\rho) = \int_{E^k} \varphi \rho(dv_1) \dots \rho(dv_k)$$

and S_t^{NL} is the nonlinear semigroup associated to the nonlinear mean-field limit equation by $g_0 \mapsto S_t^{NL} g_0 := g_t$.

$$\begin{aligned}
 |T_1| &= \left| \left\langle F_t^N, \varphi \otimes 1^{\otimes(N-k)}(V) - R_\varphi(\mu_V^N) \right\rangle \right| \\
 &= \left| \left\langle F_t^N, \widetilde{\varphi \otimes 1^{\otimes(N-k)}(V)} - R_\varphi(\mu_V^N) \right\rangle \right| \\
 &\leq \left\langle F_t^N, \frac{2k^2}{N} \|\varphi\|_{L^\infty(E^k)} \right\rangle = \mathcal{O}\left(\frac{1}{N}\right)
 \end{aligned}$$

where we use that F^N is symmetric and a probability and we introduce the symmetrization function associated to $\varphi \otimes 1^{\otimes(N-k)}$ by

$$\widetilde{\varphi \otimes 1^{\otimes(N-k)}(V)} = \frac{1}{\#\mathfrak{S}_N} \sum_{\sigma \in \mathfrak{S}_N} \varphi \otimes 1^{\otimes(N-k)}(V_\sigma).$$

$$\begin{aligned}
 |T_3| &= |\langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) - R_\varphi(S_t^{NL} f_0) \rangle| \\
 &\leq [R_\varphi]_{C^{0,1}} \langle F_0^N, W_1(S_t^{NL} \mu_V^N, S_t^{NL} f_0) \rangle \\
 &\leq k \|\nabla \varphi\|_{L^\infty(E^k)} C_T \langle F_0^N, W_1(\mu_V^N, f_0) \rangle \\
 &\leq k \|\nabla \varphi\|_{L^\infty(E)} C_T \mathcal{W}_{W_1}(\hat{F}_0^N, \delta_{f_0}) \\
 &= \mathcal{O}(\mathcal{D}_\infty(F_0^N, f_0)) \quad \text{but in fact} \quad \stackrel{!}{=} \mathcal{O}\left(\frac{1}{\log N}\right)
 \end{aligned}$$

where

$$[R_\varphi]_{C^{0,1}} := \sup_{W_1(\rho, \eta) \leq 1} |R_\varphi(\eta) - R_\varphi(\rho)| = k \|\nabla \varphi\|_{L^\infty}$$

and we have to prove that the nonlinear flow satisfies

$$(A5) \quad W_1(f_t, g_t) \leq C_T W_1(f_0, g_0) \quad \forall f_0, g_0 \in \mathbb{P}(E)$$

T_2 : We write

$$T_2 = \langle F_t^N, R_\varphi(\mu_V^N) \rangle - \langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) \rangle$$

T_2 : We write

$$\begin{aligned} T_2 &= \langle F_t^N, R_\varphi(\mu_V^N) \rangle - \langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) \rangle \\ &= \langle F_0^N, T_t^N(R_\varphi \circ \mu_V^N) - (T_t^\infty R_\varphi)(\mu_V^N) \rangle \end{aligned}$$

with

- T_t^N = dual semigroup (acting on $C_b(E^N)$) of the N-particle flow $F_0^N \mapsto F_t^N$;
- T_t^∞ = pushforward semigroup (acting on $C_b(\mathbb{P}(E))$) of the nonlinear semigroup S_t^{NL} defined by $(T^\infty \Phi)(\rho) := \Phi(S_t^{NL} \rho)$;

T_2 : We write

$$\begin{aligned} T_2 &= \langle F_t^N, R_\varphi(\mu_V^N) \rangle - \langle F_0^N, R_\varphi(S_t^{NL} \mu_V^N) \rangle \\ &= \langle F_0^N, T_t^N(R_\varphi \circ \mu_V^N) - (T_t^\infty R_\varphi)(\mu_V^N) \rangle \\ &= \langle F_0^N, (T_t^N \pi_N - \pi_N T_t^\infty) R_\varphi \rangle \end{aligned}$$

with

- T_t^N = dual semigroup (acting on $C_b(E^N)$) of the N-particle flow $F_0^N \mapsto F_t^N$;
- T_t^∞ = pushforward semigroup (acting on $C_b(\mathbb{P}(E))$) of the nonlinear semigroup S_t^{NL} defined by $(T^\infty \Phi)(\rho) := \Phi(S_t^{NL} \rho)$;
- π_N = projection $C(\mathbb{P}(E)) \rightarrow C(E^N)$ defined by $(\pi_N \Phi)(V) = \Phi(\mu_V^N)$.

Semigroup method - idea 4 : duality + consistency + stability

$$\begin{aligned} T_2 &= \langle F_0^N, (T_t^N \pi_N - \pi_N T_t^\infty) R_\varphi \rangle \\ &= \left\langle F_0^N, \int_0^T T_{t-s}^N (\Lambda^N \pi_N - \pi_N \Lambda^\infty) T_s^\infty ds R_\varphi \right\rangle \\ &= \int_0^T \langle F_{t-s}^N, (\Lambda^N \pi_N - \pi_N \Lambda^\infty) (T_s^\infty R_\varphi) \rangle ds = \mathcal{O}\left(\frac{1}{N^\bullet}\right) \end{aligned}$$

where

- Λ^N is the generator associated to T_t^N and Λ^∞ is the generator associated to T_t^∞ .

Now we have to make some assumptions

- (A1) F_t^N has enough bounded moments;
- (A2) $\Lambda^\infty \Phi(\rho) = \langle Q(\rho), D\Phi(\rho) \rangle$;
- (A3) $(\Lambda^N \pi^N \Phi)(V) = \langle Q(\mu_V^N), D\Phi(\mu_V^N) \rangle + \mathcal{O}([\Phi]_{C^{1,a}}/N)$
- (A4) $S_t^{NL} \in C^{1,a}(\mathbb{P}(E); \mathbb{P}(E))$ “uniformly” in time $t \in [0, T]$

Example of result : Uniform in time propagation of chaos for the hard spheres Boltzmann-Kac model and time relaxation to the equilibrium uniformly in the number of particles

Theorem (M., Mouhot, 2013, a possible answer to Kac's problems)

For any $f_0 \in \mathbb{P}(E)$ + conditions, there exists a sequence $\mathcal{V}^N(0)$ of initial conditions for the Boltzmann-Kac process for hard spheres such that

$$\sup_{t \geq 0} \mathbb{E}(W_1(\mu_{\mathcal{V}^N(t)}^N, f(t))) \leq \frac{C}{\log N}$$

$$H(\mathcal{V}^N(t) | \gamma^N) \rightarrow H(f(t) | \gamma)$$

$$\sup_{N \geq 1} W_1(F^N(t), \gamma^N) \leq \frac{C}{\log t}.$$

▷ biblio: Kac: Foundations of kinetic theory (1956)

McKean: An exponential formula for Boltzmann eq. for a Maxwellian gas (1967)

Grünbaum: Propagation of chaos for the Boltzmann equation (1971)

Kolokoltsov: Nonlinear Markov Processes and Kinetic Equations (book, 2010)

M., Mouhot: Kac's program in kinetic theory (2013)

M., Mouhot, Wennberg: A new approach to quantitative chaos propagation ... (2015)

Carrapatoso: Propagation of chaos for .. Landau eq. for Maxwellian molecules (2016)

Coupling method - idea 1 : synchronous coupling for uniqueness

Consider two solutions to a **smooth coefficients** nonlinear Brownian SDE

$$\begin{aligned}d\mathcal{Z} &= (K * f)(\mathcal{Z}) dt + d\mathcal{B}, & f &= \mathcal{L}(\mathcal{Z}) \\d\bar{\mathcal{Z}} &= (K * \bar{f})(\bar{\mathcal{Z}}) dt + d\mathcal{B}, & \bar{f} &= \mathcal{L}(\bar{\mathcal{Z}}),\end{aligned}$$

with same (synchronous) Brownian motion. We have

$$\begin{aligned}\frac{1}{2} \frac{d}{dt} |\mathcal{Z} - \bar{\mathcal{Z}}|^2 &= [(K * f)(\mathcal{Z}) - (K * f)(\bar{\mathcal{Z}}) + K * (f - \bar{f})(\bar{\mathcal{Z}})](\mathcal{Z} - \bar{\mathcal{Z}}) \\&\lesssim |\mathcal{Z} - \bar{\mathcal{Z}}|^2 + \mathbb{E}(|\mathcal{Z} - \bar{\mathcal{Z}}|^2)\end{aligned}$$

because $\|K * (f - \bar{f})\|_{\infty} \leq \mathbb{E}(|\mathcal{Z} - \bar{\mathcal{Z}}|^2)$. Taking the expectation, we deduce from the Gronwall lemma

$$\mathbb{E}(|\mathcal{Z}_t - \bar{\mathcal{Z}}_t|^2) \leq e^{Lt} \mathbb{E}(|\mathcal{Z}_0 - \bar{\mathcal{Z}}_0|^2).$$

Coupling method - idea 2 : synchronous coupling for chaos estimate

Consider a solution to a N -particle system of Brownian SDE

$$\begin{aligned}d\mathcal{Z}_i &= \frac{1}{N} \sum_{j=1}^N K(\mathcal{Z}_i - \mathcal{Z}_j) dt + d\mathcal{B}_i \\&= (K * \mu_{\mathcal{Z}}^N)(\mathcal{Z}_i) + d\mathcal{B}_i.\end{aligned}$$

Consider a solutions to the associated the nonlinear Brownian SDE

$$\begin{aligned}d\bar{\mathcal{Z}}_i &= (K * \bar{f})(\bar{\mathcal{Z}}_i) dt + d\mathcal{B}_i \\&= (K * \mu_{\bar{\mathcal{Z}}}^N)(\bar{\mathcal{Z}}_i) dt + d\mathcal{B}_i + \{(K * \bar{f})(\bar{\mathcal{Z}}_i) - (K * \mu_{\bar{\mathcal{Z}}}^N)(\bar{\mathcal{Z}}_i)\}\end{aligned}$$

with same (synchronous and independent) Brownian motions.

Using a functional LLN estimate, we similarly get

$$\frac{d}{dt} \frac{1}{N} \sum_{i=1}^N \mathbb{E}(|\mathcal{Z}_i - \bar{\mathcal{Z}}_i|^2) \lesssim \frac{1}{N} \sum_{i=1}^N \mathbb{E}(|\mathcal{Z}_i - \bar{\mathcal{Z}}_i|^2) + \mathcal{O}\left(\frac{1}{N}\right).$$

Coupling method - other ideas : distances, truncation, not synchronous coupling

- Write a differential inequality on an appropriate distance and use Gronwall lemma

$$\mathbb{E}(|Z^N - \bar{Z}^N|^q) \quad \text{or} \quad \mathcal{D}_\infty(\mu_{Z^N}^N, \bar{f}) \sim \mathcal{D}_N(Z^N, \bar{f}) \sim \mathbb{E}(W_2^2(\mu_{Z^N}^N, \bar{f}))^{1/2}$$

- Uniqueness for Boltzmann and Landau equation

$$\frac{d}{dt} W_2^2(f_t, g_t) \leq 0 \quad \text{or even} \quad \leq -W_2^2(f_t, g_t)^{1+\bullet}$$

for Maxwellian molecules ($\gamma = 0$) and then

$$\frac{d}{dt} W_q(f_t, g_t) \lesssim R^\gamma W_q(f_t, g_t) + e^{-R^2}$$

when f_t has exponential moment bounds (but not g_t) for hard potentials ($\gamma > 0$).

- Use a more convenient coupling than the same synchronous coupling or even more than one coupling ...

Theorem (Fournier, Guillin, 2015)

For any $f_0 \in \mathbb{P}(\mathbb{R}^3)$ + conditions, there exists a sequence (\mathcal{V}_0^N) of initial conditions for the Landau-Kac process for hard potential such that

$$\sup_{[0, T]} \mathbb{E}[W_2^2(\mu_{\mathcal{V}^N(t)}^N, f(t))] \leq C_T \left(\mathbb{E}[W_2^2(\mu_{\mathcal{V}_0^N}^N, f_0)] + \frac{1}{N^{1/3}} \right)^{1-\bullet}$$

- ▷ biblio: McKean: Propagation of chaos for a class of non-linear parabolic eq. (1967)
Dobrushin: Vlasov equations (1979)
Tanaka: Probabilistic treatment of Boltzmann eq. for Maxwellian molecules (1978/79)
Sznitman: Topics in propagation of chaos. Saint-Flour -1989 (1991)
Graham-Méléard: Convergence rate for approximations to the Boltzmann eq. (1996)
Malrieu; Convergence to equilibrium for ... and their Euler schemes. (2003)
Fontbona-Guérin-Méléard: Convergence rate for Landau particle systems (2009)
Fournier: Particle approximation of some Landau equations (2009)
Fournier-M.: Rate of convergence for Nanbu particle system (arX 2013)
Cortez-Fontbona: Quantitative propagation of chaos for Kac particle systems (arX 2014)

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Nonlinear Martingale method and the vortex model

Consider a system of N indistinguishable (exchangeable) particles, each particle being described by its position $\mathcal{X}_1^N, \dots, \mathcal{X}_N^N \in \mathbb{R}^2$, which evolves according to

$$d\mathcal{X}_i = \frac{1}{N} \sum_{j=1}^N K(\mathcal{X}_i - \mathcal{X}_j) dt + \sqrt{2\nu} dB_i \quad (\text{Brownian SDE})$$

where $\nu > 0$ is the viscosity and $K : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ is the Biot-Savart kernel defined by

$$\forall x = (x_1, x_2) \in \mathbb{R}^2, \quad K(x) = \frac{x^\perp}{|x|^2} = \left(-\frac{x_2}{|x|^2}, \frac{x_1}{|x|^2} \right) = \nabla^\perp \log |x|,$$

The associated mean field limit is the 2D Navier-Stokes equation written in vorticity formulation

$$\partial_t w_t(x) = (K \star w_t)(x) \cdot \nabla_x w_t(x) + \nu \Delta_x w_t(x), \quad (1)$$

where $w : \mathbb{R}_+ \times \mathbb{R}^2 \rightarrow \mathbb{R}_+$ is the vorticity function

All that can be done for vortices which turn in both (trigonometric and reverse) senses and thus $w : \mathbb{R}_+ \times \mathbb{R}^2 \rightarrow \mathbb{R}$

Theorem (Fournier, Hauray, M., 2014, first version)

- (1) *If \mathcal{X}_0^N is w_0 -Kac's chaotic and "appropriately bounded" then \mathcal{X}_t^N is w_t -Kac's chaotic for any time $t > 0$.*
- (2) *If \mathcal{X}_0^N is w_0 -entropy chaotic and has bounded moment of order $k \in (0, 1]$ then \mathcal{X}_t^N is w_t -entropy chaotic for any time $t > 0$.*

▷ biblio: Sznitman: Equations de type de Boltzmann, spatialement homogenes. (1984)
Osada: Propagation of chaos for the 2D Navier-Stokes equation (1985)–(1987)
Fournier-Hauray: Chaos propag for Landau eq with moderate soft potentials (arX 2015)
Fournier-Jourdain. Stochastic particle approximation of Keller-Segel equation (arX 2015)

We say that $\mathcal{X} = (\mathcal{X}_t)_{t \geq 0}$ a continuous stochastic process with values in \mathbb{R}^2 is a solution to the stochastic NS vortex equation if it satisfies the nonlinear Brownian SDE

$$d\mathcal{X}_t = (K * w_t)(\mathcal{X}_t) + \sqrt{2\nu} dB_t$$

for some given brownian motion B and where $w_t = \mathcal{L}(\mathcal{X}_t)$ is the law of \mathcal{X}_t .

It is important to point out that (thanks to Ito formula) the law w_t of \mathcal{X}_t then satisfies the NS vortex equation

$$\partial_t w_t = (K * w_t) \cdot \nabla_x w_t + \nu \Delta_x w_t.$$

Theorem (Fournier, Hauray, M., 2014, second version)

Consider $w_0 \geq 0$ a function such that

$$\int_{\mathbb{R}^2} w_0 (1 + |x|^k + |\log w_0|) dx < \infty, \quad k \in (0, 1],$$

the vortices trajectories $\mathcal{X}^N = (\mathcal{X}_t^N)_{t \geq 0}$ associated to an i.c. $\mathcal{X}_0^N \sim w_0^{\otimes N}$ and \mathcal{X} the solution to the stochastic NS vortex equation associated to an i.c. $\mathcal{X}_0 \sim w_0$. There holds

$$\mu_{\mathcal{X}^N}^N \Rightarrow \mathcal{X} \text{ in law in } \mathbb{P}(C([0, \infty); \mathbb{R}^2)) \text{ as } N \rightarrow \infty$$
$$\mathcal{L}(\mathcal{X}_1^N(t)) \rightarrow w_t = \mathcal{L}(\mathcal{X}_t) \text{ strongly in } L^1(\mathbb{R}^2) \text{ as } N \rightarrow \infty$$

The first convergence means

$$\mathcal{L}(\mu_{\mathcal{X}^N}^N) \rightharpoonup \delta_{\mathcal{L}(\mathcal{X})} \text{ weakly in } \mathbb{P}(\mathbb{P}(C([0, \infty); \mathbb{R}^2))) \text{ as } N \rightarrow \infty$$

and the second can be improved into

$$\mathcal{L}(\mathcal{X}_1^N(t), \dots, \mathcal{X}_j^N(t)) \rightarrow w_t^{\otimes j} \text{ strongly in } L^1(\mathbb{R}^2)^j \text{ as } N \rightarrow \infty$$

The proof follows the by-now well-known “weak stability on nonlinear martingales” approach, which goes back to Sznitman 1984.

Everything is standard except the fact that we use the Fisher information bound in each step.

- A priori estimates (on entropy, moment and Fisher information)
- tightness of the law Q^N of the empirical process $\mu_{\mathcal{X}^N}^N$ in $\mathbb{P}(\mathbb{P}(E))$
- pass to the limit and identify the set of limit points \mathcal{S} as the probability measures $q \in \mathbb{P}(E)$ associated to a process \mathcal{X} which solves the (Martingale problem associated to the) stochastic NS vortex equation and has finite Fisher information.
- if $q \in \mathcal{S}$ and $q = \mathcal{L}(\mathcal{X})$ then $w_t := \mathcal{L}(\mathcal{X}_t)$ is the **unique** solution to the NS vortex PDE
- the Martingale problem has a **unique** solution $\bar{\mathcal{X}}$ and then $\mathcal{S} = \{\bar{q}\}$ where $\mathcal{L}(\bar{q}) = \bar{\mathcal{X}}$.

In conclusion, $Q^N \rightharpoonup \delta_{\bar{q}}$ in $\mathbb{P}(\mathbb{P}(E))$. (that Kac's chaos)

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Using $\operatorname{div} K = 0$, we get the entropy identity

$$H(F_t^N) + \nu \int_0^t I(F_s^N) ds = H(F_0^N).$$

As usually we need a control of a moment of F_t^N in order to take advantage of the entropy bound (we need a lower bound on H).

We define the moment M_k of order $k \in (0, 1]$ by

$$M_k(F^N) = \int_{\mathbb{R}^{2N}} F^N \frac{1}{N} \sum_{j=1}^N \langle x_j \rangle^k = \int_{\mathbb{R}^2} F_1^N \langle x \rangle^k dx$$

We then compute

$$\begin{aligned} \frac{d}{dt} M_k(F_t^N) &= \nu \int_{\mathbb{R}^2} F_{1t}^N \Delta \langle x \rangle^k + \int_{\mathbb{R}^4} F_{2t}^N K(x_1 - x_2) \cdot \nabla_1 \langle x_1 \rangle^k \\ &\leq C_1 \int_{\mathbb{R}^2} F_{1t}^N + C_2 \int_{\mathbb{R}^4} F_{2t}^N \frac{1}{|x_1 - x_2|} \end{aligned}$$

Control given by the Fisher information

Define $g^N := \mathcal{L}(X_2 - X_1)$ and use (Carlen 1991) Fisher information inequalities

$$\frac{1}{2} I_1(g^N) \leq I_2(F_2^N) \leq I_N(F^N)$$

as well as Gagliardo-Nirenberg type inequalities in 2D

$$\forall g \in \mathbb{P}(\mathbb{R}^2), \forall p \in [1, \infty) \quad \|g\|_{L^p} \leq C_p I(g)^{1-1/p}.$$

Coming back to the singular term in the moment equation, we compute

$$\begin{aligned} \int_{\mathbb{R}^4} \frac{F_{2t}^N}{|x_1 - x_2|} dx_1 dx_2 &= \sqrt{2} \int_{B_1} \frac{g_t^N(x)}{|x|} dx + \sqrt{2} \int_{B_1^c} \frac{g_t^N(x)}{|x|} dx \\ &\leq \sqrt{2} \| |\cdot|^{-1} \|_{L^{3/2}(B_1)} \|g_t^N\|_{L^3(B_1)} + \sqrt{2} \|g_t^N\|_{L^1(B_1^c)} \\ &\leq C_3 I(g_t^N)^{2/3} + C_4 \\ &\leq \frac{\nu}{4C_2} I(g_t^N) + C_5 \\ &\leq \frac{\nu}{2C_2} I(F_t^N) + C_5 \end{aligned}$$

Summing up the two equations on the entropy and on the moment of order k , we find

Lemma (a priori estimates)

Uniformly in N

$$\begin{aligned} H(F_t^N) + M_k(F_t^N) + \frac{\nu}{2} \int_0^t I(F_s^N) ds \\ \leq H(F_0^N) + M_k(F_0^N) + (C_1 + C_2)t \end{aligned}$$

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We denote

$\mathcal{X}^N := (\mathcal{X}_1^N, \dots, \mathcal{X}_N^N)$ the **exchangeable** r.v. with value in E^N

where $\mathcal{X}_i^N = (\mathcal{X}_i^N(t))_{t \geq 0} \in E := C([0, \infty); \mathbb{R}^2)$ solution to the SDE

$$\mathcal{X}_i(t) = \mathcal{X}_i(0) + \int_0^t (K * \mu_{\mathcal{X}(s)}^N)(\mathcal{X}_i(s)) ds + \sqrt{2\nu} B_i(t)$$

and we want to show that each particle behaves asymptotically like N independent copies of the same process $\mathcal{X} = (\mathcal{X}(t))_{t \geq 0}$ defined as the solution to the nonlinear SDE

$$\mathcal{X}(t) = \mathcal{X}(0) + \int_0^t (K * w_s)(\mathcal{X}(s)) ds + \sqrt{2\nu} B(t),$$

where $w_s := \mathcal{L}(\mathcal{X}(s))$ and then is a solution (Ito formula) to the NS vortex equation

$$\partial_t w = (K \star w) \cdot \nabla_x w + \nu \Delta_x w.$$

Lemma

the family of laws $\mathcal{L}(\mu_{\mathcal{X}_N}^N)_{N \geq 1}$ is tight in $\mathbb{P}(\mathbb{P}(E))$

From classical compactness criterium (Sznitman 1984) it is enough to prove that the family of laws $\mathcal{L}(\mathcal{X}_1^N)_{N \geq 1}$ is tight in $\mathbb{P}(E)$. That is a consequence of

Lemma

For all $T > 0$, $\theta \in (0, 1/2)$

$$\mathbb{E} \left[\sup_{0 < s < t < T} \frac{|\mathcal{X}_1^N(t) - \mathcal{X}_1^N(s)|}{(t-s)^\theta} \right] \leq C \left(1 + \int_0^T I(G_u^N) du \right)$$

By Prokhorov, we get

Lemma

There exists $Q \in \mathbb{P}(\mathbb{P}(E))$ such that

$$Q^N \rightharpoonup Q \text{ in } \mathbb{P}(\mathbb{P}(E)).$$

Identification of the the limit thanks to “Sznitman” argument.

Lemma

Assume that $Q \in \mathbb{P}(\mathbb{P}(C([0, +\infty), \mathbb{R}^2))) = \mathbb{P}(\mathbb{P}(E))$ is a mixture measure obtained as a limit point of some subsequence of Q^N . Then $\text{supp } Q \subset \mathcal{S}$

$$\mathcal{S} := \left\{ \begin{array}{l} q \text{ is the law of some } \mathcal{X} \text{ solution to stoch. NS vortex eq.} \\ \forall T > 0, \quad \int_0^T I(\mathcal{L}(X_s)) ds < +\infty \end{array} \right\} = \mathcal{S}_0 \cap \mathcal{S}_1$$

- $q \approx \mathcal{X}$ solves the stoch. NS vortex eq. iif for all times $s, t, \psi, \varphi \dots$

$$\begin{aligned} \mathcal{F}(q) := \iint_{E^2} q(dx)q(dy) \psi(x(s \leq t)) & \left[\varphi(x(t)) - \varphi(x(s)) \right. \\ & \left. - \int_s^t K(x(u) - y(u)) \cdot \nabla \varphi(x(u)) du - \nu \int_s^t \Delta \varphi(x(u)) du \right] = 0 \end{aligned}$$

- Q concentrated on $\mathcal{S}_0 \iff \mathbb{E}_Q[|\mathcal{F}(\cdot)|^2] = 0$ for all s, t, ψ, φ .
- $\mathbb{E}_{Q^N}[|\mathcal{F}(\cdot)|^2] \rightarrow 0$ as $N \rightarrow +\infty$.
- Continuity $\mathbb{P}(\mathbb{P}(E)) \ni R \mapsto \mathbb{E}_R[|\mathcal{F}(\cdot)|^2]$ under the condition $\mathbb{E}_R[\int_0^t I(\cdot_s) ds] < +\infty$.
- $\mathbb{E}_P[\int_0^t I(\cdot_s) ds] < +\infty$, which is equivalently $P \in \mathcal{S}_1$.

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Level 3 Fisher information (for a mixture of probability measures)

Consider $\pi \in \mathbb{P}(\mathbb{P}(E))$, $E = \mathbb{R}^2$, and define

$$\mathcal{I}(\pi) := \int_{\mathbb{P}(E)} I(\rho) \pi(d\rho), \quad \mathcal{I}'(\pi) := \sup_{j \geq 1} I(\pi_j) = \lim_{j \rightarrow \infty} I(\pi_j)$$

where π_j is given by (the easy part of) Hewitt and Savage theorem

$$\pi_j := \int_{\mathbb{P}(E)} \rho^{\otimes j} \pi(d\rho) \in \mathbb{P}_{\text{sym}}(E^j).$$

From $I(f^{\otimes j}) = I(f)$ (good normalization), I is lsc, convex, proper and ≥ 0 on $\mathbb{P}_{\text{sym}}(E^j)$, $\forall j \geq 1$, and \mathcal{I}' is linear on disjoint convex combination, we deduce

Theorem (representation formula, Hauray-M.)

$$\forall \pi \in \mathbb{P}(\mathbb{P}(E)) \quad \mathcal{I}(\pi) = \mathcal{I}'(\pi).$$

A similar formula is known for the entropy (Robinson-Ruelle, 1967)

Application: the Fisher information is Γ -lsc in the sense

$$\mathbb{P}_{\text{sym}}(E^N) \ni F^N \rightharpoonup \pi \in \mathbb{P}(\mathbb{P}(E)) \text{ implies } \mathcal{I}(\pi) \leq \liminf I(F^N).$$

One line proof: for any $j \geq 1$ by lsc of I_j : $I_j(\pi_j) \leq \liminf I_j(F_j^N) \leq \liminf I_N(F^N)$.

Consequence for the vortex problem

We know (from tightness) that

$$\mathcal{L}(\mu_{\mathcal{X}}^N) \rightharpoonup Q \quad \text{weakly in } \mathbb{P}(\mathbb{P}(E))$$

with here $E := C([0, \infty); \mathbb{R}^2)$. We define $Q_t :=$ projection on the section $\mathbb{P}(\mathbb{P}(\{t\} \times \mathbb{R}^2))$ so that

$$G_t^N = \mathcal{L}(\mathcal{X}_t^N), \mathcal{L}(\mu_{\mathcal{X}_t^N}^N) \rightharpoonup Q_t \quad \text{weakly in } \mathbb{P}(\mathbb{P}(\mathbb{R}^2))$$

As a consequence, by Fubini, Γ -lsc property of the Fisher information and Fatou

$$\begin{aligned} \int_{\mathbb{P}(E)} \int_0^T I(q_t) dt Q(dq) &= \int_0^T \int_{\mathbb{P}(E)} I(q_t) Q(dq) dt \\ &= \int_0^T \mathcal{I}(Q_t) dt \\ &\leq \int_0^T \liminf_N I(G_t^N) dt \leq \liminf_N \int_0^T I(G_t^N) dt. \end{aligned}$$

This last quantity is finite, so that $\int_0^T I(q_t) dt < \infty$ Q -a.s. and $\text{supp } Q \subset \mathcal{S}_1$.

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Uniqueness of the solution to the NS vortex equation

We claim that

$\forall q \in \mathcal{S}, q = \mathcal{L}(\mathcal{X}), \quad w_t := \mathcal{L}(\mathcal{X}_t) = \bar{w}_t :=$ unique solution of NS vortex.

- First, for $q \in \mathcal{S}$, it is clear that $w_t := \mathcal{L}(\mathcal{X}_t)$ satisfies

$$w \in C([0, T]; \mathbb{P}(R^2)), \quad I(w) \in L^1(0, T)$$

and w is a weak solution to (take $\nu = 1$)

$$\partial_t w = \Delta w + (K * w) \cdot \nabla w.$$

- Second, the uniqueness is known (Ben-Artzi 1994, Brézis 1994, improved by Gallagher-Gallay 2005) in the class of function

$$t^{1/4} \|w(t, \cdot)\|_{L^{4/3}} \rightarrow 0 \text{ as } t \rightarrow 0.$$

- We have to prove by a “regularity argument” (through a renormalization trick) that w satisfies the Ben-Artzi & Brézis criterium

Uniqueness (in law) of linear SDE under the a priori condition.

If $q \in \mathcal{S}$ we consider the associated linear SDE

$$\mathcal{X}_t = \mathcal{X}_0 + \int_0^t u_s(\mathcal{X}_s) ds + \nu B_t, \quad u_s = K * \bar{w}_s,$$

Lemma

Strong uniqueness for the previous linear SDE holds (and thus weak uniqueness by Yamada-Watanabe theorem). In other words, $\mathcal{S} = \{\bar{q}\}$.

Sketch of the proof

- Use argument used by Crippa-De Lellis for uniqueness in ODE with low regularity.
- Two solutions \mathcal{X} and \mathcal{Y} satisfies

$$\forall \delta > 0, \quad \mathbb{E} \left[\ln \left(1 + \frac{1}{\delta} \sup_{s \leq t} |\mathcal{X}_s - \mathcal{Y}_s| \right) \right] \leq \mathbb{E} \left[\int_0^t [M \nabla u_s(\mathcal{X}_s) + M \nabla u_s(\mathcal{Y}_s)] ds \right]$$

where M stands for maximal function.

- Standard estimates + bounds on Fischer information helps to bound the r.h.s.
- A variant of Chebichev ineq. allows to conclude.

Plan

- 1 Introduction
- 2 Short discussion about chaos
- 3 Short discussion about methods
- 4 Nonlinear Martingale method and the vortex model
- 5 sketch of the proof - a priori estimates
- 6 sketch of the proof - probability argument
- 7 Sketch of the proof - functional analysis argument
- 8 Sketch of the proof - PDE/SDE argument
- 9 Sketch of the proof - entropy argument

From

$$H(F_t^N) + \int_0^t I(F_s^N) ds = H(F_0^N)$$

and

$$H(w_t) + \int_0^t I(w_s) ds = H(w_0),$$

as well as the Γ -lsc of H and I we get if

$$H(F_0^N) \rightarrow H(w_0),$$

the conclusion

$$\begin{aligned} H(w_t) + \int_0^t I(w_s) ds &\leq \liminf_{N \rightarrow \infty} \left\{ H(F_t^N) + \int_0^t I(F_s^N) ds \right\} \\ &\leq \limsup_{N \rightarrow \infty} \left\{ H(F_t^N) + \int_0^t I(F_s^N) ds \right\} \\ &= \limsup_{N \rightarrow \infty} H(F_0^N) = H(w_0) \end{aligned}$$

and then

$$H(F_t^N) \rightarrow H(w_t) \quad \forall t > 0$$