

Uniform in time propagation of chaos for the generalized Dyson Brownian motion and 1D Riesz gases.

Pierre Le Bris

Joint work with : Arnaud Guillin (LMBP), Pierre Monmarché (LJLL)

LJLL, Sorbonne Université - Paris

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The model

1D N-particle system in mean field interaction

$$dX_t^i = \sqrt{2\sigma_N} dB_t^i - U'(X_t^i) dt - \frac{1}{N} \sum_{j \neq i} V'(X_t^i - X_t^j) dt,$$

where

- σ_N diffusion coefficient,
- $(B^i)_i$ independent Brownian motions,
- U confining potential such that either U' is Lipschitz continuous or $U'(x) = \lambda x$,
- $\exists \alpha \geq 0, \forall x \in \mathbb{R}^*, V'(x) = -\frac{x}{|x|^{\alpha+1}}$.

Motivation

The (generalized) Dyson Brownian motion

$$dX_t^i = \sqrt{\frac{2\sigma}{N}} dB_t^i - \lambda X_t^i dt + \frac{1}{N} \sum_{j \neq i} \frac{1}{X_t^i - X_t^j} dt.$$

Motivation

The (generalized) Dyson Brownian motion

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Question : What happens when $N \rightarrow \infty$?

Finding the limit

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Finding the limit

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Formally, notice $\frac{1}{N} \sum_{j=1}^N V'(X_t^i - X_t^j) = V' * \mu_t^N(X_t^i)$, where
 $\mu_t^N := \frac{1}{N} \sum_{i=1}^N \delta_{X_t^i}$.

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$$\begin{cases} dX_t = \sqrt{2\sigma} dB_t - U'(X_t) dt - V' * \bar{\rho}_t(X_t) dt, \\ \bar{\rho}_t = \text{Law}(X_t), \end{cases}$$

Finding the limit

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$$\begin{cases} dX_t = \sqrt{2\sigma} dB_t - U'(X_t) dt - V' * \bar{\rho}_t(X_t) dt, \\ \bar{\rho}_t = \text{Law}(X_t), \end{cases}$$

which is linked to

$$\partial_t \bar{\rho}_t = \partial_x \left((U' + V' * \bar{\rho}_t) \bar{\rho}_t \right) + \sigma \partial_{xx}^2 \bar{\rho}_t.$$

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Goal : Show $\mu_t^N \rightarrow \bar{\rho}_t$.

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Some methods :

- Coupling methods (*McKean, Sznitman, Eberle...*) :

$$\mathcal{W}_2(\mu, \nu)^2 = \inf_{X \sim \mu, Y \sim \nu} \mathbb{E}(|X - Y|^2).$$

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Let $\rho_t^N = \text{Law}(X_t^1, \dots, X_t^N)$, show $\mathcal{W}_2(\rho_t^N, \bar{\rho}_t^{\otimes N}) \rightarrow 0$.

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- Energy/Entropy estimates (*Serfaty, Jabin, Wang...*).

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Goal : Show $\mu_t^N \rightarrow \bar{\rho}_t$.

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- Coupling methods (*McKean, Sznitman, Eberle...*) :

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- Energy/Entropy estimates (*Serfaty, Jabin, Wang...*).
- Tightness (*Rogers, Zhi, Cépa, Lépingle...*).

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Existence, uniqueness, no collisions

Theorem

Consider $N \geq 2$, and $-\infty < x_1 < \dots < x_N < \infty$.

- If $\alpha > 1$, for any $\sigma_N \geq 0$, there exists a unique strong solution $X = (X^1, \dots, X^N)$ to the particle system with initial condition $X_0^1 = x_1, \dots, X_0^N = x_N$, which furthermore satisfies $X_t^1 < \dots < X_t^N$ for all $t \geq 0$, \mathbb{P} -a.s.
- The same result holds for $\alpha = 1$ and $\sigma_N \leq \frac{1}{N}$.

"Cauchy sequence"

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Lemma

Let $(\mu_t^N)_{N \in \mathbb{N}}$ be any sequence of independent empirical measures, such that μ_t^N is the empirical measure of the N particle system at time t . Then (for $\lambda > 0$, $\alpha = 1$, $U'(x) = \lambda x$ and $\sigma_N = \frac{1}{N}$), we have for all $t \geq 0$ and all $N, M \geq 1$

$$\mathbb{E} \left(\mathcal{W}_2 \left(\mu_t^N, \mu_t^M \right)^2 \right) \leq e^{-2\lambda t} \mathbb{E} \left(\mathcal{W}_2 \left(\mu_0^N, \mu_0^M \right)^2 \right) + \frac{C}{N \wedge M}.$$

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Remark : The same result holds...

- for $U = 0$, but no longer uniform in time,
- for $\alpha \in [1, 2[$, with rate $N^{-\frac{2-\alpha}{\alpha}}$,

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Remark : The same result holds...

- for $U = 0$, but no longer uniform in time,
- for $\alpha \in [1, 2[$, with rate $N^{-\frac{2-\alpha}{\alpha}}$,
- for U' only Lipschitz continuous, but no longer uniform in time,

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Lemma

Let $(\mu_t^N)_{N \in \mathbb{N}}$ be any sequence of independent empirical measures, such that μ_t^N is the empirical measure of the N particle system at time t . Then (for $\lambda > 0$, $\alpha = 1$, $U'(x) = \lambda x$ and $\sigma_N = \frac{1}{N}$), we have for all $t \geq 0$ and all $N, M \geq 1$

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Remark : The same result holds...

- for $U = 0$, but no longer uniform in time,
- for $\alpha \in [1, 2[$, with rate $N^{-\frac{2-\alpha}{\alpha}}$,
- for U' only Lipschitz continuous, but no longer uniform in time,
- for the supremum, but no longer uniform in time.

Conclusion

Using independence, this implies that there exists a (deterministic) $\bar{\rho}_t$ such that

$$\mathbb{E} \left(\mathcal{W}_2(\mu_t^N, \bar{\rho}_t)^2 \right) \rightarrow 0.$$

Proof of the estimate

For two sets of points $(x_i)_{i \in \{1, \dots, N\}}$ and $(y_j)_{j \in \{1, \dots, N\}}$, with $x_1 \leq \dots \leq x_N$ and $y_1 \leq \dots \leq y_N$, and two measures $\mu = \frac{1}{N} \sum_i \delta_{x_i}$ and $\nu = \frac{1}{N} \sum_j \delta_{y_j}$:

$$\mathcal{W}_2(\mu, \nu)^2 = \frac{1}{N} \sum_i |x_i - y_i|^2.$$

Proof of the estimate

For two sets of points $(x_i)_{i \in \{1, \dots, N\}}$ and $(y_j)_{j \in \{1, \dots, N\}}$, with $x_1 \leq \dots \leq x_N$ and $y_1 \leq \dots \leq y_N$, and two measures $\mu = \frac{1}{N} \sum_i \delta_{x_i}$ and $\nu = \frac{1}{N} \sum_j \delta_{y_j}$:

$$\mathcal{W}_2(\mu, \nu)^2 = \frac{1}{N} \sum_i |x_i - y_i|^2.$$

Let

$$-\infty < X_t^1 = \dots = X_t^N < \dots < X_t^{N(M-1)+1} = \dots = X_t^{NM} < \infty$$

$$-\infty < Y_t^1 = \dots = Y_t^M < \dots < Y_t^{M(N-1)+1} = \dots = Y_t^{NM} < \infty.$$

Thus

$$\mu_t^M = \frac{1}{M} \sum_{i=1}^M \delta_{\bar{x}_t^{i,M}} = \frac{1}{NM} \sum_{i=1}^{NM} \delta_{X_t^i} \quad \text{and} \quad \mu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{\bar{y}_t^{i,N}} = \frac{1}{NM} \sum_{i=1}^{NM} \delta_{Y_t^i},$$

and

$$\mathcal{W}_2(\mu_t^N, \mu_t^M)^2 = \frac{1}{NM} \sum_{i=1}^{NM} |X_t^i - Y_t^i|^2.$$

Proof of the estimate-2

Direct calculations yield :

$$\begin{aligned} & d \left(\mathcal{W}_2 \left(\mu_t^N, \mu_t^M \right)^2 \right) \\ &= -2\lambda \mathcal{W}_2 \left(\mu_t^N, \mu_t^M \right)^2 dt + 2\sigma \left(\frac{1}{N} + \frac{1}{M} \right) dt + dM_t \\ &\quad - \frac{2}{(NM)^2} \sum_i \left(X_t^i - Y_t^i \right) \sum_j \left(V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j) \right) dt. \end{aligned}$$

Proof of the estimate-3

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$$\begin{aligned} & \sum_i (X_t^i - Y_t^i) \sum_j (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) \\ &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - Y_t^i) - (X_t^j - Y_t^j)) \\ &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - X_t^j) - (Y_t^i - Y_t^j)). \end{aligned}$$

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$$\begin{aligned}
 & \sum_i (X_t^i - Y_t^i) \sum_j (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) \\
 &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - Y_t^i) - (X_t^j - Y_t^j)) \\
 &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - X_t^j) - (Y_t^i - Y_t^j)). \\
 &\geq \sum_{i>j \text{ s.t. } Y_t^i = Y_t^j} V'(X_t^i - X_t^j) (X_t^i - X_t^j) + \sum_{i>j \text{ s.t. } X_t^i = X_t^j} V'(Y_t^i - Y_t^j) (Y_t^i - Y_t^j) \\
 &\geq \sum_{i>j \text{ s.t. } Y_t^i = Y_t^j} -1 + \sum_{i>j \text{ s.t. } X_t^i = X_t^j} -1 \\
 &= -\frac{M(M-1)}{2}N - \frac{N(N-1)}{2}M.
 \end{aligned}$$

Proof of the estimate-3

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$$\begin{aligned}
 & \sum_i (X_t^i - Y_t^i) \sum_j (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) \\
 &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - Y_t^i) - (X_t^j - Y_t^j)) \\
 &= \sum_{i>j} (V'(X_t^i - X_t^j) - V'(Y_t^i - Y_t^j)) ((X_t^i - X_t^j) - (Y_t^i - Y_t^j)). \\
 &\geq \sum_{i>j \text{ s.t. } Y_t^i = Y_t^j} V'(X_t^i - X_t^j) (X_t^i - X_t^j) + \sum_{i>j \text{ s.t. } X_t^i = X_t^j} V'(Y_t^i - Y_t^j) (Y_t^i - Y_t^j) \\
 &\geq \sum_{i>j \text{ s.t. } Y_t^i = Y_t^j} -1 + \sum_{i>j \text{ s.t. } X_t^i = X_t^j} -1 \\
 &= -\frac{M(M-1)}{2}N - \frac{N(N-1)}{2}M.
 \end{aligned}$$

Hence

$$\mathbb{E} \left(\mathcal{W}_2 \left(\mu_t^N, \mu_t^M \right)^2 \right) \leq e^{-2\lambda t} \mathbb{E} \left(\mathcal{W}_2 \left(\mu_0^N, \mu_0^M \right)^2 \right) + \frac{C}{N \wedge M}.$$

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