

Mean-field limit of (non)-exchangeable multi-agent systems

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Goal:

Mean-field limit for multi-agent systems

⇒ Singular (Coulomb) interaction agents

Major open problem

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⇒ Non-exchangeable agents: Remove

Type of cooperations (symmetry)

Connectivities of dense graphs

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Connectivities of dense graphs

Results:

Novel hierarchy of observables

New L^p estimates of the marginals of the system

Allowing very singular interaction kernels

Non-exchangeable agents on sparse-graphs

→ New concept of limits: extended graphons

A general class of N-agents interacting system:

$$dX_i = \sum_{j=1}^N w_{ij}^N K(X_i - X_j) dt + \sqrt{2\sigma} dW_i$$
$$X_i(0) = X_i^0 \in \mathbb{R}^d$$

where

- X_i positions/activities
- w_{ij} weights/connectivities
- W_i N independent Wiener processes
- K interaction kernel

Classical mean-field theory:

$$dX_i = \frac{1}{N} \sum_{j=1}^N K(X_i - X_j) dt + \sqrt{2\sigma} dW_i$$
$$X_i(0) = X_i^0 \in \mathbb{R}^d$$

$$w_{ij} \sim \frac{1}{N}$$

K

exchangeable particles

interaction kernel



$$\frac{\partial f}{\partial t} + \operatorname{div} (f(K * f) - \sigma \nabla f) = 0$$

Classical mean-field theory: **Second order systems**

$$\frac{d}{dt}X_i(t) = V_i(t)$$

$$dV_i(t) = \frac{1}{N} \sum_{j \neq i} K(X_i - X_j) dt + \sqrt{2\sigma} dW_i$$

Rigorous limit \downarrow $N \rightarrow \infty$

$$\frac{\partial f}{\partial t} + v \cdot \nabla_x f + K * \rho \cdot \nabla_v f = \sigma \Delta_v f$$

Vlasov-Fokker-Planck equation

distribution: $f(t, x, v)$

density: $\rho(t, x) = \int f(t, x, v) dv$

Classical mean-field theory: Bibliographic background

$K \in W^{1,\infty}$ + f is smooth enough

McKean 1967, Braun-Hepp 1977, Dobrushin 1979, ...

Neunzert, Sznitmann, ..., Golse

Estimates on the trajectories distance one by one

$$\left(\frac{1}{N} \mathbb{E} \left(\sum_{i=1}^N |(X_i, V_i) - (\bar{X}_i, \bar{V}_i)|^p \right) \right)^{\frac{1}{p}}$$

where (\bar{X}_i, \bar{V}_i) are N identical copies of

$$\frac{d}{dt} \bar{X}(t) = \bar{V}(t)$$

$$d\bar{V}(t) = \left(K \star \int f(t, \bar{X}, v) dv \right) dt + \sqrt{2\sigma} dW$$

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$$\|\mu_N - f\|_{W^{-1,1}} \leq c e^{t \|\nabla K\|_{L^\infty}} \|\mu_N^0 - f^0\|_{W^{-1,1}}$$

$$\mu_N(t, x, v) := \frac{1}{N} \sum_{i=1}^N \delta_{X_i(t)}(x) \delta_{V_i(t)}(v)$$

(empirical measure)

\implies This trajectorial approach requires $K \in W^{1,\infty}$

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(empirical measure)

$$|K| \leq \frac{c}{|x|^\theta}, \theta < 1$$

Hauray-Jabin 2015

→ only in the deterministic case

→ in the stochastic case

$$d = 1$$

Hauray-Salem 2019

→ in the stochastic case

$d = 1$

Hauray-Salem 2019

$d > 1$: Truncated kernels

Huang-Liu-Pickl 2020, Pickl *et al.* $\theta < \frac{1}{d}$, $d = 3$

Typically $\phi_\varepsilon = \frac{C}{(\varepsilon_N + |x|)^{d-2}}$, $K_\varepsilon = -\nabla\phi_\varepsilon$, $d \geq 3$

where

$$\varepsilon_N = N^{-\theta}$$

with the critical scale

$$\theta = \frac{1}{d}$$

Interaction kernel:

$$K = -\nabla\phi, \text{ for a repulsive and nonnegative } \phi$$

Main example:

Coulombian interactions

$$\rightarrow \phi = -c \ln|x|, \quad \text{if } d = 2$$

$$\rightarrow \phi = \frac{c}{|x|^{d-2}}, \quad \text{if } d \geq 3$$

For simplicity:

$$X_i \in \mathbb{T}^d$$

$$V_i \in \mathbb{R}^d$$

Classical mean-field theory: Statistical approach

- **New object:** We put individual trajectories aside

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- **New object:** We put individual trajectories aside
- **Full joint law** at time t : $f_N(t, x_1, v_1, \dots, x_N, v_N)$ satisfies the Liouville equation:

$$\begin{aligned} \partial_t f_N + \sum_{i=1}^N v_i \cdot \nabla_{x_i} f_N \\ + \sum_{i=1}^N \frac{1}{N} \sum_{j=1}^N K(x_i - x_j) \cdot \nabla_{v_i} f_N = \sigma \sum_i \Delta_{v_i} f_N, \end{aligned}$$

We need velocity decay to control ρ_N

Statistical approach

- Introduce a joint law of the “trajectories” as a kind of projection: k-point marginals
- **Marginal laws** of $X_1, V_1, \dots, X_k, V_k$ at time t :

$$f_{k,N}(t, x_1, v_1, \dots, x_k, v_k) = \int_{\mathbb{T}^{d(N-k)} \times \mathbb{R}^{d(N-k)}} f_N(t, x_1, v_1, \dots, x_N, v_N) dx_{k+1} dv_{k+1} \dots dx_N dv_N.$$

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- **Question:** $f_{k,N} \rightharpoonup f^{\otimes k}$?

$$\begin{aligned}
& \partial_t f_{k,N} + \sum_{i=1}^k v_i \cdot \nabla_{x_i} f_{k,N} + \sum_{i \leq k} \frac{1}{N} \sum_{j \leq k} K(x_i - x_j) \cdot \nabla_{v_i} f_{k,N} \\
& + \frac{N-k}{N} \sum_{i \leq k} \nabla_{v_i} \cdot \int_{\mathbb{T}^d \times \mathbb{R}^d} f_{k+1,N} K(x_i - x_{k+1}) dx_{k+1} dv_{k+1} \\
& = \sigma \sum_{i \leq k} \Delta_{v_i} f_{k,N}.
\end{aligned}$$

Critical new idea

the new term has the same scaling as
the convolution at the limit

Moment propagation:

$$e^{\sum_{i \leq k} (1 + |v_i|^2)}$$

$$\begin{aligned}
& \partial_t f_{k,N} + \sum_{i=1}^k v_i \cdot \nabla_{x_i} f_{k,N} + \sum_{i \leq k} \frac{1}{N} \sum_{j \leq k} K(x_i - x_j) \cdot \nabla_{v_i} f_{k,N} \\
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$$e^{e_k(x_1, v_1, \dots, x_k, v_k)} = e^{\sum_{i \leq k} (1 + |v_i|^2) + \frac{1}{N} \sum_{i, j \leq k} \phi(x_i - x_j)}.$$

$$\begin{aligned}
& \partial_t f_{k,N} + \sum_{i=1}^k v_i \cdot \nabla_{x_i} f_{k,N} + \sum_{i \leq k} \frac{1}{N} \sum_{j \leq k} K(x_i - x_j) \cdot \nabla_{v_i} f_{k,N} \\
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Critical new idea

the new term has the same scaling as
the convolution at the limit:

⇒ behaves well with (weighted) L^p norms

$$\|f_{k,N}\|_{L_{\lambda(t)e_k}^q}^q = \int_{\mathbb{T}^{kd} \times \mathbb{R}^{kd}} |f_{k,N}|^q e^{\lambda(t) e_k}$$

$$e_k(x_1, v_1, \dots, x_k, v_k) = \sum_{i \leq k} (1 + |v_i|^2) + \frac{1}{N} \sum_{i, j \leq k} \phi(x_i - x_j).$$

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Avoid losing a derivative in v_i

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Avoid losing a derivative in v_i

In parallel:

- Lacker (2021) studies the propagation of marginals on 1st (2nd) order systems with non-degenerate diffusion using the relative entropy, but require interaction kernels K in some exponential Orlicz space
- Jabin-Poyato-S, 2021, for non-exchangeable systems

New results: $d = 2, 3$.

Assume $K \in L^p(\mathbb{T}^d)$, $p > 1$, and define

$$\lambda(t) = \frac{1}{\Lambda(1+t)}, \quad L = \frac{C}{\lambda(1)^\theta} \|K\|_{L^p}^q, \quad q > p'.$$

Consider a renormalized solution f_N to the Liouville eq satisfying the Gaussian decay with $f_N^0 \in L^\infty(\mathbb{T}^{dN} \times \mathbb{R}^{dN})$, such that

$$\int_{\mathbb{T}^{kd} \times \mathbb{R}^{kd}} |f_{k,N}^0|^q e^{\lambda(0) e_k} \leq F_0^k, \quad \sup_{t \leq 1} \int_{\mathbb{T}^{Nd} \times \mathbb{R}^{Nd}} |f_N|^q e^{\lambda(t) e_N} \leq F^N$$

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Then, one has that

$$\sup_{t \leq T} \int_{\mathbb{T}^{kd} \times \mathbb{R}^{kd}} |f_{k,N}|^q e^{\lambda(t) e_k} \leq 2^k F_0^k + F^k 2^{2k-N-1}$$

New results (Bresch, Jabin & JS)

Assume that ($d=2$)

- $K = -\nabla\phi \in L^p(\mathbb{T}^d)$, for some $p > 1$,
- $\int_{\mathbb{T}^d} e^{\theta\phi(x)} dx < +\infty$, $\theta > 0$,
- Let f be the unique smooth solution to the Vlasov equation with initial data $f^0 \in C^\infty(\mathbb{T}^d \times \mathbb{R}^d)$ & $\int_{\mathbb{T}^d \times \mathbb{R}^d} f^0 e^{\beta|v|^2} < \infty$
- $f_{k,N}^0 \rightharpoonup (f^0)^{\otimes k}$ in L^1 , $f^{\otimes k} = \prod_{i=1}^k f(t, x_i, v_i)$
- $\|f_{k,N}^0\|_{L^\infty(\mathbb{T}^{dN} \times \mathbb{R}^{dN})} \leq M^k$, for some $M > 0$, $\forall k < N$

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- $\|f_{k,N}^0\|_{L^\infty(\mathbb{T}^{dN} \times \mathbb{R}^{dN})} \leq M^k$, for some $M > 0$, $\forall k < N$

Then, there exists T^* such that

$$f_{k,N} \rightharpoonup f^{\otimes k}, \text{ in } L_{loc}^q([0, T^*] \times \mathbb{T}^{kd} \times \mathbb{R}^{kd})$$

Strong propagation of Chaotic/Tensorized law

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$$f_{k,N} - f^{\otimes k}.$$

Quantitative estimate for the case $K \in L^2$.

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Quantitative estimate for the case $K \in L^2$.

- Extension to the stochastic case of **mildly singular kernels**
- Does not apply directly in $d \geq 3$ to i.i.d. X_i^0 as $\int e^{\lambda\phi(x)} dx = +\infty$, $\forall \lambda > 0$, if $\phi = \frac{c}{|x|^{d-2}}$.

New results: **First order systems**

$$dX_i(t) = \frac{1}{N} \sum_{j \neq i} K(X_i - X_j) dt + \sqrt{2\sigma} dW_i$$



$$\frac{\partial f}{\partial t} + \operatorname{div} (f(K * f) - \sigma \nabla f) = 0$$

Applications:

- Keller-Segel, Euler / Navier-Stokes

Fetecau, Huang-Liu, Pickl, Bresch-Jabin-Wang,...

Chorin, Goodman, Beale-Majda, Cottet, Fournier, S.,

Hauray, Jabin-Wang, Wynter,

Rosenzweig, Duerinckx, Serfaty ($u \in W^{1,\infty}$),...

New results: First order models (Bresch, Jabin & JS).

Assume that

- $K \in L^p(\mathbb{T}^d)$, for some $p > 1$, $K \sim \frac{1}{|x|^s}$ $s < d$
- $(\operatorname{div} K)_- \in L^\infty(\mathbb{T}^d)$,
- Let f be the unique smooth solution to the transport equation with initial data $f^0 \in C^\infty(\mathbb{T}^d)$,
- $f_{k,N}^0 \rightharpoonup (f^0)^{\otimes k}$ in L^1 ,
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Then, there exists T^* such that

$$f_{k,N} \rightharpoonup f^{\otimes k}, \text{ in } L_{loc}^q([0, T^*] \times \mathbb{T}^{kd})$$

\implies Strong propagation of chaos

Non-exchangeable systems:

A prototype to study non-exchangeable systems is

$$dX_i = \sum_{j=1}^N w_{ij}^N K(X_i - X_j) dt + \sqrt{2\sigma} dW_i,$$

where

$$\frac{1}{N} \rightarrow w_{ij}^N$$

non-necessarily symmetric
interaction weights

Assumptions on the interaction weights:

Objective: stay within the mean-field limit

$\max_{1 \leq i \leq N} \sum_{j=1}^N |w_{ij}^N| = O(1)$ Total interaction with any object must be finite

$\max_{1 \leq i, j \leq N} |w_{ij}^N| \xrightarrow{N \rightarrow \infty} 0$ Individual coefficients should be small

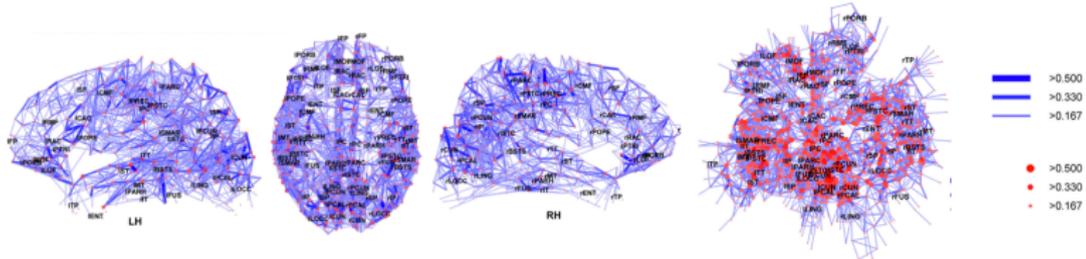
Some examples:

Brain neural networks

Many neurons: Human brains contain $\sim 86 \cdot 10^9$ neurons.

Sparseness & modularity: Each neuron has synaptic connections with only $7 \cdot 10^3$ neurons. Human Connectome is organized into structural cores and modules.

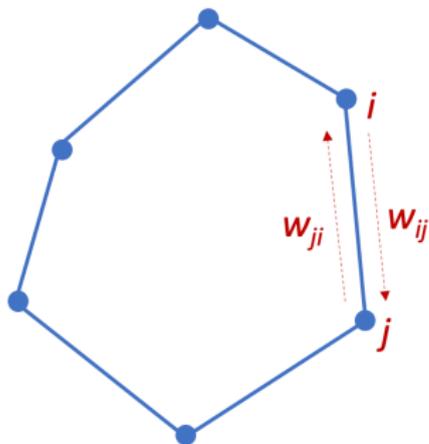
Synchronization: emerges as a consequence of the interplay of the local dynamics in regions with large intra connectivity and the topology of non-symmetric connectivities.



Some examples:

Agents on a graph

Brain structure, Epidemiology, Machine Learning, ...



Sparseness: We don't expect to have many interactions

$w_{ij} = 0$ for most i, j .

There is another scale within the graph: $\forall i$, the number $\#j$ of order N^θ such that $w_{ij} \sim \frac{1}{N^\theta}$

$\implies w_{ij}$ satisfy the assumptions and the law of large numbers still applies

How different is the study of the mean-field limit of the classical case with respect to this context?

Strategies regarding the structure of w_{ij} : **simple structure**

For $w_{ij} = m_j$, define

$$\nu_N = \sum_{j=1}^N m_j \delta(x - X_j)$$

which verifies (symmetrization)

$$\frac{\partial \nu_N}{\partial t} + \operatorname{div} ((K \star \nu_N) \nu_N) = \sigma \Delta \nu_N$$

\implies In Fluid Mechanics gives the total vorticity,
but ν_N can lose the probability measure character
and modulated energy techniques could give problems.

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For m_j random and i.i.d. Then $m_j \sim \frac{1}{N}$

Defining $\tilde{X}_j = m_j X_j \implies$ case of exchangeable systems
extend the entropy approaches

Strategies regarding the structure of w_{ij}^N :

When w_{ij}^N do not have any simple structure

- It no longer seems possible to define an empirical measure

↪ Try to incorporate w_{ij}^N as part of the state

Can we write

$$w_{ij}^N = w(w_{ij}^N) = w(\xi_i, \xi_j), \quad \xi_j \in [0, 1]$$

for some distribution of ξ_j and some kernel w ?

⇒ we go back to a classical mean-field context for which the regularity of w is crucial

Strategies regarding the structure of w_{ij}^N :

We would like to solve the mean-field limit equation associated with this new kernel

$$\begin{aligned} & \partial_t \bar{f}(t, x, \xi) \\ & + \operatorname{div}_x \left(\bar{f}(t, x, \xi) \int_0^1 w(\xi, \zeta) \int_{\mathbb{R}^d} K(x-y) \bar{f}(t, dy, d\zeta) \right) \\ & = \sigma \Delta \bar{f}(t, x, \xi) \end{aligned}$$

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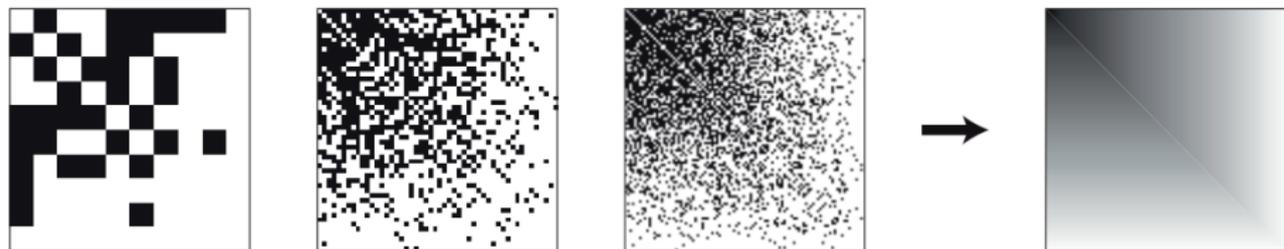
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- There is no initial data
- We are not interested in \bar{f} , but in $\int_0^1 \bar{f}(t, x, d\xi)$ which should be the limit of the “empirical measure”
- Need for an analytical context: Graph Theory

Large-scale limits of symmetric dense graphs

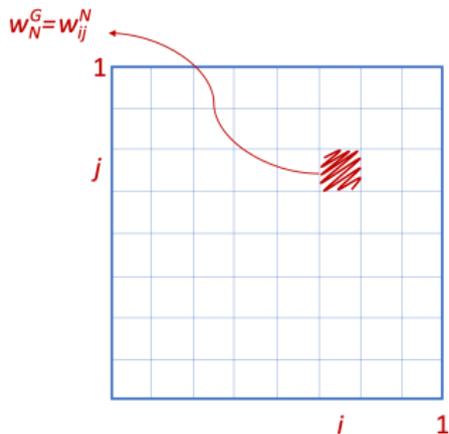
Let $G = (V, E, W_N)$ a finite graph with vertex $V = \llbracket 1; N \rrbracket$, edges E , and adjacency matrix $W_N = (w_{ij}^N)_{ij}$



Large-scale limits of symmetric dense graphs

Let $G = (V, E, W_N)$ a finite graph with vertex $V = \llbracket 1; N \rrbracket$, edges E , and adjacency matrix $W_N = (w_{ij}^N)_{ij}$

From w_{ij}^N we construct a piecewise constant function w_N^G



$$w_N^G(\xi, \zeta) = \sum_{i,j=1}^N w_{ij}^N \mathbf{1}_{[\frac{i-1}{N}, \frac{i}{N})}(\xi) \mathbf{1}_{[\frac{j-1}{N}, \frac{j}{N})}(\zeta)$$

After permutation of indexes

$w_N^G \rightarrow w \in L^\infty([0, 1]^2)$, w symmetric,
for **dense graphs** $|E(G_N)| \approx |V(G_N)|^2$.

Large-scale limits of symmetric dense graphs

Let $G = (V, E, W_N)$ a finite graph with vertex $V = \llbracket 1; N \rrbracket$, edges E , and adjacency matrix $W_N = (w_{ij}^N)_{ij}$

- $w_N^G(\xi, \zeta) = \sum_{i,j=1}^N w_{ij} \mathbb{1}_{[\frac{i-1}{N}, \frac{i}{N})}(\xi) \mathbb{1}_{[\frac{j-1}{N}, \frac{j}{N})}(\zeta)$
- $w_N^G \rightarrow w \in \{L^\infty([0, 1]^2), w \text{ symmetric}\} \equiv$ **Graphons**,
for **dense graphs** $|E(G_N)| \approx |V(G_N)|^2$.
- **The cut metric:** $\delta_\square(w, w_N^G)$
there exists a measure preserving map ϕ_N on $[0, 1]$ st

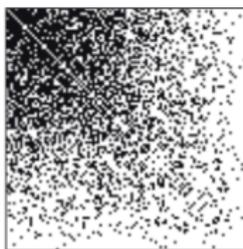
$$\left\| w - w_{\phi_N(i), \phi_N(j)}^N \right\|_{L^\infty \rightarrow L^1} \xrightarrow{N \rightarrow \infty} 0$$

What about mean-field limits on symmetric dense graphs?

Medvedev '18-'19

- For any **graphon** w , we can find finite graphs G_N approximating $0 \leq w \leq 1$, with weights

$$w_{ij}^N = N \int_{[\frac{i-1}{N}, \frac{i}{N}] \times [\frac{j-1}{N}, \frac{j}{N})} w(\xi, \zeta) d\xi d\zeta,$$



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$$w_N^G \xrightarrow{N \rightarrow \infty} w \text{ in } L^1([0, 1]^2) \implies \delta_{\square}(w, w_N^G) \rightarrow 0$$

- $K \in W^{1, \infty}$
- Stability estimate:

$$\frac{d}{dt} \int_0^1 \int_{\mathbb{R}^d} |f_N - f| dx d\xi \leq C_1 \int_0^1 \int_{\mathbb{R}^d} |f_N - f| dx d\xi + C_2 \|w_N^G - w\|_{L^\infty \rightarrow L^1}$$

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$\implies \mu_N \rightharpoonup \int_0^1 f(t, x, \xi) d\xi$ and f satisfies

$$\partial_t f(t, x, \xi) + \operatorname{div}_x \left(f(t, x, \xi) \int_0^1 \int_{\mathbb{R}^d} w(\xi, \zeta) K(x - y) f(t, y, \zeta) dy d\zeta \right) = 0$$

What about mean-field limits on symmetric dense graphs?

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$$w_{ij}^N = N \int_{[\frac{i-1}{N}, \frac{i}{N}] \times [\frac{j-1}{N}, \frac{j}{N}]} w(\xi, \zeta) d\xi d\zeta, \quad \boxed{\sup_{i,j} N |w_{ij}^N| = O(1)}$$

$$w_N^G \xrightarrow{N \rightarrow \infty} w \text{ in } L^1([0, 1]^2) \implies \delta_{\square}(w, w_N^G) \rightarrow 0$$

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What about mean-field limits on sparse graphs, $|E(G_N)| \approx |V(G_N)|$?

• $\sup_{N \in \mathbb{N}} \left[\max_i \sum_j |w_{ij}^N| + \max_j \sum_i |w_{ij}^N| \right] \leq C$ $\lim_{N \rightarrow \infty} \max_{i,j} |w_{ij}^N| = 0.$

• $K \in W^{1,1} \cap W^{1,\infty}$

• X_i^0 independent (but not *i.i.d.*) st their laws f_i^0 verify

$$\sup_{N \in \mathbb{N}} \max_i \left\{ \int_{\mathbb{R}^d} |x|^2 f_i^0(x) dx, \|f_i^0\|_{W^{1,1} \cap W^{1,\infty}} \right\} < \infty$$

What about mean-field limits on sparse graphs, $|E(G_N)| \approx |V(G_N)|$?

- $\sup_{N \in \mathbb{N}} \max_i \sum_j |w_{ij}^N| + \max_j \sum_i |w_{ij}^N| \leq C, \quad \lim_{N \rightarrow \infty} \max_{i,j} |w_{ij}^N| = 0.$
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$$\sup_{N \in \mathbb{N}} \max_i \left\{ \int_{\mathbb{R}^d} |x|^2 f_i^0(x) dx, \|f_i^0\|_{W^{1,1} \cap W^{1,\infty}} \right\} < \infty$$

\implies the mean-field limit is described by (*extended graphon*)

$$\begin{aligned} w &\in L_\xi^\infty([0, 1], \mathcal{M}_\zeta([0, 1])) \cap L_\zeta^\infty([0, 1], \mathcal{M}_\xi([0, 1])) \\ f &\in L_{t,\xi}^\infty([0, T] \times [0, 1], W_x^{1,1} \cap W_x^{1,\infty}(\mathbb{R}^d)) \end{aligned}$$

$$\lim_{k \rightarrow \infty} \sup_{0 \leq t \leq T} \mathbb{E} W_1 \left(\frac{1}{N_k} \sum_{i=1}^{N_k} \delta_{X_i(t)}, \int_0^1 f(t, \cdot, \xi) d\xi \right) = 0.$$

Idea of the proof

Propagation of independence

$$\frac{d\bar{X}_i}{dt} = \sum_{j=1}^N w_{ij} \int_{\mathbb{R}^d} K(\bar{X}_i - y) \bar{f}_j(t, y) dy,$$

$$\bar{X}_i(0) = X_i^0 \quad (\text{independent})$$

$\bar{f}_i(t, \cdot) = \text{Law}(\bar{X}_i(t))$ verify

$$\partial_t \bar{f}_i + \text{div}_x \left(\bar{f}_i(t, x) \sum_{j=1}^N w_{ij} \int_{\mathbb{R}^d} K(x - y) \bar{f}_j(t, dy) \right) = 0,$$

$$\bar{f}_i(0, x) = f_i(0, x).$$

\implies

$$\mathbb{E} W_1 \left(\frac{1}{N} \sum_{i=1}^N \delta_{X_i(t)}, \frac{1}{N} \sum_{i=1}^N \bar{f}_i(t, \cdot) \right) \leq C_1(t) \max_{i,j} |w_{ij}|^{1/2} + \frac{C_2(t)}{N^{\theta_d}}.$$

Consequences of the propagation of independence:

- For general connectivities w_{ij}^N ,
 $\not\Rightarrow$ the limit of the corresponding 1-particle distribution

$$\frac{1}{N} \sum_{i=1}^N \bar{f}_i(t, x).$$

- The equation for 1-particle distribution evolves
a kind of 2-particle distribution

$$\frac{1}{N^2} \sum_{i,j=1}^N w_{ij}^N \bar{f}_i(t, x) \bar{f}_j(t, x)$$

- \Rightarrow There is a hierarchy of equations indexed by trees

Graphon-like reformulation

- We can recast the equation for \bar{f}_i using graphons:

$$w_N(\xi, \zeta) := \sum_{i,j} N w_{ij}^N \mathbb{1}_{[\frac{i-1}{N}, \frac{i}{N})}(\xi) \mathbb{1}_{[\frac{j-1}{N}, \frac{j}{N})}(\zeta),$$

$$f_N(x, \xi) := \sum_i \bar{f}_i(t, x) \mathbb{1}_{[\frac{i-1}{N}, \frac{i}{N})}(\xi).$$

- (w_N, f_N) is a solution of the generalized Vlasov equation

$$\partial_t f_N(t, x, \xi) + \operatorname{div}_x \left(f_N(t, x, \xi) \int_0^1 w_N(\xi, \zeta) \int_{\mathbb{R}^d} K(x-y) f_N(t, dy, d\zeta) \right) = 0$$

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Q1) Compactness of (w_N, f_N)

Q2) Identification of the limit (w, f) in an appropriate topology.

Q3) The limit (w, f) satisfies the generalized Vlasov equation

⇒ **New compactness result in the spirit of Lovasz & Szegedy.**

New hierarchy of observables

- For finite tree T (and not any arbitrary graph), we define

$$\begin{aligned} \tau(T, w_N, f_N)(t, x_1, \dots, x_{|T|}) \\ = \int_{[0, 1]^{|T|}} \prod_{(k,l) \in \mathcal{E}(T)} w_N(\xi_k, \xi_l) \prod_{m=1}^{|T|} f_N(t, x_m, \xi_m) d\xi_1 \dots d\xi_{|T|} \end{aligned}$$

These observables include the 1-particle distribution for the tree $T = T_1$ with only one vertex:

$$\tau(T_1, w_N, f_N)(t, x) = \frac{1}{N} \sum_{i=1}^N \bar{f}_i(t, x).$$

- Our observables completely entangle the kernel with the initial conditions

New hierarchy of observables

- A critical point is that these observables solve an **independent hierarchy of equations**

$$\partial_t \tau(T, w_N, f_N) + \sum_{i=1}^{|T|} \operatorname{div}_{x_i} \left(\int_{\mathbb{R}^d} K(x_i - z) \tau(T + i, w_N, f_N)(t, x_1, \dots, x_{|T|}, z) dz \right) = 0$$

where $T + i$ denotes the tree obtained from T by adding a leaf on the i -th vertex

- They naturally extend the notion of marginals, and hierarchy of marginals to **non-exchangeable** systems
If the particles were exchangeable, observables would depend only on the number of nodes \rightarrow classical hierarchy

New hierarchy of observables

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- The kernel w does not appear explicitly in this equation, it only appears in the definition of observable

New hierarchy of observables

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$$\partial_t \tau(T, w_N, f_N) + \sum_{i=1}^{|T|} \operatorname{div}_{x_i} \left(\int_{\mathbb{R}^d} K(x_i - z) \tau(T + i, w_N, f_N)(t, x_1, \dots, x_{|T|}, z) dz \right) = 0$$

where $T + i$ denotes the tree obtained from T by adding a leaf on the i -th vertex

- We don't prove the convergence of f_N to f in a direct sense. This must be inferred from the convergence of $\tau(T, w_N, f_N)$ that gives the correct topology for the convergence

\implies we only need that $\tau(T, w_N, f_N)(t = 0, x_1, \dots, x_{|T|})$ converges

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- In our strategy there are two types of limits:
 - 1) Propagation of independence, limit of zero correlation: the coupled system: the PDE for \bar{f}_i and the DS for \bar{X}_i
 - 2) Push the graph to infinity: the limit of $\lim_{N \rightarrow \infty} \tau(T, w_N, f_N)$

What about the metric to estimate compactness? Stability

- For $\lambda > 0$ and any $w, \tilde{w} \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$
 $f^0, \tilde{f}^0 \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$, we define

$$\mathcal{D}_\lambda((w, f^0), (\tilde{w}, \tilde{f}^0)) = \sup_{\text{trees } \tau} \lambda^{|\tau|/2} \left\| \tau(T, w, f^0) - \tau(T, \tilde{w}, \tilde{f}^0) \right\|_{L^2(\mathbb{R}^{d|\tau|})}$$

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- **Stability:**

$f, \tilde{f} \in L_t^\infty([0, t_*]; L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty}))$ unique solutions of

$$\partial_t f(t, x, \xi) + \operatorname{div}_x \left(f(t, x, \xi) \int_0^1 \int_{\mathbb{R}^d} w(\xi, \zeta) K(x-y) f(t, y, \zeta) dy d\zeta \right) = 0$$

$$\Rightarrow \left\| \int_0^1 f d\xi - \int_0^1 \tilde{f} d\xi \right\|_{L_t^\infty([0, t_*], L_x^2)} \lesssim \frac{1}{(\ln |\ln \mathcal{D}_\lambda((w, f^0), (\tilde{w}, \tilde{f}^0))|)^{\frac{1}{2}+}}$$

Compactness result towards extended graphons

- Consider $w_N \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$ and $f_N \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$

$$\sup_{N \in \mathbb{N}} \|w_N\|_{L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi} < \infty, \quad \sup_{N \in \mathbb{N}} \|f_N\|_{L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})} < \infty$$

- Then, there is a subsequence $\{N_k\}_{k \in \mathbb{N}}$ and there are $w \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$ and $f \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$ st

$$\tau(T, w_{N_k}, f_{N_k}) \rightarrow \tau(T, w, f) \quad \text{in} \quad L_{\text{loc}}^p(\mathbb{R}^{d|T|}), \quad 1 \leq p < \infty$$

Compactness result towards extended graphons

- Consider $w_N \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$ and $f_N \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$

$$\sup_{N \in \mathbb{N}} \|w_N\|_{L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi} < \infty, \quad \sup_{N \in \mathbb{N}} \|f_N\|_{L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})} < \infty$$

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$$\tau(T, w_{N_k}, f_{N_k}) \rightarrow \tau(T, w, f) \quad \text{in} \quad L_{\text{loc}}^p(\mathbb{R}^{d|T|}), \quad 1 \leq p < \infty$$

- **Reformulation** $\tau(T, w, f) = \int_0^1 F(w, f) d\xi$ in terms of a countable algebra of transforms $F \in \mathcal{F}$ consistent with the adding-leafs process.
- New **compactness-by-rearrangement lemma** reminiscent of Szemerédi lemma proving that $F(w_N, f_N)$ must converge in L_{loc}^p modulo measure-preserving rearrangements w.r.t. ξ .
- **Invariance under rearrangements** of $\tau(T, w, f)$.

Compactness result towards extended graphons

- Consider $w_N \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$ and $f_N^0 \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$

$$\sup_{N \in \mathbb{N}} \|w_N\|_{L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi} < \infty, \quad \sup_{N \in \mathbb{N}} \|f_N^0\|_{L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})} < \infty$$

- Then, there is a subsequence $\{N_k\}_{k \in \mathbb{N}}$ and there are $w \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi$ and $f^0 \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$ st

$$\mathcal{D}_\lambda((w_{N_k}, f_{N_k}^0), (w, f^0)) \rightarrow 0, \quad \text{as } N \rightarrow \infty$$

for a sufficiently small $\lambda > 0$.

The identification of the limit

1) Once the limit (w, f^0) has been calculated

$$w \in L_\xi^\infty \mathcal{M}_\zeta \cap L_\zeta^\infty \mathcal{M}_\xi \text{ and } f^0 \in L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty})$$

2) We solve

$$\partial_t f(t, x, \xi) + \operatorname{div}_x \left(f(t, x, \xi) \int_0^1 \int_{\mathbb{R}^d} w(\xi, \zeta) K(x-y) f(t, y, \zeta) dy d\zeta \right) = 0$$

$$\text{to get } f \in L^\infty([0, t_*], L_\xi^\infty(W_x^{1,1} \cap W_x^{1,\infty}))$$

3) The stability result gives

$$\left\| \int_0^1 f d\xi - \frac{1}{N_k} \sum_i \bar{f}_i \right\|_{L_t^\infty([0, t_*], L_x^2)} \lesssim \frac{1}{(\ln |\ln \mathcal{D}_\lambda((w, f^0), (w_{N_k}, f_{N_k}^0))|)_+^{\frac{1}{2}}} \rightarrow 0$$

$\implies W_1$ -convergence by the propagation of moments.

The identification of the limit

4) Therefore, we can conclude that

$$\begin{aligned} \mathbb{E} W_1 \left(\frac{1}{N_k} \sum_i \delta_{X_i}, \int_0^1 f(t, \cdot, \xi) d\xi \right) \\ \leq \mathbb{E} W_1 \left(\frac{1}{N_k} \sum_i \delta_{X_i}, \frac{1}{N_k} \sum_i \bar{f}_i(t, \cdot) \right) \\ + W_1 \left(\frac{1}{N_k} \sum_i \bar{f}_i(t, \cdot), \int_0^1 f(t, \cdot, \xi) d\xi \right) \rightarrow 0 \end{aligned}$$

Conclusions and perspectives

Advantages

- Joint approach mean-field and large-graph limit.
- Large class of sparse non-symmetric graphs.
- Weights not coming as discretization of fixed continuous objects.
- Better results in the noisy case:

$$\|\tau(T, w, f) - \tau(T, w_N, f_N)\|_{L^2(\mathbb{R}^{d|T|})} \rightarrow 0$$

Open questions

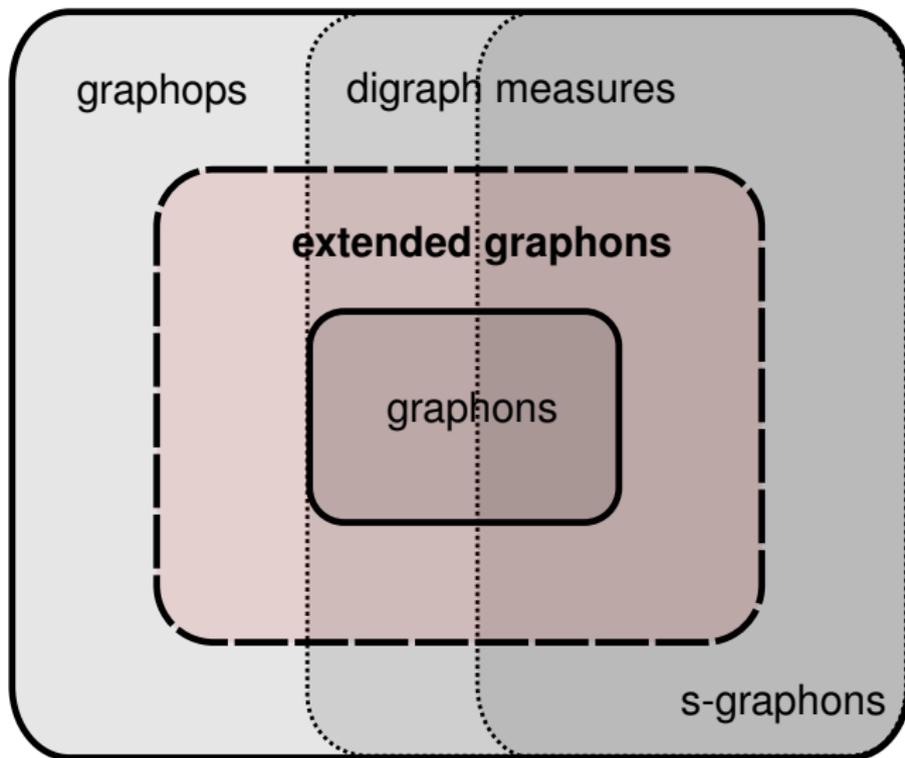
- What if $K \notin W^{1,\infty}$?
only used in the propagation of independence

Thanks for your attention!

Further previous results in the non-exchangeable case

- ▷ **Dense graph limits** $|E(G_N)| \approx |V(G_N)|^2$.
(LGL) Lovasz, Szegedy '06: **Graphons** $w \in L^\infty([0, 1]^2)$.
(MFL) Chiba, Medvedev '19, Kaliuzhnyi-Verbovetski, Medvedev '18.
- ▷ **Sparse graph limits** $|E(G_n)| \approx |V(G_n)|$.
(LGL) Benjamini, Schram '01: **Graphings**.
- ▷ **Intermediate density** $|V(G_n)| \ll |E(G_n)| \lesssim |V(G_n)|^2$.
(LGL) Borgs, Chayes, Cohn, Zhao '18-'19: **L^p graphons**.
- ▷ **Intermediate density** $|V(G_n)| \lesssim |E(G_n)| \lesssim |V(G_n)|^2$.
(LGL) Backhausz, Szegedy '20: **Graphops**
(LGL) Kunszenti-Kovasz, Lovasz, Szegedy '19: **s-graphons**
(LGL) Kuehn, Xu '21: **Digraph-measures**
(MFL) Gkogkas, Kuehn '20; Kuehn, Xu '21.

Comparison of the various graph limit theories



Entropy solution

- A function $f_N \in L^\infty([0, 1] \times \mathbb{T}^{dN} \times \mathbb{R}^{dN})$ satisfying the Gaussian decay

$$\sup_{t \leq 1} \int_{\mathbb{T}^{dN} \times \mathbb{R}^{dN}} e^{\beta \sum_{i \leq N} |v_i|^2} f_N dx_1 dv_1 \dots dx_N dv_N \leq V^N,$$

Entropy solution

- A function $f_N \in L^\infty([0, 1] \times \mathbb{T}^{dN} \times \mathbb{R}^{dN})$ satisfying the Gaussian decay is an **entropy solution** iff all marginals $f_{k,N}$ for $1 \leq k \leq N$ verify

$$\int_0^T \int_{\mathbb{T}^{dk} \times \mathbb{R}^{dk}} e^{\lambda e_k} |f_{k,N}|^{q-1} \operatorname{sign}(f_{k,N}) L_k[f_{k,N}] dx_1 dv_1 \dots dx_k dv_k dt \geq 0,$$

Reduced energy

$$e_k(x_1, v_1, \dots, x_k, v_k) = \sum_{i \leq k} (1 + |v_i|^2) + \frac{1}{N} \sum_{i, j \leq k} \phi(x_i - x_j).$$

e_k is invariant under the advection component

$$L_k = \sum_{i \leq k} v_i \cdot \nabla_{x_i} + \frac{1}{N} \sum_{i, j \leq k} K(x_i - x_j) \cdot \nabla_{v_i}$$

Entropy solution

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- Renormalized or mild solutions can offer a natural way to prove it, while it is automatically satisfied if we have classical solutions.