

- Really impressed by the number of preprints in Enrique's office with clear and deep comments on.

- Stress tensor effects,
- Incompressibility/compressibility,
- Micro-meso-macro \implies Geophysical applications in mind.

Thanks to several Action Intégrée Picasso projects

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Je suis un mec pas très cool dans James Bond. Je m'appelle ?

- 1) Le nombre
- 2) Le chiffre
- 3) La virgule
- 4) La barre de fraction

Complétez cette citation de Georges Cantor :

" L'essence des mathématiques c'est....."

- 1) La liberté
- 2) L'égalité
- 3) La fraternité
- 4) Grave frais de ouf Askip

- 1) Un Mexicagone
- 2) Un Chiliagone
- 3) Un Brésiliagone
- 4) Un Andalousiagone

- 1) Un dodécaèdre
- 2) Un icosaèdre tronqué
- 3) Un triacontraèdre tronqué
- 4) Un ballondefootaèdre tout simplement.

Quel type de nombres n'existe pas ?

- 1) Les nombres têtus
- 2) Les nombres tordus
- 3) Les nombres rigolos
- 4) Les nombres bizarres

Lequel de ces personnages célèbres a donné son nom à un théorème en maths

- 1) Charlemagne
- 2) Napoléon
- 3) Louis XIV
- 4) Felipe VI

- 1) Les nombres cousins
- 2) Les nombres sexys
- 3) Les nombres vampires
- 4) Les nombres jumeaux

Prendre du poids, on voit pas ça ?

- 1) En théorie du contrôle
- 2) En limite champ moyen dans système de particules
- 3) En méca des fluides pour les endroits où ça se passe mal.
- 4) À un moment en avion pour un vol parabolique

Introduire un problème dual, on voit pas ça ?

- 1) En théorie du contrôle.
- 2) En limite champ moyen dans système de particules.
- 3) En méca des fluides pour des estimations magiques.
- 4) Si on est raisonnable.

The goal of the talk is exactly
a story on weights and duality for fluid mechanics / mean field
limit.

Joint works with

- P.–E. Jabin (Pennsylvania State University),
- M. Duerinckx (Université Libre Bruxelles).
- J. Soler (Granada University).
- Z. Wang (Pekin University)

- A set of small navigation icons typically found in Beamer presentations, including symbols for back, forward, search, and other slide controls.

Goal of the first part:

Well adapted dynamical weights for mean-field limit
to get result with low regularity

Topics:

- Mean Field limits (with singular kernels)
 - First order systems
 - Second order systems

"Idea" coming from Compressible Navier-Stokes equations:

⇒ Global existence à la Leray

Here large difference in the procedure and weights to be chosen.

Develop a duality method for mean field limit.

A) Compressible Navier-Stokes equations:

with velocity field u such that

and

where

- Compressible framework for weak regularity namely

$\operatorname{div} u$ not necessarily $L_t^1 L_x^\infty$.

1) Let us look at the propagation of the information

$$\int_{\Omega^2} \frac{|\rho_n(t, x) - \rho_n(t, y)|}{(h + |x - y|)^d} (w_n(t, x) w_n(t, y)) dx dy$$

with w_n solution of

$$\partial_t w_n + u_n \cdot \nabla w_n + \lambda P_n w_n = 0,$$

where $w_n|_{t=0} = 1$ and with P_n a positive penalization associated to (ρ_n, u_n) to be chosen and λ a large enough parameter to be chosen.

We look at the propagation of

$$R_h = \int_{\Omega} K_h(x-y) |\rho_n(t, x) - \rho_n(t, y)| w_n(t, x) w_n(t, y) dx dy.$$

Calculating the time derivative of R_h , we get

$$\begin{aligned} \frac{d}{dt} R(t) &= \int_{\Omega^2} \nabla K_h(x-y) \cdot (u_n(t, x) - u_n(t, y)) |\rho_n(t, x) - \rho_n(t, y)| w_n(t, x) w_n(t, y) \\ &\quad - \int_{\Omega^2} K_h(x-y) (\operatorname{div} u_n(t, x) - \operatorname{div} u_n(t, y)) \rho_n(t, x) s_n w_n(t, x) w_n(t, y) \\ &\quad + 2 \int_{\Omega^2} K_h(x-y) |\rho_n(t, x) - \rho_n(t, y)| x \\ &\quad (\partial_t w_n(t, x) + u_n(t, x) \cdot \nabla_x w_n(t, x) + \operatorname{div}_x u_n(t, x) w_n(t, x)) w_n(t, y) \end{aligned}$$

where s_n is the sign of $\rho_n(t, x) - \rho_n(t, y)$.

Choose $P_n \geq 0$ in terms of the unknowns appropriately
To conclude with a nice Gronwall Lemma.

For instance $P_n = M|\nabla u_n| + |\operatorname{div} u_n| + p(\rho_n)$
where $M(f)$ the maximal function of f .

2) We must show some properties on the weights w_n i.e.

$$0 \leq w_n \leq c_n \leq 1, \quad \int_{\Omega} \rho_n |\log w_n|^q < +\infty$$

with $q > 0$ to hope to get rid of the weights at the end
 \implies use the compact embedding given in the previous slide.

Remark:

Vacuum state \implies Necessity to consider

$w_n(t, x) + w_n(t, y)$ instead of $w_n(t, x)w_n(t, y)$.

+ Square functions instead of Maximal functions + averaging in h

Use translation properties of the operators.

B) Mean Field limits

1) First order:

Liouville of forward Kolmogorov equation (general kernels)

$$\partial_t \rho_N + \sum_{i=1}^N \operatorname{div}_{x_i} \left(\rho_N \frac{1}{N-1} \sum_{j \neq i}^N K(x_i - x_j) \right) = \sum_{i=1}^N \sigma_N \Delta_{x_i} \rho_N$$

2) Second order:

Linear advection-diffusion equation (repulsive)

$$\partial_t f_N + \sum_{i=1}^N v_i \cdot \nabla_{x_i} f_N + \sum_{i=1}^N \frac{1}{N-1} \sum_{j \neq i}^N K(x_i - x_j) \cdot \nabla_{v_i} f_N = \frac{\sigma^2}{2} \Delta_{v_i} f_N$$

with

$$K = \nabla V.$$

div K not necessarily L^∞ .

Goal: Quantitative estimates or at least uniform bounds
 \implies To get mean field limit justification

Tools:

Appropriate weights: modification of pure Gaussian

- 1) Modulated free energy (First order system).
- 2) L^p estimate (Second order system).

Our ideas:

- Choose a norm used with "regular" kernel K (truncation)
- Introduce dynamical weights
to cancel quantities containing $\operatorname{div} K$
- Study the effect of the weights
in terms of control with respect to N
- Go back to the original norm and conclude
using Lemma comparing kernel and regularized kernel.

I.1) A modulated free energy for first order systems

Consider N particles, identical and interacting two by two through the kernel K . For $X_i(t) \in \Pi^d$ the position of the i -th particle,

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

with mean field scaling and N independent Brownian motions W_i where

$$K = -\nabla V$$

Gradient flow

Main question: Behavior of the system as $N \rightarrow \infty$.

For simplicity in the talk: σ is fixed but $\sigma = \sigma_N$ is also of interest.

The most classical cases:

- **Poisson law** with $d = 2$ (Patlak-Keller-Segel)

$$V(x) = \lambda \log |x| + \text{perturbation},$$

- **Coulomb law**: For $n \geq 2$:

$$V(x) = d\lambda/|x|^{d-2} + \text{perturbation} \text{ if } d \geq 3,$$

$$V(x) = -\lambda \log |x| + \text{perturbation} \text{ if } d = 2.$$

Motivations and kernels covered by modulated free energy:

- Attractive Kernel :
The Patlak-Keller-Segel.
- Repulsive Kernel :
More general kernel than Riesz potential, Coulomb potential.

Focus on the joint law $\rho_N(t, x_1, \dots, x_N)$ of the process (X_1, \cdot, X_N) which solves the Liouville or forward Kolmogorov equation

$$\partial_t \rho_N + \sum_{i=1}^N \operatorname{div}_{x_i} \left(\rho_N \frac{1}{N-1} \sum_{j \neq i}^N K(x_i - x_j) \right) = \sigma \sum_{i=1}^N \Delta_{x_i} \rho_N$$

$$\rho_N|_{t=0} = \rho_N^0 \text{ such that } \int_{\Pi^{dN}} \rho_N^0 = 1.$$

Two previous approaches

Two key results appeared recently

- In Jabin-Wang, new estimates **relative entropy** of joint law

$$\frac{1}{N} \int_{\Pi^{dN}} \rho_N(t, X^N) \log \left(\frac{\rho_N(t, X^N)}{\bar{\rho}_N(t, X^N)} \right) dX^N$$

where $\bar{\rho}_N = \bar{\rho}^{\otimes N} = \prod_{i=1}^N \bar{\rho}(t, x_i)$ and ρ_N the joint law of the process (X_1, \dots, X_N) which satisfies the Liouville equation

$$\partial_t \rho_N + \sum_{i=1}^N \operatorname{div}_{x_i} \left(\rho_N \frac{1}{N} \sum_{j \neq i}^N K(x_i - x_j) \right) = \sigma \sum_{i=1}^N \Delta_{x_i} \rho_N.$$

They give **optimal rates of convergence** in $\frac{1}{\sqrt{N}}$
provided that $K, \operatorname{div} K \in W^{-1, \infty}$.

- **very well** – 2d Navier-Stokes because $\operatorname{div} K = 0$.
- **very poorly** – gradient flows: Log-Lipschitz Kernels.

See L. St Raymond, Bourbaki 70ème année, 2017–2018, no 1143.

Remark/idea: $\operatorname{div} K$ in relative entropy propagation is a bad term!

Similar quantity $\operatorname{div} u$ appears for compressible Navier-Stokes eqs. D.B., P.–E. Jabin (published *Ann. Math.* 2018) introduced weights satisfying PDE related to the unknowns in the quantity encoding the low regularity to cancel bad terms and prove some quantitative estimates on the density leading to compactness.

$$\begin{aligned} & \frac{1}{N} \int_{\Pi^{dN}} \rho_N(t, X^n) \log \frac{\rho_N(t, X^n)}{\bar{\rho}_N(t, X^n)} dX^N(t) \\ & \leq \frac{1}{N} \int_{\Pi^{dN}} \rho_N(0, X^N) \log \frac{\rho_N(0, X^N)}{\bar{\rho}_N(0, X^n)} dX^N(0) \\ & \quad - \frac{\sigma_N}{N} \int_0^t \int_{\Pi^{dN}} d\rho_N \left| \nabla \log \frac{\rho_N}{\bar{\rho}_N} \right|^2 \\ & \quad - \frac{1}{N^2} \sum_{i,j=1}^N \int_0^t \int_{\Pi^{dN}} \rho_N(K(x_i - x_j) - K \star_x \bar{\rho}(x_i)) \cdot \nabla_{x_i} \log \bar{\rho}_N dX^N ds \\ & \quad - \frac{1}{N^2} \sum_{i,j=1}^N \int_0^t \int_{\Pi^{dN}} \rho_N(\operatorname{div} K(x_i - x_j) - \operatorname{div} K \star_x \bar{\rho}(x_i)) dX^N ds \end{aligned}$$

where we call $\bar{\rho}_N(t, X^N) = \prod_{i=1}^N \bar{\rho}(t, x_i)$.

Modify the relative entropy here to cancel $\operatorname{div} K$ and conclude ?

This was the starting point question with P.-E. Jabin and Z. Wang.

- Duerinckx-Serfaty and Serfaty (for $\sigma = 0$):

Modulated potential energy

$$\frac{1}{2} \int_{\Pi^{2d} \cap \{x \neq y\}} V(x-y)(\mu_N(dx) - \bar{\rho}(x)dx)(\mu_N(dy) - \bar{\rho}(y)dy)$$

(with $\mu_N = \left(\sum_{i=1}^N \delta(x - X_i(t)) \right) / N$ the empirical measure)
allows to deal with repulsive singular Riesz potential of the type

$$V = \frac{C}{|x|^\alpha} \text{ for } C > 0 \text{ and } \alpha < d$$

→ Works **beyond Poisson kernel**.

→ Does not work

for stochastic systems with attractive potentials.

→ Use the explicit formula of the kernel

allowing to reformulate the energy

in terms of potential or extension representation

(for the fractional laplacian) by Caffarelli-Silvestre

The Gibbs equilibria

Denote by G_N the **Gibbs equilibrium** of the system, and by $G_{\bar{\rho}_N}$ the corresponding distribution where the exact field is replaced by the mean field limit according to the law $\bar{\rho}$,

$$G_N(t, X^N) = \exp \left(- \frac{1}{2N\sigma} \sum_{i \neq j} V(x_i - x_j) \right),$$

$$G_{\bar{\rho}_N}(t, X^N) = \exp \left(- \frac{1}{\sigma} \sum_{i=1}^N V \star \bar{\rho}(x_i) + \frac{N}{2\sigma} \int_{\Pi^d} V \star \bar{\rho} \bar{\rho} \right),$$

$$G_{\bar{\rho}}(t, x) = \exp \left(- \frac{1}{\sigma} V \star \bar{\rho}(x) + \frac{1}{2\sigma} \int_{\Pi^d} V \star \bar{\rho} \bar{\rho} \right).$$

Our method uses the modified relative entropy

$$E_N \left(\frac{\rho_N}{G_N} \mid \frac{\bar{\rho}_N}{G_{\bar{\rho}_N}} \right) = \frac{1}{N} \int_{\Pi^{dN}} \rho_N(t, X^N) \log \left(\frac{\rho_N(t, X^N)}{G_N(X^N)} \frac{G_{\bar{\rho}_N}(t, X^N)}{\bar{\rho}_N(t, X^N)} \right) dX^N.$$

A modified free energy

One may also write

$$E_N\left(\frac{\rho_N}{G_N} \mid \frac{\bar{\rho}_N}{G_{\bar{\rho}_N}}\right) = \mathcal{H}_N(\rho_N | \bar{\rho}_N) + \mathcal{K}_N(G_N | G_{\bar{\rho}_N}),$$

where

$$\mathcal{H}_N(\rho_N | \bar{\rho}_N) = \frac{1}{N} \int_{\Pi^{dN}} \rho_N(t, X^N) \log\left(\frac{\rho_N(t, X^N)}{\bar{\rho}_N(t, X^N)}\right) dX^N$$

is exactly the relative entropy introduced in Jabin-Wang and

$$\mathcal{K}_N(G_N | G_{\bar{\rho}_N}) = -\frac{1}{N} \int_{\Pi^{dN}} \rho_N(t, X^N) \log\left(\frac{G_N(t, X^N)}{G_{\bar{\rho}_N}(t, X^N)}\right) dX^N$$

is expectation of modulated potential energy in Serfaty, Duerinckx multiplied by $1/\sigma$.

→ E_N is a modulated **free energy** for the system.

The time evolution of E_N

The modulated free energy E_N has the right algebraic structure with for any V even that

$$\begin{aligned}
 E_N \left(\frac{\rho_N}{G_N} \mid \frac{\bar{\rho}_N}{G_{\bar{\rho}_N}} \right) (t) &\leq E_N \left(\frac{\rho_N}{G_N} \mid \frac{\bar{\rho}_N}{G_{\bar{\rho}_N}} \right) (0) \\
 &- \frac{\sigma}{N} \int_0^t \int_{\Pi^{dN}} d\rho_N \left| \nabla \log \frac{\rho_N}{\bar{\rho}_N} - \nabla \log \frac{G_N}{G_{\bar{\rho}_N}} \right|^2 \\
 &- \frac{1}{2} \int_0^t \int_{\Pi^{dN}} \int_{\Pi^{2d} \cap \{x \neq y\}} \nabla V(x-y) \cdot \left(\nabla \log \frac{\bar{\rho}}{G_{\bar{\rho}}}(x) - \nabla \log \frac{\bar{\rho}}{G_{\bar{\rho}}}(y) \right) \\
 &\quad (d\mu_N - d\bar{\rho})^{\otimes 2} d\rho_N,
 \end{aligned}$$

where $\mu_N = \frac{1}{N} \sum_{i=1}^N \delta(x - x_i)$ is the empirical measure.

The main points of the proof

The previous simple expression leaves two main points in the proof

- Bound the right-hand side in terms of E_N .
- Show that E_N is almost positive or more specifically that for some constant C

$$E_N \left(\frac{\rho_N}{G_N} \mid \frac{\bar{\rho}_N}{G_{\bar{\rho}_N}} \right) (t) \geq \frac{1}{C} \mathcal{H}_N (\rho_N \mid \bar{\rho}_N) (t) - \frac{C}{N^\theta}.$$

Remark. See recent papers by A. Chodron de Courcel, M. Rosenzweig, S. Serfaty.

I-2) A weighted L_q quantity

Ssecond order systems of Mean field limit with singular kernels

D. B., P.–E. Jabin, J. Soler. A new approach to the mean-field limit of Vlasov-Fokker-Plank equations. Analysis & PDE vol. 18, No. 4, 2025.

- Strongly uses the diffusion in velocity
- Bound weighted L^p norms of the marginals.
- Mean-field limit for very singular interaction kernels including repulsive Poisson interactions in $2D$.
- Quantitative estimates for a general kernel in L^2

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Posed on the limiting 1-particle density $f(t, x, v)$

$$\partial_t f + v \cdot \nabla_x f + (K \star_x \rho) \cdot \nabla_v f = \frac{\sigma^2}{2} \Delta_v f \quad \text{with} \quad \rho = \int_{\mathbb{R}^d} f dv.$$

The mean-field of the Vlasov-Poisson equation has remained a long-standing open problem due to the difficulty in general of handling singular kernels for second-order model.

Known results:

In one dimension: Singular kernels.

Higher space dimensions: Regularized and cutted-off Kernels ; convergence results of discrete approximation for multidimensional Vlasov-Poisson systems with mollified mass and charge density.

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Uses BBGKY hierarchy equations for k point marginals: $f_{k,N}$
as introduced in Lacker's paper
(definition given later on in the duality approach).

Consider the weighted quantity

$$X_k(t) = \int |f_{k,N}|^q e^{\lambda(t)e_k}, \quad \lambda(t) = \frac{1}{\Lambda(1+t)}$$

with Λ a parameter and e_k the dynamical weight to be chosen.

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$$\begin{cases} dX_{i,N} = V_{i,N} dt, \\ dV_{i,N} = \frac{1}{N-1} \sum_{j:j \neq i}^N K(X_{i,N} - X_{j,N}) dt + \sqrt{2\alpha} dB_{i,N}, \end{cases} \quad (2)$$

where $\{B_{i,N}\}_{1 \leq i \leq N}$ are N independent Brownian motions and where the temperature $0 \leq \alpha < \infty$ is a fixed parameter (possibly vanishing).

$$\partial_t F_N + \sum_{i=1}^N \left(v_i \cdot \nabla_{x_i} F_N + \frac{1}{N-1} \sum_{j:j \neq i}^N K(x_i - x_j) \cdot \nabla_{v_i} F_N \right) = \alpha \sum_{i=1}^N \Delta_{v_i} F_N. \quad (3)$$

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$$F_N|_{t=0} = (f^\circ)^{\otimes N}, \quad (4)$$

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$$F_{N,1}(z) := \int_{\mathbb{D}^{N-1}} F_N(z, z_2, \dots, z_N) dz_2 \dots dz_N.$$

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Rigorous derivation of mean-field limit for Vlasov-Poisson still fully open in multi-dimension:

- The case of Lipschitz interactions $K(x)$ was handled by McKean in for the stochastic setting and by Braun and Hepp, and Dobrushin in the deterministic case.
- Mild singularities $K(x) \ll |x|^{-1}$ in Hauray-Jabin 09 and 15.
- Truncated kernels (essential for numerics) in Boers-Pickl 16, Lazarovici-Pickl 17, Pickl 19.
- Swarming models: Carrillo-Choi-Hauray-Salem 18 for cones of vision, Mucha-Peszek 18 for mild singular communication weights in Cucker-Smale.
- So-called Monokinetic limits with the full singularity were obtained in Duerinckx-Serfaty 20.
- Repulsive 2d Vlasov-Poisson-Fokker-Planck in Bresch-Jabin-Soler 22, with a partial result in 3d.

Hypothesis (H)

- Let $0 \leq \alpha < +\infty$ and let $K \in L^2_{loc}(\Omega; \mathbb{R}^d)$, and assume for convenience $K \in L^\infty_{loc}(|x| > 1)$.
- Consider a global weak duality solution $F_N \in L^\infty_{loc}(\mathbb{R}^+; L^1(\mathbb{D}^N) \cap L^\infty(\mathbb{D}^N))$ of the Liouville equation with f° -chaotic initial data for some density $f^\circ \in \mathcal{P}(\mathbb{D}) \cap L^\infty(\mathbb{D})$.
- Let $f \in L^\infty_{loc}(\mathbb{R}^+; \mathcal{P}(\mathbb{D}) \cap L^\infty(\mathbb{D}))$ be a bounded weak solution of the Vlasov equation with initial data f° ,
- Assume that for some $T > 0$ it satisfies $K * f \in L^\infty([0, T] \times \Omega)$ and has bounded Fisher information

$$\int_0^T \left(\int_{\mathbb{D}} |\nabla_v \log f|^2 f \right)^{\frac{1}{2}} < \infty.$$

Our method

- Requires the interaction kernel K to be in L^2 .
- Captures the structure of correlations in the dynamics.
- Works with or without diffusion
- Works for first and second order systems.
- Valuable for general domains.

A Duality approach (on first order for the talk)

The joint law solves the Liouville or Forward Kolmogorov equation

$$\partial_t F_N + \frac{1}{N-1} \sum_{i=1}^N \sum_{j \neq i} K(X_i - X_j) \cdot \nabla_{x_j} F_N = \alpha \sum_{i=1}^N \Delta_{x_i} F_N$$

with

$$F_N|_{t=0} = (f^0)^{\otimes N}.$$

Dual Backward Kolmogorov equation

$$\partial_t \Phi_N + \frac{1}{N-1} \sum_{i=1}^N \sum_{j \neq i} K(x_i - X_j) \cdot \nabla_{x_j} \Phi_N = -\alpha \sum_{i=1}^N \Delta_{x_i} \Phi_N$$

with

$$\Phi_N|_{t=T} = \bar{\Phi}_N.$$

We have

$$\int F_N(t = T) \bar{\Phi}_N dx_1 \cdots dx_N = \int (f^0)^{\otimes N} \Phi_N(t = 0) dx_1 \cdots dx_N.$$

The Goal : Show that as $N \rightarrow \infty$ for the limiting solution f

$$\int (f(t = T))^{\otimes N} \bar{\Phi}_N dx_1 \cdots dx_N - \int (f^0)^{\otimes N} \Phi_N(t = 0) dx_1 \cdots dx_N \rightarrow 0.$$

Choosing $\bar{\Phi}_N = \frac{1}{N} \sum_{i=1}^N \bar{\varphi}(X_i)$

$$\begin{aligned} & \int F_N(t = T) \bar{\Phi}_N dx_1 \cdots dx_N - \int (f(t = T))^{\otimes N} \bar{\Phi}_N dx_1 \cdots dx_N \\ &= \int F_{N,1}(t = T, x) \bar{\varphi}(x) dx - \int f(t = T, x) \bar{\varphi}(x) dx \rightarrow 0 \end{aligned}$$

That means weak convergence of the 1-marginal
(similar formula for other marginals)

$$\frac{d}{dt} \int (f(t))^{\otimes N} \phi_N dx_1 \cdots dx_N \rightarrow 0$$

Formal calculation seems to give a first order term.....

is equivalent to the dual convergence property

$$N \int_0^T \left(\int_{\Omega^N} V_f(z_1, z_2) \Phi_N f^{\otimes N} \right) dt \xrightarrow{N \uparrow \infty} 0,$$

- Write the system on $M_{N,n}$ and therefore on $C_{N,n}$.
- Remark that

$$\begin{aligned} & N \int_0^T \left(\int_{\Omega^N} V_f(z_1, z_2) \Phi_N f^{\otimes N} \right) dt \\ &= N \int_0^T \left(\int_{\Omega^2} V_f M_{N,2} f^{\otimes 2} \right) dt = N \int_0^T \left(\int_{\Omega^2} V_f C_{N,2} f^{\otimes 2} \right) dt. \end{aligned}$$

- Show that

$$N^{\frac{n}{2}} C_{N,n} \xrightarrow{*} n!^{\frac{1}{2}} \bar{C}_n, \quad \text{in } L^\infty(0, T; L_f^2(\Omega^n))$$

- Write the equation on \bar{C}_n and appropriate final condition to perform a uniqueness result to show that $\bar{C}_2 = 0$ and its friends.

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