



# Some Variations on the Mean Field Limit

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## Plan of the Talk

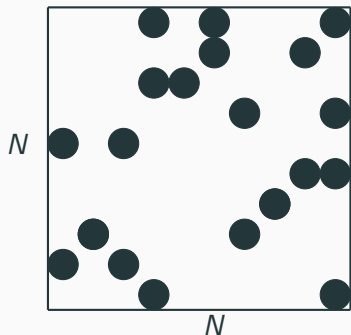
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- ▲ Mean Field Interaction vs Other Type of Interaction [joint work with F. Flandoli (SNS), C. Ricci (Univerisity of Pisa)]
  - a brief introduction
  - our contribution
  
- ▲ First Order Approximation [joint work with F. Flandoli (SNS), M. Aleandri(SNS)]
  - a brief introduction
  - technical difficulties

# Mean Field Interaction vs Other Type of Interaction

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# An introduction to scaling limit



$$\left(X_t^{i,N}\right)_{i=1,\dots,K_{\bar{\rho}}} \in \left(\mathbb{T}_N^d\right)^{K_{\bar{\rho}}}$$

$$K_{\bar{\rho}} = \lfloor \bar{\rho} N^d \rfloor$$

$\bar{\rho}$  = density of particles

$\text{supp}V$  = radius of a particle

$$dX_t^{i,N} = - \sum_{j=1}^{K_{\bar{\rho}}} \nabla V(X_t^{j,K} - X_t^{i,N}) dt + \varepsilon dB_t^i \quad i = 1, \dots, K_{\bar{\rho}}$$

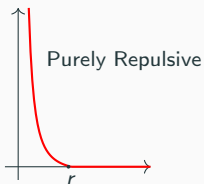
where  $V : \mathbb{R}^d \rightarrow \mathbb{R}$  and  $B_t^i$  are independent Brownian motion in  $\mathbb{R}^d$ ,  
 $0 < \varepsilon \ll 1$ .

# An introduction to scaling limit

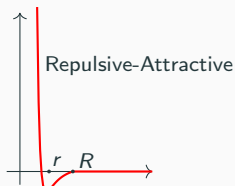
Training examples are related to biology, in particular to cellular adhesion. We consider  $V$  compactly supported and radial,

$$\text{supp}V = B(0, 1), \quad \nabla V(x) = V'(|x|) \frac{x}{|x|}.$$

Moreover we consider purely repulsive or attractive-repulsive potential,

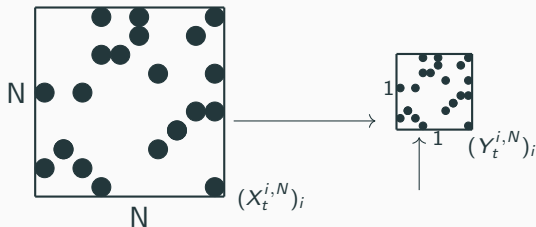


$$V(|x|) = \left( \frac{1}{|x|^\alpha} - C \right) \mathbb{1}_{\{|x| \leq r\}}, \quad \alpha > 0$$



$$V(|x|) = \left( \frac{r^\alpha}{|x|^\alpha} - \frac{r^\beta}{|x|^\beta} + C \right) \mathbb{1}_{\{|x| \leq R\}} \quad \alpha > \beta$$

## An introduction to scaling limit.



$$\left( Y_t^{i,N} \right)_{i=1, \dots, K_{\bar{\rho}}} \in \left( \mathbb{T}_1^d \right)^K$$

$$K_{\bar{\rho}} = \lfloor \bar{\rho} N^d \rfloor, \rho \in (0, 1),$$

- Scaling in space: in order to have  $Y_t^{i,N} \in \mathbb{T}_1^d$ ,  $Y_t^{i,N} = \frac{X_t^{i,N}}{N}$ .
- Scaling in time:  $Y_t^{i,N} = \frac{X_{t \cdot N^2}^{i,N}}{N}$ , in order to see the brownian movement at the macro scale, we need to accelerate the time.

## An introduction to scaling limit.

Note that  $W_t = \frac{1}{\sqrt{c}} B_{ct}$  for each  $c > 0$  is still a Brownian motion.

By Ito formula, calling  $W_t^i = \frac{1}{N} B_{tN^2}^i$  and by some integration by substitution, we obtain the dynamic for  $Y_t^{i,N}$ :

$$dY_t^{i,N} = -N \sum_{j=1}^{K_{\bar{\rho}}} \nabla V(N(Y_t^{j,N} - Y_t^{i,N})) dt + \varepsilon dW_t^i \quad i = 1, \dots, K_{\bar{\rho}}$$

$$dY_t^{i,N} = -\frac{1}{N^d} \sum_{j=1}^{K_{\bar{\rho}}} \nabla V_N(Y_t^{j,N} - Y_t^{i,N}) dt + \varepsilon dW_t^i \quad i = 1, \dots, K_{\bar{\rho}}$$

$$\text{with } V_N(x) = N^d V(Nx)$$

then  $\text{supp} V_N = B(0, N^{-1})$ .



## An introduction to scaling limit

If we forget about the scaling and we assume  $\nabla V_N(x) = \nabla V(x)$  we relapse on the Mean Field case.

$$\partial_t \rho = \frac{\epsilon^2}{2} \Delta \rho - \operatorname{div}((\nabla V * \rho)\rho)$$

- **Intermediate Interaction**  $\nabla V_N(x) = N^{\beta d} \nabla V(N^\beta x)$ ,  
 $\operatorname{supp} V_N = B(0, N^{-\beta})$  [Oeschlager '90]



$$\partial_t \rho = \frac{\epsilon^2}{2} \Delta \rho + \frac{C_V}{2} \Delta \rho^2$$

Volume fraction  $\approx N^{(1-\beta)d}$

- **Local Interaction**  $\nabla V_N(x) = N^d \nabla V(Nx)$ ,  $\operatorname{supp} V_N = B(0, N^{-1})$   
[Varadhan '91, Uchiyama '00]

$$\partial_t \rho = \Delta(P_V(\rho))$$

Volume fraction  $\approx 1$

# Macroscopic limit of Brownian particles with local interaction

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## Identity for the empirical measure

By Itô formula, if  $\varphi : \mathbb{T}^d \rightarrow \mathbb{R}$  is a smooth compact support test function, then

$$\begin{aligned} d\varphi\left(Y_t^{i,N}\right) &= -(\nabla\varphi)\left(Y_t^{i,N}\right) \frac{1}{N^d} \sum_{j \neq i} \nabla V_N\left(Y_t^{i,N} - Y_t^{j,N}\right) dt \\ &\quad + (\nabla\varphi)\left(Y_t^{i,N}\right) \epsilon dW_t^i + \frac{\epsilon^2}{2} \Delta\varphi\left(Y_t^{i,N}\right) dt. \end{aligned}$$

Therefore, for  $S_t^N := \frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} \delta_{X_t^{i,N}}$ ,

$$\begin{aligned} d\langle\varphi, S_t^N\rangle &= -\left\langle\nabla\varphi \int_{\mathbb{T}^d} \nabla V_N(\cdot - y) S_t^N(dy), S_t^N\right\rangle dt + \\ &\quad + \frac{\epsilon^2}{2} \langle\Delta\varphi, S_t^N\rangle dt + dM_t^{N,\varphi} \end{aligned}$$

with  $M_t^{N,\varphi} = \epsilon \frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} \int_0^t (\nabla\varphi)\left(Y_t^{i,N}\right) dW_t$

## Mean field case

If  $\nabla V_N$  is independent of  $N$ , and it is continuous and bounded, if  $S^N \rightarrow \rho$

$$\begin{aligned}\lim_{N \rightarrow \infty} \left\langle \nabla \varphi \int_{\mathbb{T}^d} \nabla V(\cdot - y) S_t^N(dy), S_t^N \right\rangle &= \\ &= \lim_{N \rightarrow \infty} \left\langle \nabla \varphi(\nabla V * S_t^N), S_t^N \right\rangle = \langle \nabla \varphi(\nabla V * \rho), \rho \rangle\end{aligned}$$

Then, passing to the limit in

$$\begin{aligned}d \langle \varphi, S_t^N \rangle &= - \left\langle \nabla \varphi \int_{\mathbb{T}^d} \nabla V(\cdot - y) S_t^N(dy), S_t^N \right\rangle dt + \\ &\quad + \frac{\epsilon^2}{2} \langle \Delta \varphi, S_t^N \rangle dt + dM_t^{N, \varphi}\end{aligned}$$

we get the weak formulation of

$$\partial_t \rho = \frac{\epsilon^2}{2} \Delta \rho + \operatorname{div}((\nabla V * \rho)\rho)$$

## Manipulation of the non linear term

$$\langle S_t^N, \nabla \varphi \cdot \nabla (V_N * S_t^N) \rangle = \frac{1}{N^{2d}} \sum_{i,j=1}^{K_{\bar{p}}} \nabla \varphi (Y_t^{i,N}) \cdot \nabla V_N (Y_t^{i,N} - Y_t^{j,N}).$$

By a symmetry argument, since  $\nabla V_N(-x) = -\nabla V_N(x)$

$$\begin{aligned} &= \frac{1}{N^{2d}} \sum_{i < j}^{K_{\bar{p}}} \left( \nabla \varphi (Y_t^{i,N}) - \nabla \varphi (Y_t^{j,N}) \right) \cdot \nabla V_N (Y_t^{i,N} - Y_t^{j,N}) \\ &= \frac{1}{N^{2d}} \sum_{i < j} \sum_{\alpha=1}^d \partial_{\alpha} \left( \varphi (Y_t^{i,N}) - \varphi (Y_t^{j,N}) \right) \partial_{\alpha} V_N (Y_t^{i,N} - Y_t^{j,N}) \end{aligned}$$

By Taylor expansion,  $\psi_{\alpha,\beta} := \partial_{\alpha} V(x) \cdot (x)_{\beta}$  and  $\psi := \partial_1 V(x) \cdot (x)_1$

$$\begin{aligned} &\approx \frac{1}{N^{2d}} \sum_{i < j} \sum_{\alpha,\beta=1}^d \partial_{\alpha} \partial_{\beta} \varphi (Y_t^{j,N}) (Y_t^{i,N} - Y_t^{j,N})_{\beta} \partial_{\alpha} V_N (Y_t^{i,N} - Y_t^{j,N}). \\ &= \frac{1}{N^{2d}} \sum_{i < j} \sum_{\alpha,\beta=1}^d \partial_{\alpha} \partial_{\beta} \varphi (Y_t^{j,N}) N^d \psi_{\alpha,\beta} \left( N(Y_t^{i,N} - Y_t^{j,N}) \right). \end{aligned}$$

# Local Equilibrium

**Local equilibrium property:** namely for each  $\varphi, \psi \in C_c(\mathbb{T}^d)$ . Let  $\rho : [0, T] \times \mathbb{T}^d \rightarrow \mathbb{R}$  and  $\bar{\rho} = \int \rho(y) dy$  and  $K = \lfloor \bar{\rho} N^d \rfloor$

$$\begin{aligned} \frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} \varphi \left( Y_t^{j,N} \right) \sum_j \psi \left( N(Y_t^{i,N} - Y_t^{j,N}) \right) &= \\ &= \frac{1}{N^d} \sum_j \varphi \left( Y_t^{j,N} \right) \sum_{|Y^i - Y^j| \leq \frac{r}{N}} \psi \left( N(Y_t^{i,N} - Y_t^{j,N}) \right) \\ &\rightarrow \int_{\mathbb{T}^d} \varphi(y) \Psi_V(\rho(y)) dy \end{aligned}$$

## Local Equilibrium

Given  $\psi \in C_c(\mathbb{R}^d)$  (it is on of  $\psi_{\alpha\beta}(x) = x_\beta \partial_\alpha V(x)$ )  $\bar{\rho} > 0$

$$\Psi_V(\bar{\rho}) := \lim_N \mathbb{E}_{\mu_G} \left[ \frac{1}{N^d} \sum_{i,j=1}^{K_{\bar{\rho}}} \psi(x^i - x^j) \right], \quad (\text{Virial Formula})$$

where  $\mu_G$ , is the invariant measure of the SDEs system ,

$$\mu_G(dx) = \frac{1}{Z_N} \exp \left( -\frac{1}{2} \sum_{i \neq j} V(x_i - x_j) \right) dx$$

with  $\mathbf{x} = (x_1, \dots, x_K) \in (\mathbb{T}_N^d)^{K_{\bar{\rho}}}$ . Interpretation:

$$\frac{1}{N^d} \sum_{i,j=1}^{K_{\bar{\rho}}} \psi(x^i - x^j) = \frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} F(x_i), \quad \text{with } F(x_i) = \sum_{|x^i - x^j| \leq r} \psi(x^i - x^j)$$

$\Psi_V$  is a sort of spatial average of local observables.

$$\begin{aligned}
&= \frac{1}{2N^d} \sum_{i,j=1}^{K_{\bar{\rho}}} \sum_{\alpha,\beta=1}^d \partial_{\alpha} \partial_{\beta} \varphi \left( Y_t^{j,N} \right) \psi_{\alpha,\beta} \left( N \left( Y_t^{i,N} - Y_t^{j,N} \right) \right) \\
&= \frac{1}{2N^d} \sum_{\alpha,\beta=1}^d \sum_{i=1}^{K_{\bar{\rho}}} \partial_{\alpha} \partial_{\beta} \varphi \left( Y_t^{j,N} \right) \sum_{j: |Y_t^{i,N} - Y_t^{j,N}| \leq \frac{r}{N}} \psi_{\alpha\beta} \left( N \left( Y_t^{i,N} - Y_t^{j,N} \right) \right). \\
&\quad \rightarrow \frac{1}{2} \int \partial_{\alpha} \partial_{\beta} \varphi (x) \Psi_V^{\alpha,\beta} (\rho_t(x)) dx.
\end{aligned}$$

Summing up, passing to the limit in the weak formulation

$$\begin{aligned}
d \langle \varphi, S_t^N \rangle &= - \left\langle \nabla \varphi \int_{\mathbb{T}^d} \nabla V_N (\cdot - y) S_t^N (dy), S_t^N \right\rangle dt + \\
&\quad + \frac{\epsilon^2}{2} \langle \Delta \varphi, S_t^N \rangle dt + dM_t^{N,\varphi}
\end{aligned}$$

we get the weak formulation of  $\partial_t \rho = \frac{\epsilon^2}{2} \Delta \rho + \sum_{\alpha,\beta=1}^d \Delta \Psi_V^{\alpha,\beta} (\rho)$ .



## Manipulation of the non linear term

By an isotropy argument

$$\partial_t \rho = \frac{\epsilon^2}{2} \Delta \rho - \Delta \Psi_V(\rho),$$

whre  $\Psi_V(\rho) = \Psi_V^{\alpha,\alpha}(\rho) = \Psi_V^{1,1}(\rho)$ . We reformulate the PDE as

$$\partial_t \rho = \Delta P_V(\rho), \quad P_V(\rho) = \frac{\epsilon^2}{2} \rho - \Psi_V(\rho).$$

- [Varadhan, '91], for  $d = 1$  and purely repulsive potential,
- [Uchiyama, '00], for repulsive potential and repulsive- attractive potential.
- [Flandoli, Leocata, Ricci, '20] investigations on the explicit form of  $\Psi_V(\rho)$ .

## Investigation on $\Psi_V$

Our aim is to obtain some insights on  $\Psi_V$

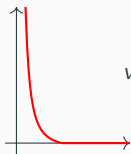
$$\Psi_V(\rho) = \lim_N \mathbb{E}_{\mu_G} \left[ \frac{1}{Nd} \sum_{i,j=1}^{K_{\bar{\rho}}} \partial_1 V(x^i - x^j) (x^i - x^j)_1 \right]$$

Two different tools:

- heuristic arguments based on intuitions on the Gibbs measure;
- numerics. By ergodicity property, we can produce realizations of such measure  $\mu_G^\rho$  by simulations on large times of a SDE whose invariant measure is  $\mu_G^\rho$ :

$$dX^i = - \sum_{j \neq i}^{K_{\bar{\rho}}} \nabla U(X_t^j - X_t^i) dt + \epsilon dB_t^i \quad (1)$$

# Heuristical investigation on $\Psi_V$ , Repulsive case, $d = 1$



$$V(|x|) = \left( \frac{1}{|x|^\alpha} - c \right) \mathbb{1}_{\{|x| \leq r\}},$$

Let us imagine realizations of points according to such measure. We look for an equilibrium configuration  $(\bar{x}^i, N)_{i=1, \dots, N}$  for a system of deterministic ODEs satisfying

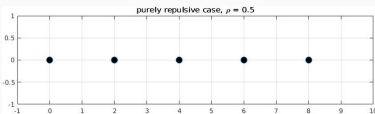
$$\dot{x}_t^{i, N} = - \sum_{j=1, j \neq i}^{K_{\bar{\rho}}} V'(x_t^{j, N} - x_t^{i, N}) \quad (2)$$

Then we approximate,

$$\mu_G \approx \frac{1}{N} \sum_{i=1}^{K_{\bar{\rho}}} \delta_{\bar{x}^i}, \quad \text{with } \bar{x}^i = \frac{i}{\bar{\rho}}, \quad i = 1, \dots, K.$$

# Heuristical investigation on $\Psi_V$ , Repulsive case, $d = 1$

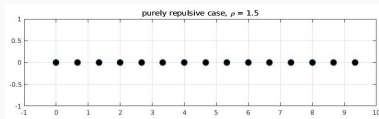
In the case of low density,  $\bar{\rho} < 1/r$



$$\begin{aligned}
 & -\frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} \sum_{j=1}^{K_{\bar{\rho}}} V' \left( \left| \bar{x}^{i,N} - \bar{x}^{j,N} \right| \right) \left| \bar{x}^{i,N} - \bar{x}^{j,N} \right| = \\
 & = -\frac{1}{N^d} \sum_{i=1}^{K_{\bar{\rho}}} \sum_{j=1}^{K_{\bar{\rho}}} V' \left( \left| \frac{i-j}{\bar{\rho}} \right| \right) \left| \frac{i-j}{\bar{\rho}} \right| = \\
 & = -\bar{\rho} \sum_{j=1}^{K_{\bar{\rho}}} V' \left( \frac{j}{\bar{\rho}} \right) \frac{j}{\bar{\rho}} = -\bar{\rho} \sum_{j=1}^{K_{\bar{\rho}}} \frac{\bar{\rho}^{\alpha+1} j}{j^{\alpha+1}} \frac{j}{\bar{\rho}} \mathbb{1}_{\left| \frac{j}{\bar{\rho}} \right| \leq r} = 0
 \end{aligned}$$

# Heuristical investigation on $\Psi_V$ , Strong Repulsive case, $d = 1$

In the case of high density,  $\bar{\rho} > 1/r$



$$\bar{\rho} \sum_{i=1}^K \frac{\bar{\rho}^{\alpha+1}}{i^{\alpha+1}} \frac{i}{\bar{\rho}} \mathbb{1}_{\left|\frac{i}{\bar{\rho}}\right| \leq r} = \bar{\rho}^{1+\alpha} \sum_{i=1}^{\lfloor \bar{\rho} r \rfloor} \frac{1}{i^{\alpha}}$$

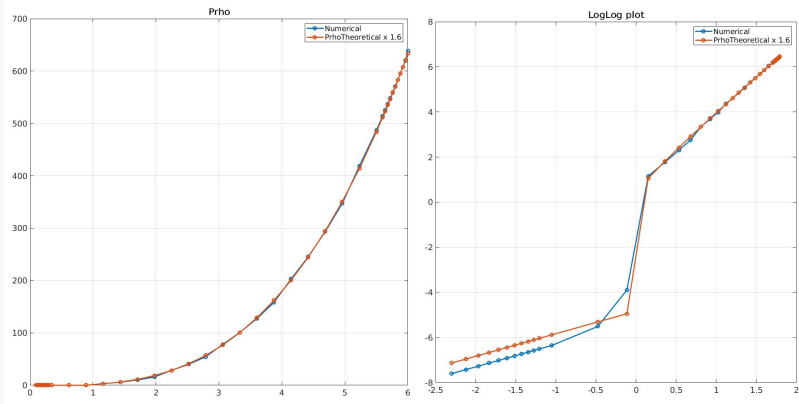
Since  $\sum_{i=1}^{\bar{\rho} r} \frac{1}{i^{\alpha}} \approx C_{\alpha}^1 + \frac{C_{\alpha}^2}{\bar{\rho}^{\alpha-1}}$ , we conclude

$$\bar{\rho}^{1+\alpha} \sum_{i=1}^{\lfloor \bar{\rho} R_1 \rfloor} \frac{1}{i^{\alpha}} = C_{\alpha}^1 \bar{\rho}^{\alpha+1} + C_{\alpha}^2 \bar{\rho}^2$$

Summing up

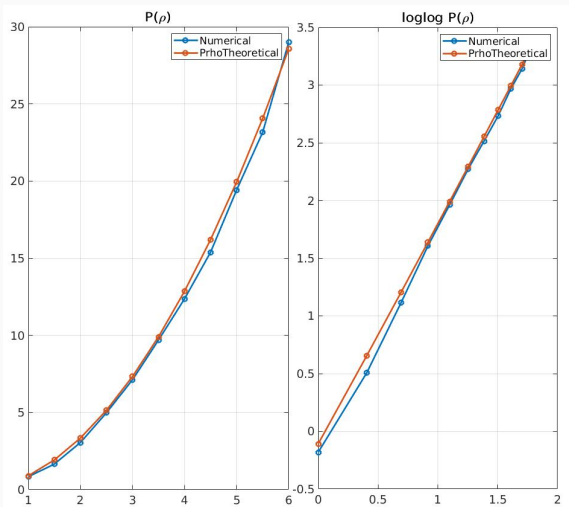
$$-\Psi_V(\bar{\rho}) = \begin{cases} 0 & \text{if } \bar{\rho} < r \\ C_{\alpha}^1 \bar{\rho}^{\alpha+1} + C_{\alpha}^2 \bar{\rho}^2 & \text{if } \bar{\rho} \geq r \end{cases}$$

# Numerical investigation on $\Psi_V$ , Strong Repulsive case, $d = 1$ , $\alpha = 2$



If  $f(x) = x^\alpha$ , then  $\log(f(x)) = \alpha \log(x)$ . By a change of variable  $y = \log(x)$ ,  $\log(f(x)) = \alpha y$ . Then  $\alpha$  will play the role of the angular coefficient in the loglog plot.

# Numerical investigation on $\Psi_V$ , Integrable potential, $d = 1$



## Repulsive-Attractive case

By a similar argument, we approximate  $\Psi_V(\rho)$  in the case of attractive-repulsive potential,  $V(x) = \left( \frac{R^\alpha}{\alpha|x|^\alpha} - \frac{R^\beta}{\beta|x|^\beta} + C \right) \mathbb{1}_{|x| \leq r}$ ,

$$\Psi_V(\rho) = \begin{cases} 0 & \rho < r, \\ C_\alpha \rho^{1+\alpha} - C_\beta \rho^{1+\beta} - C_{\alpha,\beta} \rho^2 & \rho \geq r \end{cases}$$



# First-order Approximation

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Let's go back to the Mean Field case, where particles interact through a interaction kernel,  $K$ .

### Motivating Example

Consider  $\mu_0^N \rightarrow 0$ , then the MF approximation does not hold. What is the equation to consider in this case?

By the Mean Field Limit:

$$\mu_t^N = \rho_t + \frac{1}{\sqrt{N}}.$$

By the Central Limit Theorem:

$$\mu_t^N = \rho_t + \frac{1}{\sqrt{N}}\eta_t + o\left(\frac{1}{\sqrt{N}}\right).$$

Our aim is to derive an equation furnishing a more detailed description of  $\mu_t^N$  with respect to  $\rho_t$ .

Up to our knowledge (Suggestions are welcome)

- Diffusion Approximation in Ethier-Kurtz,
- Derivation of the first-order approximation in [Chevallier, Ost'20] for Hawkes Process.

Does  $\rho_t^N := \rho_t + \frac{1}{\sqrt{N}}\eta_t$  satisfy an equation (at least at a formal level)?

$$\begin{aligned}d\rho^N(t) &= \Delta\rho(t) + \operatorname{div}((K * \rho(t))\rho(t)) + \frac{1}{\sqrt{N}}\Delta\eta(t) + \\ &+ \frac{1}{\sqrt{N}}\operatorname{div}((K * \eta(t))\rho(t)) + \frac{1}{\sqrt{N}}\operatorname{div}((K * \rho(t))\eta(t)) + dW_t[\rho_t] \\ &= \Delta\rho^N(t) + \operatorname{div}(F(\rho(t))) + \operatorname{div}\left(DF[\rho]\left(\frac{1}{\sqrt{N}}\eta(t)\right)\right) + dW_t[\rho_t]\end{aligned}$$

where  $W_t$  is a gaussian process on some negative sobolev Space  $(W_0^{-(4+2D,D)}(\mathbb{R}^d), [\text{Meleard '96}])$  with covariance

$$\mathbb{E}[W_t(\varphi)W_s(\psi)] = \int_0^{t \wedge s} \rho_s \nabla\varphi \cdot \nabla\psi ds$$

By Taylor expansion in infinite dimension,

$$F\left(\rho(t) + \frac{1}{\sqrt{N}}\eta(t)\right) = F(\rho(t)) + DF[\rho]\left(\frac{1}{\sqrt{N}}\eta(t)\right) + \frac{1}{2}D^2F[\rho]\left(\frac{1}{\sqrt{N}}\eta(t), \frac{1}{\sqrt{N}}\eta(t)\right)$$

Then,

$$\begin{aligned}d\rho^N(t) &= \Delta\rho^N(t) + \operatorname{div}(F(\rho(t))) + \operatorname{div}\left(DF[\rho]\left(\frac{1}{\sqrt{N}}\eta(t)\right)\right) + dW_t[\rho_t] \\ &= \Delta\rho^N(t) + \operatorname{div}(F(\rho^N(t))) + dW_t[\rho_t] - \frac{1}{2N}\operatorname{div}((K * \eta(t))\eta(t))\end{aligned}$$

in other terms,  $\rho_t^N$  is almost a solution of a SPDE. "Almost" because there is an error term of order  $o\left(\frac{1}{\sqrt{N}}\right)$

Our purpose is to study the SPDE,

$$du^N(t) = \Delta u^N(t)dt + \operatorname{div} \left( F(u^N(t)) \right) dt + dW_t[\rho_t^N]$$

and to prove that  $v^N$  is in some sense close to  $u^N$ . Indeed,  $u^n$

$$\begin{aligned} \mu_t^N &= u_t^N + \left( \rho_t + \frac{1}{\sqrt{N}}\eta_t - u_t^N \right) + o\left(\frac{1}{N}\right) \\ &= u_t^N + \left( \rho_t + \frac{1}{\sqrt{N}}\eta_t - u_t^N \right) + o\left(\frac{1}{N}\right) = u_t^N + o\left(\frac{1}{N}\right) \end{aligned}$$

$u_t^N$  is close to  $\mu_t^N$  at least as  $v_t^N$ , but differently from  $v_t^N$ , it satisfies an equation.

## Difficulties:

- Is the Gaussian Process a Stochastic Integral in infinite dimension,  $W_t = \text{div} \left( \sum_k \int_0^T \sqrt{v_t^N} e_k d\beta_t^k \right)$ ?
- If the previous is true, the process  $W_t$  will depend on  $\sqrt{u_t^N}$ . So  $u_t^N \geq 0$  and in particular it must be a function.
- Well-posedness of  $W_t$ :

$$\begin{aligned} \mathbb{E} \left\| \int_0^t \nabla \cdot e^{(t-s)\Delta} \sum_k \sqrt{v_t^N} e_k d\beta_t^k \right\|_{L^2}^2 &\leq \\ &\leq \int_0^t \frac{c}{(t-s)} \sum_k \left\| \sqrt{v_t^N} e_k \right\|_{L^2}^2 dt = \infty \end{aligned}$$

$\Rightarrow v_t^N$  should be a regular function.

•

$$\mathbb{E} \left\| v_t^N \right\|_{L^2}^2 \leq \int_0^t \frac{c}{\sqrt{t-s}} \mathbb{E} \left\| v_s^N (K * v_s^N) \right\|_{L^2}^2 dt$$

- The natural space for  $v_t^N$  seems to be the space of fluctuations. But this space is not compatible with all the requirements above.

Thanks for the attention!