



EFFECTIVE
SEDIMENTATION SPEED
OF RANDOM SUSPENSIONS

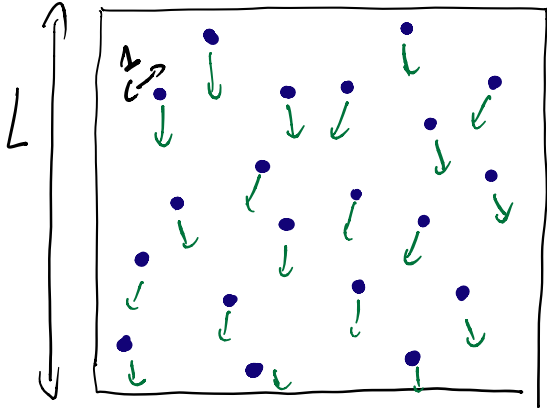
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Collective behavior of particles in fluids, Dec 14-17, 2020

Introduction

times $\Theta_L, L \gg 1$

→ size 1



Model: heavy rigid particles fall by gravity in a Stokes fluid

Setting: no inertia (steady Stokes)

Question: analysis of the map

$\{x_i\}$ → $\{v_i\}$
(positions of particles) (velocity of particles)

Interested in statistical rather than pointwise properties

Assume $\{x_i\}$ are distributed according to \mathbb{P}_L

what can we say on the distribution of $\{v_i\}$?

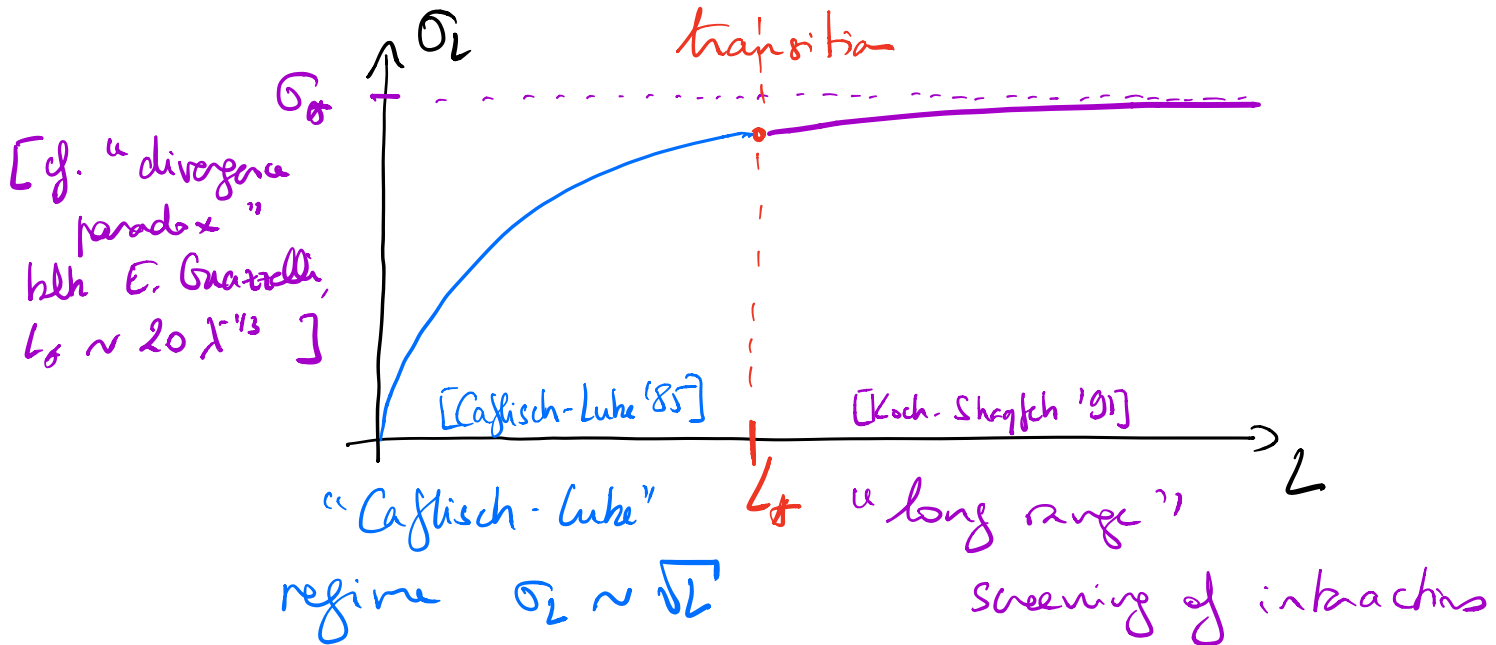
Observables: $\bar{V}_L = \mathbb{E}_L(v_i)$, $\sigma_L^2 = \mathbb{E}_L(|v_i - \bar{V}_L|^2)$

↑ effective sedimentation speed

↑ variance of the sedimentation speed

Physical insight from experiments (dimension $d=3$)

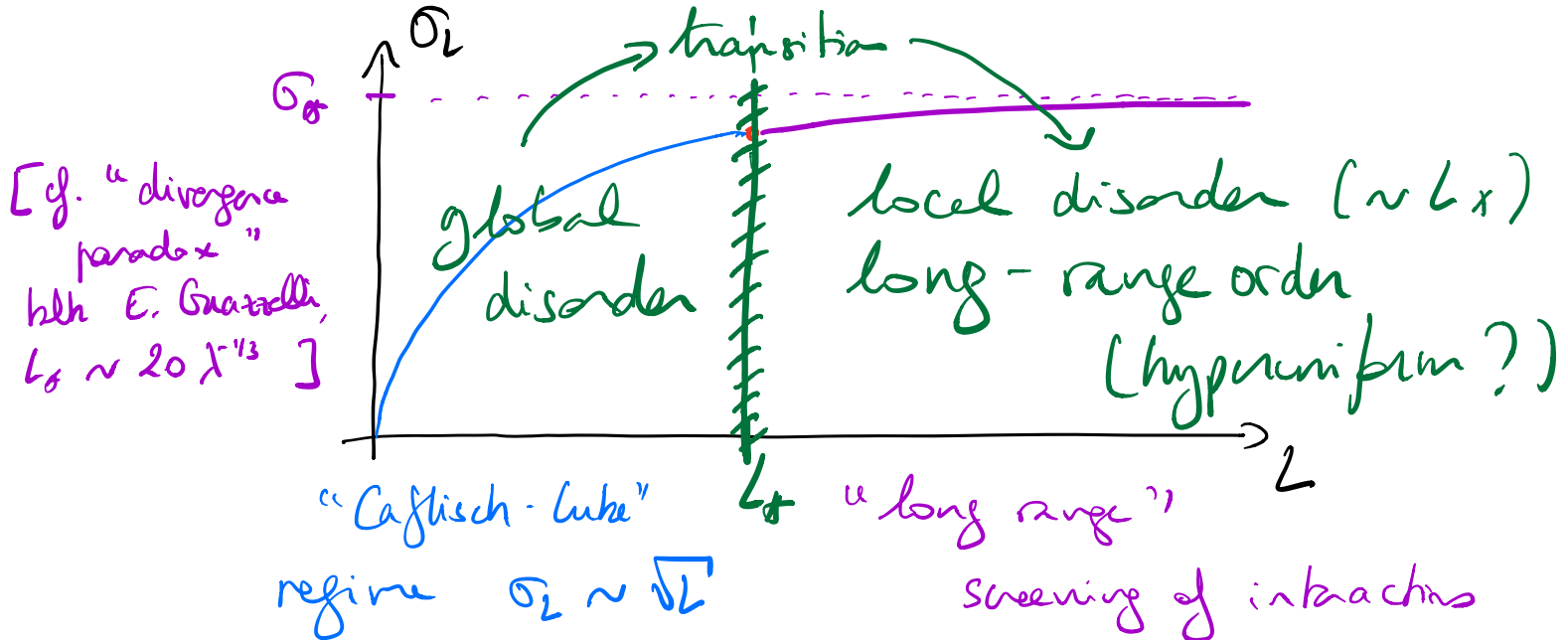
- Effective sedimentation speed: $\bar{v}_L \rightarrow \bar{v}$ as $L \rightarrow \infty$
- Variance of the sedimentation speed σ_L^2 :



- Aim of the talk:
- ① Identify a measure P_L likely to yield such a transition
 - ② Rigorously prove this behavior

Physical insight from experiments (dimension $d=3$)

- Effective sedimentation speed: $\bar{v}_L \rightarrow \bar{v}$ as $L \rightarrow \infty$
- Variance of the sedimentation speed:



- Aim of the talk:
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 - ② Rigorously prove this behavior

Physical relevance?

- Assumptions
- $\inf_{n \neq m} d(I_n, I_m) > 0$: positive minimal distance
 - any shape, polydispersity etc (crucial input: P_L)

E. Guazzelli's lecture

- 1941 Burgers : problem !
- 1972 Batchelor \bar{v} well-defined
- 1985 Geflick-Luke $\text{Var}[V_i] \sim L$
- 1991 Koch-Schlichter $\text{Var}[V_i] \sim L$
with long-range order
- 1997 Segé-Herstzheimmer - Chaikin
- 2017 Goldfried - Diamond - Wilton
- ...

Based on linear analysis
and various stochastic cancellations

Aim of the talk

Show why the linear
analysis yields
a good description
of the nonlinear
long-range multiphase
interactions, even
for hyperuniform structures

Outline of the talk

- (I) Equations & difficulties
- (II) Linear analysis: CL versus screening
a matter of stochastic cancellations
- (III) (Hyperuniform) functional calculus
CLT scaling, and more
- (IV) nonlinear analysis: a matter of locality
- (V) Main results and main steps of the proof
 - + (Hyperuniform) variance inequality
 - + sensitivity calculus and duality
 - + annealed Calderón-Zygmund estimates
for the "colloidal Stokes equation"

(I) Equations and difficulties

(I) Equations & difficulties

$$\left\{ \begin{array}{ll}
 -\Delta \varphi_L + \nabla \Pi_L = -d_L e & \mathcal{O}_L \setminus I_L \\
 \operatorname{div} \varphi_L = 0 & \mathcal{O}_L \setminus I_L \\
 \mathbb{D}(\varphi_L) = 0 & I_L = U I_n \\
 e | I_n | + \int_{\partial I_n} \sigma(\varphi_L, \Pi_L) \nu = 0 & (1) \quad \forall n \\
 \int_{\partial I_n} \mathbb{H} \nu \cdot \sigma(\varphi_L, \Pi_L) \nu = 0 & (2) \quad \forall n, \forall \mathbb{H} \in \mathbb{M}^{\text{skew}}
 \end{array} \right.$$

φ_L : { velocity of the fluid $\mathcal{O}_L \setminus I_L$ symmetrized gradient
 { velocity of the particles I_n (rigid body: $\mathbb{D}(\varphi_L) = 0$)

Π_L : pressure (fluid is incompressible $\operatorname{div} \varphi_L = 0$)

$-d_L e$: backflow (pressure in experiments), $\sigma = 2\mathbb{D}(\varphi_L) - \Pi_L \operatorname{Id}$
 (cf. talk by E. Guazzelli) stress tensor

(1): free equilibrium on particles (gravity & fluid)

(2): momentum equilibrium on particles (fluid)

$$\varphi_L(x) = \mathbb{H}_L(x - a_n) + v_n$$

(I) Equations & difficulties

Well-posedness is standard (with $\alpha_L = \frac{\lambda_L}{1-\lambda_L}$, $\alpha_L = L^{-d} \sum_n |\mathbb{I}_n|$)

Reformulation of the eqns in \mathcal{O}_L :

$$-\Delta \varphi_L + \nabla(\pi_L \mathbb{1}_{\mathcal{O}_L \setminus \mathbb{I}_L}) = -\alpha_L e \mathbb{1}_{\mathcal{O}_L \setminus \mathbb{I}_L} - \sum_n \int_{\partial \mathbb{I}_n} \sigma(\varphi_L, \pi_L) \nu$$

Test with φ_L and use BC + rigid body motion

$$\int_{\mathcal{O}_L} |\nabla \varphi_L|^2 = -\alpha_L e \cdot \int_{\mathcal{O}_L \setminus \mathbb{I}_L} \varphi_L - \sum_n \int_{\partial \mathbb{I}_n} \varphi_L \cdot \sigma(\varphi_L, \pi_L) \nu$$

$$= -\alpha_L e \int_{\mathcal{O}_L \setminus \mathbb{I}_L} \varphi_L + \sum_n e \cdot \int_{\mathbb{I}_n} \varphi_L$$

$$= (1 + \alpha_L) \sum_n e \cdot \int_{\mathbb{I}_n} \varphi_L$$

$$\Rightarrow V_L := \frac{1}{N} \sum_n e \cdot \int_{\mathbb{I}_n} \varphi_L = \frac{1}{\alpha_L} \int_{\mathcal{O}_L} |\nabla \varphi_L|^2$$

By Poincaré's inequality: $|V_L| \lesssim L^2$ (pb encountered by Burgers '41)

NEED OF STOCHASTIC CANCELLATIONS

(II) Linear analysis:

Cajlisch - Luke versus screening

(cf. talks by E. Guazzelli

and the papers by

Batchelor	'72
Cajlisch-Luke	'85
Koch-Shaughfeh	'81

)

(II) Linear analysis

Other formulation: $\varphi_L = \frac{1}{1-\lambda_L} P_L \varphi_L^0$
(cf. R. Höfer)

+ P_L orthogonal project from $H\text{div}(\mathcal{O}_L) \rightarrow \{\varphi \in H\text{div} \mid \mathcal{D}(\varphi) = 0 \text{ on } \mathbb{T}^d\}$.

+ φ_L^0 solves $-\Delta \varphi_L^0 + \nabla \pi_L^0 = (\mathbb{1}_{\mathcal{I}_L} - \lambda_L) e$, $\text{div} \varphi_L^0 = 0$ \mathcal{O}_L .

→ equation for φ_L^0 is linear wrt P_L

→ P_L : multibody nonlinear hydrodyn. interactions

$$\int_{\mathcal{O}_L} |\nabla \varphi_L|^2 \leq \int_{\mathcal{O}_L} |\nabla \varphi_L^0|^2$$

By linearity, $\varphi_L^0(x) = \sum_n U_L(x - x_n)$

where $-\Delta U_L + \nabla P_L = (\mathbb{1}_B - L^{-d} |B|) e$ \mathcal{O}_L

is the (locally averaged) periodic Stokeslet.

Hence, $\lambda_L |e| |\nabla_L^0| = \mathbb{E}_L [|\nabla \varphi_L^0|^2] = \mathbb{E}_L \left[\left| \sum_n \nabla U_L(x_n) \right|^2 \right] = |\text{var}_L \left[\sum_n \nabla U_L(x_n) \right]|$

$$(\sigma_L^0)^2 = |\text{var}_L [\varphi_L^0]| = |\text{var}_L \left[\sum_n U_L(x_n) \right]|$$

(II) Linear analysis : control of variances

To have a hint on what to expect,

look at **variance** of number of points in a box Q_L :

→ **Poisson point process** $P = \{x_n\}_n$, intensity λ

$$\text{Var} [L^{-d} \# (P \cap Q_L)] = \lambda^2 L^{-d}$$

↑ CLT scaling

→ **Hyperuniform point process** $P = \{x_n\}_n$

$$\text{Var} [L^{-d} \# (P \cap Q_L)] \sim L^{-d-2}$$

↑ better than CLT

→ **suppression of number fluctuations**

In this section : alternative formulation in terms
of variance estimates

(II) Linear analysis: Caflisch-Luke behavior

Assumption: P_L is weakly correlated in the sense that

$$\forall \text{ function } \varphi, \quad \text{var}_L \left[\sum_n \varphi(x_n) \right] \lesssim \rho_L^2 \int_{\mathcal{Q}} |\varphi|^2$$

$\mathbb{E}[L^{-d} \# \langle x_n \rangle]$
 (intensity of point process)

\uparrow CLT scaling

Let $\varphi: \mathcal{Q} \rightarrow \mathbb{R}_+$ and set $\varphi_L := L^{-d} \varphi(\cdot/L)$
 with $\int_{\mathcal{Q}} \varphi = 1$, $\int_{\mathcal{Q}} \varphi^2 \lesssim 1$ so that $\int_{\mathcal{Q}_L} \varphi_L = 1$, $\int_{\mathcal{Q}_L} \varphi_L^2 \lesssim L^{-d}$

Then, since $|U_L(x)| \sim |x| (1+|x|)^{2-d}$, $|\nabla U_L(x)| \sim |x| (1+|x|)^{1-d}$

$$\lambda_L |x| \bar{\sigma}_L^0 \lesssim \rho_L^2 \int_{\mathcal{Q}_L} |\nabla U_L|^2 \lesssim \rho_L^2 |x|^2 \left| \begin{array}{l} L \quad d=1 \\ \log L \quad d=2 \\ 1 \quad d>2 \end{array} \right. \quad \text{Batchelor}$$

$$(\sigma_L^0)^2 \lesssim \rho_L^2 \int_{\mathcal{Q}_L} |U_L|^2 \lesssim \rho_L^2 |x|^2 \left| \begin{array}{l} L^3 \quad d=1 \\ L^2 \quad d=2 \\ L \quad d=3 \\ \log L \quad d=4 \\ 1 \quad d>4 \end{array} \right. \quad \text{Caflisch-Luke}$$

(II) Linear analysis: additional screening

Assumption: P_L is weakly correlated and hyperuniform in the sense

$$\text{that } \forall \text{ function } \varphi, \quad \text{var}_L \left[\sum_n \varphi(x_n) \right] \lesssim \rho_L^2 \int_{\mathcal{Q}_L} |\nabla \varphi|^2$$

$\mathbb{E}[L^{-d} \# \langle x_n \rangle]$
(intensity of point process)

↑ better than
CLT scaling

$$\left[\begin{array}{l} L^d \varphi : \mathcal{Q} \rightarrow \mathbb{R}_+ \quad \text{and set } \varphi_L := L^{-d} \varphi(\cdot/L) \\ \text{with } \int_{\mathcal{Q}} \varphi = 1, \int_{\mathcal{Q}} \varphi^2 \leq 1 \quad \text{so that } \int_{\mathcal{Q}_L} \varphi_L = 1, \int_{\mathcal{Q}_L} |\nabla \varphi_L|^2 \leq L^{-d-2} \end{array} \right]$$

Then, since $|\nabla U_L(x)| \sim |x| (1+|x|)^{1-d}$, $|\nabla^2 U_L(x)| \sim |x| (1+|x|)^{-d}$

$$\lambda_L |x| \bar{V}_L^0 \lesssim \rho_L^2 \int_{\mathcal{Q}_L} |\nabla^2 U_L|^2 \lesssim \rho_L^2 |x|^2 \quad (d \geq 1)$$

$$(\sigma_L^0)^2 \lesssim \rho_L^2 \int_{\mathcal{Q}_L} |\nabla U_L|^2 \lesssim \rho_L^2 |x|^2 \quad \left| \begin{array}{l} L \quad d=1 \\ \log L \quad d=2 \\ 1 \quad d>2 \end{array} \right. \quad \begin{array}{l} \text{screening} \\ \text{Koch \& Sierpinski} \end{array}$$

(II) Linear analysis: conclusion

- (1) \bar{V}_L well defined : $|d > 2$ for "CLT scaling"
 $P_L: \int_{\mathcal{Q}} |\nabla \psi_L|^2 \leq \int_{\mathcal{Q}_L} |\nabla \psi_L|^2$ $d \geq 1$ for "hyperuniform scaling"
- (2) σ_L : no rigorous bound yet (P_L in the way)
 σ_L gives a hint : |Cagliostro-Luke for "CLT scaling"
screening for "hyperuniform scaling"
- (3) A matter of stochastic cancellations
→ when do we have $\text{var}_L[\sum_n \mathcal{P}(a_n)] \stackrel{?}{\sim} P_L^2 \left\{ \begin{array}{l} \int_{\mathcal{Q}_L} |g|^2 \\ \int_{\mathcal{Q}_L} |\nabla g|^2 \end{array} \right.$
- where does L_* enter the picture (condition)?
- P_L is nonlinear: it would be convenient to have
"nonlinear versions" of these inequalities
- where would these stochastic cancellations
come from physically?

(III) Standard and hyperuniform
functional calculus

(III) Standard functional calculus

Standard functional calculus:

$X = X(P)$ with $P = \{\lambda_n\}_{n \in \mathbb{N}}$ **Poisson point process**

$$\text{then } \text{var}[X] \leq \int_{\mathbb{R}^d} \mathbb{E} \left[\left(\partial_{B(x)}^{\text{osc}} X \right)^2 \right] dx$$

$$\text{with } \partial_D^{\text{osc}} X = \cosup_{P' = P} X(P') - \text{osinf}_{P'' = P} X(P'') \\ \text{on } \mathbb{R}^d \setminus D \quad \text{on } \mathbb{R}^d \setminus D$$

For a **hardcore Poisson process**, multiscale version:

$$\text{var}[X] \leq \int_1^{+\infty} \int_{\mathbb{R}^d} \mathbb{E} \left[\left(\partial_{B_\ell(x)}^{\text{osc}} X \right)^2 \right] dx e^{-\ell} d\ell$$

Example: $X_L(P) = \sum_n \mathbb{1}_{B_L(\lambda_n)}$

$$\text{var}[X_L] \leq \int_1^{\infty} \int_{\mathbb{R}^d} \mathbb{E} \left[\sup_{P' = P} \left(\sum_n \mathbb{1}_{B_L(\lambda_n)} \mathbb{1}_{B_\ell(x)} \right)^2 \right] dx e^{-\ell} d\ell \\ \text{on } \mathbb{R}^d \setminus B_\ell(x) \quad \approx \int_{\mathbb{Q}_L \cap B_\ell(x)} |\mathbb{1}_{B_L}|$$

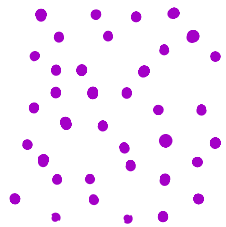
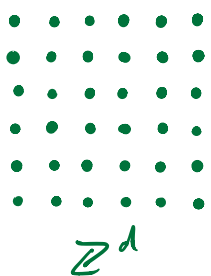
$$\leq \int_1^{\infty} \int_{\mathbb{R}^d} |\mathbb{Q}_L \cap B_\ell(x)| \int_{B_\ell(x) \cap \mathbb{Q}_L} |\mathbb{1}_{B_L}|^2 dx e^{-\ell} d\ell \leq L^{-d} \int_{\mathbb{Q}_L} |\mathbb{1}_{B_L}|^2$$

$$[\ell=1 : \int_{\mathbb{Q}_L} |\mathbb{1}_{B_L}|^2, \ell \geq 1 : \ell^{2d} e^{-\ell} \int_{\mathbb{Q}_L} |\mathbb{1}_{B_L}|^2]$$

(III) Hyperuniform functional calculus ($L_0 = 1$)

Hyperuniform functional calculus:

Elementary example: $\mathcal{P} = \{x + u_x, x \in \mathbb{Z}^d\} = \{x_n\}_n$
 with $\{u_x\}_x$ iid, uniformly distributed in $(-1/3, 1/3)^d$



\mathcal{P} (random displacement model)

Efron-Stein inequality: $\text{var}[X] \leq \sum_y \mathbb{E}[(X(\mathcal{P}) - X(\mathcal{P}_y))^2]$,

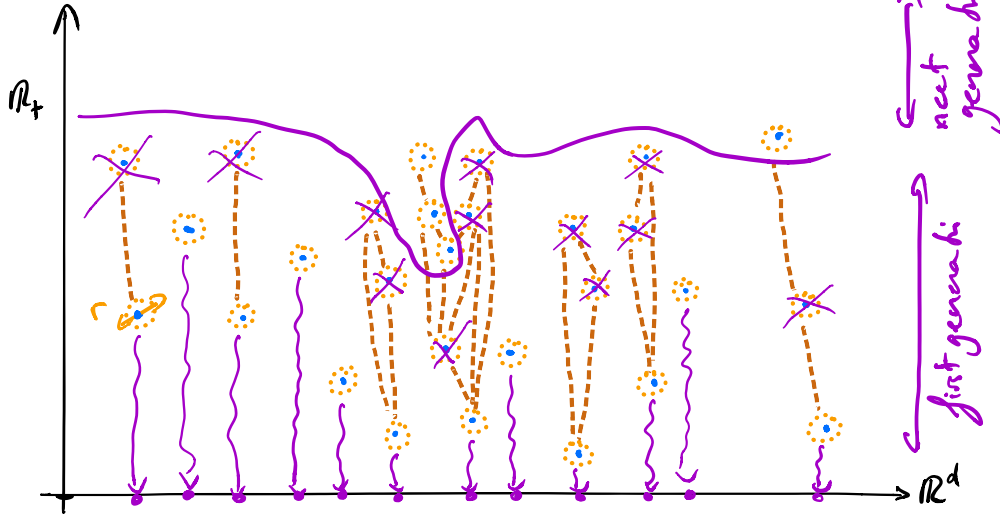
where $\mathcal{P}_y = \{x + u_x, x \in \mathbb{Z}^d\}$ with $\begin{cases} u_{x,y} = u_x & \text{for } x \neq y \\ u_{y,y} & \text{independent of } \{u_x\}_x \end{cases}$

Example: $X_L(\mathcal{P}) = \sum_n \rho_L(x_n) = \sum_{x \in \mathbb{Z}^d} \rho_L(x + u_x)$

$$\begin{aligned} \text{var}[X_L] &\leq \sum_y \mathbb{E}[(\rho_L(y + u_y) - \rho_L(y + u_{yy}))^2] \\ &\leq \sum_y \mathbb{E}[\int_{\mathbb{Q}^d} |\rho_L|^2] = \int_{\mathbb{R}^d} |\rho_L|^2 \leq L^{-d-2} \int_{\mathbb{Q}} |\rho_L|^2 \end{aligned}$$

(II) A more sophisticated hyperuniform model ($L_* \gg 1$)

Step 0: Penrose graphical construction \mathcal{G}



$$\mathcal{P}_r^t(\mathbb{R}^d)$$

hardcore

$$= \mathcal{G}(\text{Poisson } \mathbb{R}^d \times (0, t])$$

$$\mathcal{P}_r^\infty(\mathbb{R}^d)$$

$$= \mathcal{G}(\text{Poisson } \mathbb{R}^d \times \mathbb{R}_+)$$

Random parking measure (packing)

Step 1: $\mathcal{P}_* := \mathcal{P}_{L_*}^\infty(\mathbb{R}^d)$ for hardcore radius $L_* \gg 1$

$\mathcal{V}_\theta :=$ Voronoi tessellation of $\mathbb{R}^d = \{C_n, n \in \mathbb{N}\}$

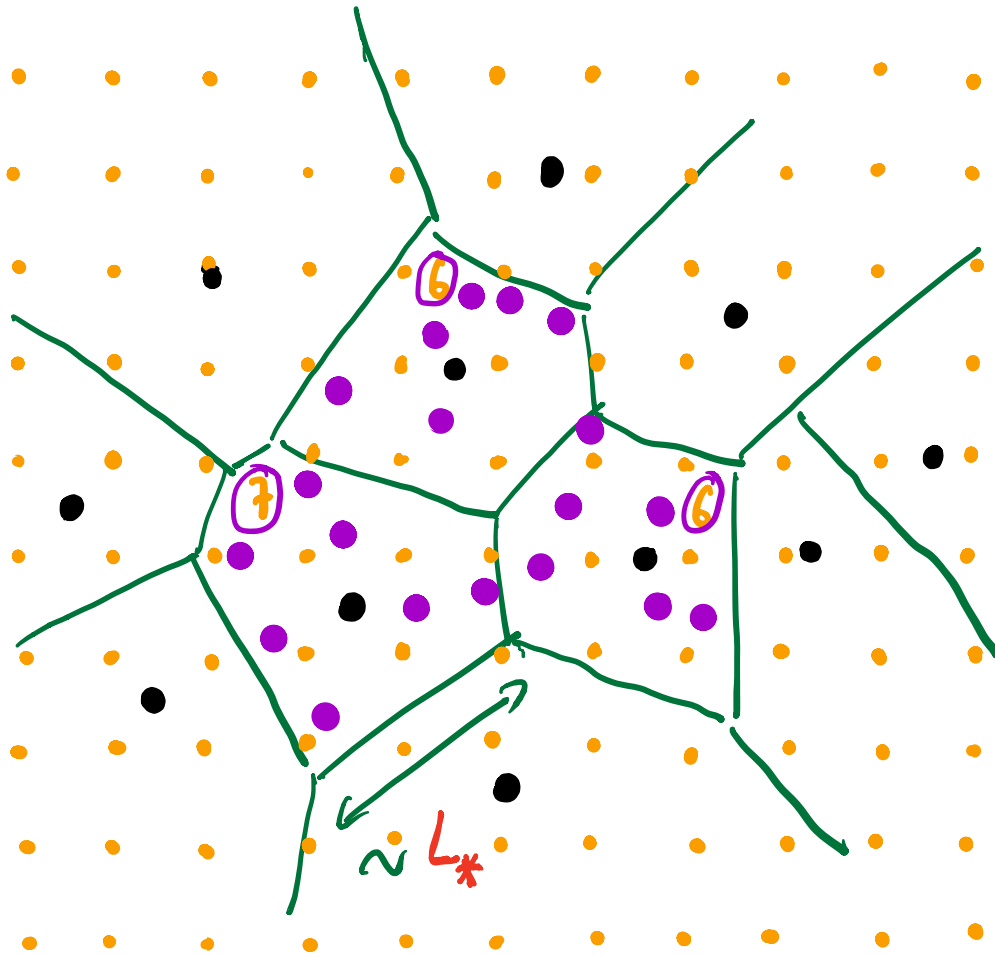
Property: $\exists \gamma$ s.t. $B(x_n, L_*) \subset C_n, |C_n| \leq \gamma L_*^d$

Step 2: $\mathcal{P}_2 = \{(x, t_x) \mid x \in \mathcal{P}_2^\infty(\mathbb{R}^d), t_x \text{ is arrival time}\}$

Set $N_n := \text{card}(C_n \cap \mathbb{Z}^d)$, and define \mathcal{P} via

$\mathcal{P}|_{C_n} := \{N \text{ points of } \mathcal{P}_1 \cap (C_n \times \mathbb{R}_+) \text{ that arrived first}\}.$

(III) A more sophisticated hyperuniform model



\mathbb{Z}^d

Voronoi: \mathcal{P}_*

\mathcal{P}

If X only depends on $\mathcal{P}|_{L \leq L^*}$, then usual spectral gap:

$$\text{Var}[X] \leq \int_0^{L^*} \int_{\mathbb{R}^d} \mathbb{E} \left[\left(\frac{\partial}{\partial \ell} \mathbb{E}[X] \right)^2 \right] dx e^{-\ell} d\ell$$

(III) A more sophisticated hyperuniform model

Multiscale hyperuniform variance inequality: $X = \sum_n \varphi(x_n; P)$

$$\text{Var}[X] \leq \int_{\mathbb{R}^d} \mathbb{E} \left[\left(\mathcal{D}_{\text{Be}(a)}^{\text{mov}} X \right)^2 \right] da$$

where $\mathcal{D}_{\text{Be}(a)}^{\text{mov}} = \sup_{P'} X(P') - \inf_{P''} X(P'')$
 $P' = \mathcal{P}$ on $\mathbb{R}^d \setminus \text{Be}(a)$ $P'' = \mathcal{P}$ on $\mathbb{R}^d \setminus \text{Be}(a)$
 $|P' \cap \text{Be}(a)| = |P \cap \text{Be}(a)|$ $|P'' \cap \text{Be}(a)| = |P \cap \text{Be}(a)|$
 ↑ same number of points ↑

Example: $X_L(P) = \sum_n \varphi_L(x_n)$, $L \gg L_0$

$$\text{var}[X_L] \leq L_0^d \int_{L_0}^{\infty} \int_{\mathbb{R}^d} \mathbb{E} \left[\sup_{\substack{x_i, x_{i'} \\ \in \text{Be}(a)}} \left| \sum_{i=1}^{N_{\text{Be}(a)}} \varphi_L(x_i) - \varphi_L(x_{i'}) \right|^2 \right] da e^{-\ell L_0^d} d\ell$$

$$\leq \int_{L_0}^{\infty} \int_{\mathbb{R}^d} \ell^2 \mathbb{1}_{\text{Be}(a) \cap \mathcal{Q}_{L_0} \neq \emptyset} \ell^{2d} \sup |\mathcal{D}_{\varphi_L}^{\varphi}|^2 da e^{-\ell L_0^d} d\ell$$

$$\leq L_0^{1+2d} \int_{L_0}^{\infty} \int_{\mathbb{R}^d} \mathbb{1}_{\text{Be}(a) \cap \mathcal{Q}_{L_0} \neq \emptyset} \sup |\mathcal{D}_{\varphi_L}^{\varphi}|^2 dx e^{-\ell} d\ell$$

$$\leq L_0^{1+d} \int_{L_0}^{\infty} L^d \sup |\mathcal{D}_{\varphi_L}^{\varphi}|^2 \leq L_*^{1+d} L^{-d-2}$$

(III) More about hyperuniformity

(1) Hyperuniformity yields improvement on CLT for (only?)
linear functionals of the random field

Counter-example: Random displacement model $\mathcal{P} = \{x + u_x, x \in \mathbb{Z}^d\}$

$$X_L = L^{-d} \sum_{x \in \mathbb{Z}^d \cap \Omega_L} \underbrace{d(\mathcal{P}, x)}_{\text{nonlinear}} = L^{-d} \sum_{x \in \mathbb{Z}^d \cap \Omega_L} \underbrace{|u_x|}_{\text{iid rv}}$$

we have $\text{var}[X_L] \sim L^{-d}$ not L^{-d-2}

(2) this section is **descriptive**: hyperuniform systems exist

→ why would a system be hyperuniform?

→ Talk by S. Sefery on Coulomb gases

→ for a systematic study:

→ Talk by S. Torquato

(IV) Nonlinear analysis :
a matter of locality

(IV) nonlinear analysis: a matter of locality

Summary: + long-range order (like hyperuniformity) combined



with linear analysis of sedimentation allows to

see both Caflisch-like and long-range screening



+ sedimentation is nonlinear (via the projector P_2)

+ hyperuniformity "only" improves linear functionals

Question: why would the linear analysis of the system be representative of the nonlinear system?

Remark: In stochastic homogenization of linear elliptic equations

in divergence and non-divergence form, assuming the



coefficients are hyperuniform does not improve the CLT scaling!

(IV) nonlinear analysis: an instructive example

Let a be random coefficients and consider

$$-\operatorname{div}(a \nabla \varphi) = (a - \mathbb{E}[a]) : \mathbb{E}$$

This is a **hybrid** between the corrector equations in homogenization of linear elliptic operators:

+ in divergence-form: $-\operatorname{div}(a \nabla \varphi) = \operatorname{div}(a e)$

+ in non-divergence form: $-a : \nabla^2 \varphi = (a - \gamma) : \mathbb{E}$

Functional calculus: replace a by $a + t \delta a$ and differentiate wrt t :

$$-\operatorname{div}(a \nabla \delta \varphi) = \underbrace{\delta a : \mathbb{E}}_{\text{linear response}} + \underbrace{\operatorname{div}(\delta a \nabla \varphi)}_{\text{nonlinear response (depends on } \varphi)}$$

Questions: order of magnitude of the contributions?
Effect of hyperuniformity?

(IV) nonlinear analysis: an instructive example

(1) Locality: Set G : Green's functn, sol of $-\text{div}(a \nabla G(x, y)) = \delta(x-y)$

Then: $\delta\varphi(x) = \int_{\mathbb{R}^d} G(x, y) \delta a(y) : E dy - \int_{\mathbb{R}^d} \nabla_z G(x, y) \cdot \delta a \nabla\varphi(y) dy$

Locality of each term: assume $\text{supp } \delta a = B(z)$ (and nothing else)

$$|\delta\varphi(x)| \sim \underbrace{G(x, z)}_{|x-z|^{2-d}} + \underbrace{|\nabla_z G(x, z)|}_{|x-z|^{1-d}} |\nabla\varphi(z)|$$

Cajlich-Luke

linear response \Rightarrow is less local than nonlinear response

Screening

(2) Hyperuniformity: In this case, δa must have vanishing average:

$$\int_{\mathbb{R}^d} G(x, y) \delta a(y) : E dy = \int_{B(z)} \underbrace{(G(x, y) - \int_{B(z)} G(x, \cdot))}_{\sim \nabla_z G(x, y) \sim |x-z|^{1-d}} \delta a(y) : E dy$$

linear response gets more local with hyperuniformity

For hyperuniform coefficients, the linear and the nonlinear responses have the same order of magnitude for this model

(IV) Main results

(V) Main results: statement

Theorem (Duerinckx - Gloria '20)

Let P_L be the L -periodization in law of the hyperuniform model of parameter $L_* \gg 1$. Then

(I) For all $\boxed{1 \leq L \ll L_*}$, we have

$$\frac{\bar{V}_L}{P_L |\ell|} \leq \begin{cases} 1 : d > 2 & \text{Batchelor} \\ \log L : d = 2 \\ L : d = 1 \end{cases}, \quad \sigma_L \leq |\ell| \begin{cases} 1 : d > 4 \\ (\log L)^{1/2} : d = 4 \\ L^{1/2} : d = 3 \\ L : d = 2 \\ L^{3/2} : d = 1 \end{cases}$$

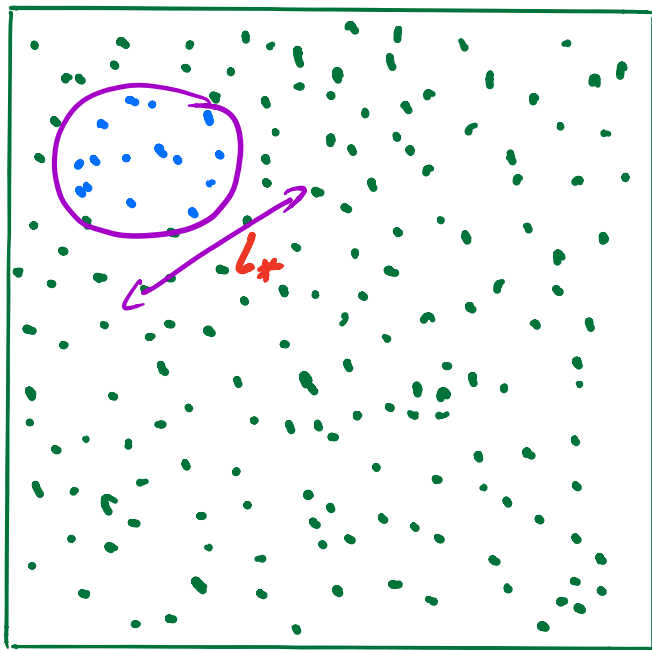
Caffisch-Luke

(II) For all $\boxed{L \gtrsim L_*}$, we have

$$\frac{\bar{V}_L}{P_L |\ell|} \leq \underset{\sim L_*}{1} : d \geq 1, \quad \sigma_L \leq \underset{\sim L_*}{|\ell|} \begin{cases} 1 : d > 2 \\ (\log L)^{1/2} : d = 2 \\ L^{1/2} : d = 1 \end{cases}$$

Screening

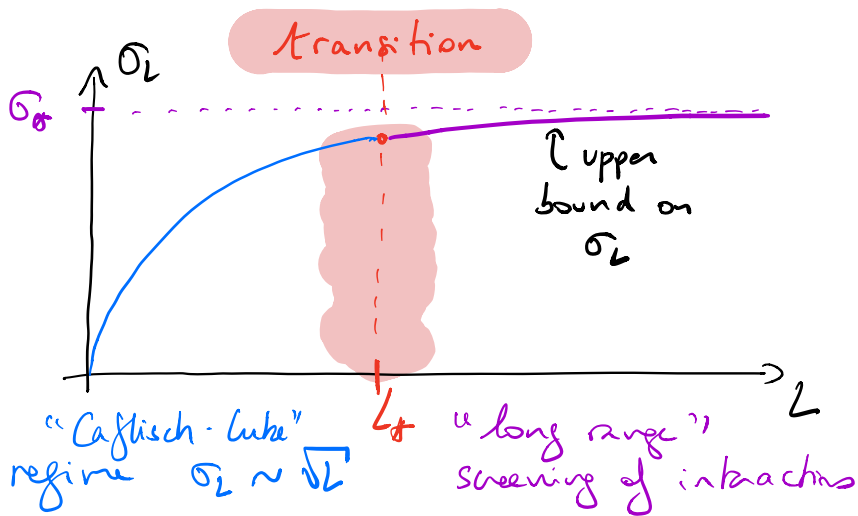
(V) Main results : interpretation $d=3$



"Hinch" - type model

like hard-core
Poisson
at scale $\leq L_*$

suppression
of number
fluctuations
at scale $\geq L_*$



Typical graph $L \mapsto \sigma_L = \sqrt{\text{Var}[V_L]}$
recovered as upper bound
(also lower bound for linear model)

(V) Main results: effective velocity

Starting point is identity

$$\int_{\mathcal{O}_L} |\nabla \varphi_L|^2 = (1 + \alpha_L) \int_{\mathcal{O}_L} \varphi_L \cdot \left(\sum_{\mathbf{n}} e(\mathbb{1}_{\mathbf{I}_n} - L^{-d} |\mathbf{B}|) \right)$$

(IPP)

$$= (1 + \alpha_L) \int_{\mathcal{O}_L} \nabla \varphi_L : \left(e \otimes \sum_{\mathbf{n}} \nabla (-\Delta)^{-1} (\mathbb{1}_{\mathbf{I}_n} - L^{-d} |\mathbf{B}|) \right)$$

(C.S.)

$$\lesssim (1 + \alpha_L)^2 |e|^2 \int_{\mathcal{O}_L} \left| \sum_{\mathbf{n}} \nabla (-\Delta)^{-1} (\mathbb{1}_{\mathbf{I}_n} - L^{-d} |\mathbf{B}|) \right|^2$$

linear analysis!

→ yields the **claimed scalings**
depending on $L \ll L_0, L \gtrsim L_0$

(even without functional-analytic version of ergodicity)

Subtle part: **fluctuations**

(V) Main results : bounds on the variance

Based on the combination of two propositions :

Proposition 1 For all $L \geq 1$, $g \in C_{\#}^{\infty}(\mathbb{Q}_L)^{d \times d}$, all $1 \leq R \leq L$,

and $q \geq 1$ and $1 < p < \infty$:

+ If $1 \leq L \ll L_{*}$ $(d \geq 2)$ ← Caflisch-Luke

$$\| \int_{\mathbb{Q}_L} g : \nabla \varphi_L \|_{L^{2p}(\Omega)}^2 \lesssim_p \| g \|_{L^{\frac{2d}{d-2}}(\mathbb{Q}_L)}^2$$

$$+ \| \langle \nabla \rangle^{\frac{1}{2}} g \|_{L^2(\mathbb{Q}_L)}^2 \| (1 + \int_{\mathbb{Q}_R} |\nabla \varphi_L|^{2q})^{1/q} \|_{L^p(\Omega)}$$

+ If $L \gtrsim L_{*}$,

$$\| \int_{\mathbb{Q}_L} g : \nabla \varphi_L \|_{L^{2p}(\Omega)}^2 \lesssim_p$$

$$\| \langle \nabla \rangle^{\frac{1}{2}} g \|_{L^2(\mathbb{Q}_L)}^2 \| (1 + \int_{\mathbb{Q}_R} |\nabla \varphi_L|^{2q})^{1/q} \|_{L^p(\Omega)}$$

Screening

stochastic integrability

(IV) Main results : bounds on the variance

Proposition 2 Let $\chi \in C_c^\infty(B)$, $\int_B \chi = 1$, set $\chi_r: x \mapsto r^{-d} \chi(x/r)$.

There exists $\eta_0 > 0$ such that for all $1 \leq r \leq R \leq L$,
 $1 \leq q \leq 1 + \eta_0$ and $p \geq 1$,

$$\left\| \left(1 + \int_{\mathbb{O}_R} |\nabla \varphi_L|^{2q} \right)^{1/q} \right\|_{L^p(\Omega)} \lesssim_\chi R^2 + \left\| \int_{\mathbb{O}_L} \chi_r \nabla \varphi_L \right\|_{L^p(\Omega)}^2$$

Proof of thm: (1) buckling on \mathbb{O}_R :

$$\begin{aligned} \mathbb{O}_R &\lesssim R^2 + \mathbb{O}_r \\ &\lesssim R^2 + \mathbb{O}_r + \mathbb{O}_r \mathbb{O}_1 \\ &\lesssim R^2 + r^{2-d} + r^{-d} R^{d(1-\nu_p)} \mathbb{O}_R \end{aligned}$$

reverse Jensen + stationarity

Take $r = R/2$, choose $R \gg 1$, and get $\mathbb{O}_R \lesssim 1$.

$$(2) \|\varphi_L(x) - \varphi_L(b)\|_{L^{2p}(\Omega)} \leq \|\nabla \varphi\|_{L^{2p}(\Omega)} + \left\| \int_{B(b)} \varphi_L - \int_{B(b)} \varphi_L \right\|_{L^{2p}(\Omega)}$$

$$\text{set } -\Delta h_L = |B|^{-1} \mathbb{O}_R - L^{-d} \text{ in } \mathbb{O}_L = \underbrace{\int_{\mathbb{O}_L} \nabla \varphi_L \cdot \nabla h_L(x)}_{\mathbb{O}_L}$$

(V) Main results: bounds on the variance

Proposition 2 solely relies on Caccioppoli's inequality

Proposition 1 relies on two ingredients:

- sensitivity calculus and use of hyperuniformity
- annealed Calderón-Zygmund estimates for colloidal suspensions

Argument 1: Let P_L and P_L' differ on $B(x)$

$$\text{Then: } \int_{\partial_L \Omega_L} g : (D\phi_L - D\phi_L') = -(\alpha_L - \alpha_L') e \cdot \int_{\Omega_L} v_L$$

$$+ (\alpha_L' + 1) \left(\sum_{x \in B(x)} e \cdot \int_{\Omega_L} v_L - \sum_{x' \in B(x)} e \cdot \int_{\Omega_L'} v_L \right)$$

[duality argument]

$$- \sum_{x \in B(x)} \int_{\partial \Omega_L'} (v_L - \int_{\Omega_L'} v_L) \cdot \sigma(\phi_L', \Pi_L' - c_L') \nu$$

$$\text{where } \begin{cases} -\Delta v_L + \nabla P_L = \nabla \cdot g & \& \operatorname{div} v_L = 0 \text{ in } \Omega_L \cup \Omega_L' \\ D(v_L) = 0 \text{ in } \Omega_L \\ \int_{\partial \Omega_L} (g + \sigma(v_L, P_L)) \nu = 0, \int_{\partial \Omega_L'} (g + \sigma(v_L, P_L)) \nu = 0 \end{cases}$$

(V) Main results : bounds on the variance

$$\begin{aligned} \int_{\mathcal{O}_L \setminus \mathcal{I}_L} \vartheta : (\mathcal{D}\phi_L - \mathcal{D}\phi'_L) &= -(\alpha_L - \alpha'_L) e \cdot \int_{\mathcal{I}_L} v_L \\ &+ (\alpha'_L + 1) \left(\sum_{x \in \mathcal{B}_L(x)} e \cdot \int_{\mathcal{I}_L} v_L - \sum_{x'_L \in \mathcal{B}_L(x)} e \cdot \int_{\mathcal{I}_L} v_L \right) \\ &- \sum_{x'_L \in \mathcal{B}_L(x)} \int_{\partial \mathcal{I}_L} (v_L - \int_{\mathcal{I}_L} v_L) \cdot \sigma(\phi'_L, \pi'_L - c'_L) \nu \end{aligned}$$

Without hyperuniformity :

$$\begin{aligned} \left| \int_{\mathcal{O}_L \setminus \mathcal{I}_L} \vartheta : \mathcal{D}\phi_L - \int_{\mathcal{O}_L \setminus \mathcal{I}_L} \vartheta : \mathcal{D}\phi'_L \right| &\leq L^{-d} |e \cdot \int_{\mathcal{I}_L} v_L| + \boxed{\int_{\mathcal{B}_L(x)} |v_L|}^{CL} \\ &+ \left(\int_{\mathcal{B}_3(x)} |v_L|^2 + |\langle \mathcal{D} \rangle^{1/2} g|^2 \right)^{1/2} \left(1 + \int_{\mathcal{B}_3(x)} |\mathcal{D}\phi'_L|^2 \right)^{1/2} \end{aligned}$$

With hyperuniformity : \mathcal{P} and \mathcal{P}' have same number of points

$\alpha_L = \alpha'_L$ and $\text{Card}\{n \mid x_n \in \mathcal{B}_L(x)\} = \text{Card}\{n' \mid x_{n'} \in \mathcal{B}_L(x)\}$
 \rightarrow gives a gradient

$$\begin{aligned} \left| \int_{\mathcal{O}_L \setminus \mathcal{I}_L} \vartheta : \mathcal{D}\phi_L - \int_{\mathcal{O}_L \setminus \mathcal{I}_L} \vartheta : \mathcal{D}\phi'_L \right| \\ \leq \boxed{\int_{\mathcal{B}_2(x)} |v_L|}^{\text{screening}} + \left(\int_{\mathcal{B}_3(x)} |v_L|^2 + |\langle \mathcal{D} \rangle^{1/2} g|^2 \right)^{1/2} \left(1 + \int_{\mathcal{B}_3(x)} |\mathcal{D}\phi'_L|^2 \right)^{1/2} \end{aligned}$$

(IV) Main results: bounds on the variance

Argument 2: related to two terms:

$$+ \| S_{\Omega_L} (|\nabla \psi_L|^2 + |\langle \nabla \rangle^{1/2} g|^2) |\nabla \phi_L|^2 \|_{L^p(\Omega)}$$

$$+ \| S_{\Omega_L} |\psi_L|^2 \|_{L^p(\Omega)}$$

stochastic improvement on reflection method: any $q!$

Annealed C2 theory for eqns for ψ_L

$$\begin{cases} -\Delta \psi_L + \nabla P_L = \operatorname{div} g & \& \operatorname{div} \psi_L = 0 \text{ in } \Omega_L \cap \Gamma_L \\ \mathcal{D}(\psi_L) = 0 \text{ in } \bar{\Gamma}_L \\ \int_{\partial \Omega_L} (g + \sigma(\psi_L, P_L)) \nu = 0, \int_{\partial \Omega_L} \psi_L \cdot (g + \sigma(\psi_L, P_L)) \nu = 0 \end{cases}$$

screening

- perturbative: $\| \psi_L \|_{L^q(\Omega_L, L^p(\Omega))} \leq \| g \|_{L^q(\Omega_L, L^p(\Omega))}$
 $|q-2|, |p-2| \leq \eta$
 (use also duality in probability)

CL

- non perturbative: $\| \psi_L \|_{L^q(\Omega_L, L^p(\Omega))} \leq \| g \|_{L^q(\Omega_L, L^{p+\eta}(\Omega))}$
 $\forall p, q, \forall \eta > 0$

$$\begin{aligned} \| S_{\Omega_L} |\psi_L|^2 \|_{L^p(\Omega)} &\stackrel{d \geq 2}{\leq} \| (S_{\Omega_L} |\psi_L|^{\frac{2d}{d-2}})^{\frac{d-2}{d}} \|_{L^p(\Omega)} = \| \psi_L \|_{L^{\frac{2d}{d-2}}(\Omega_L, L^{2p}(\Omega))}^2 \\ &\leq \| g \|_{L^{\frac{2d}{d-2}}(\Omega_L, L^{2p+\eta}(\Omega))}^2 = \| g \|_{L^{\frac{2d}{d-2}}(\Omega_L)}^2 \end{aligned}$$

A remark on annealed C τ estimates

$$\left\{ \begin{array}{l} -\Delta v_L + \nabla P_L = \operatorname{div} g \quad \& \quad \operatorname{div} v_L = 0 \quad \text{in } \Omega_L \setminus I_L \\ \mathcal{D}(v_L) = 0 \quad \text{in } I_L \\ \int_{\partial I_{in}} (g + \sigma(v_L, P_L)) \nu = 0, \quad \int_{\partial I_{in}} \Theta \nu \cdot (g + \sigma(v_L, P_L)) \nu = 0 \end{array} \right. \quad (*)$$

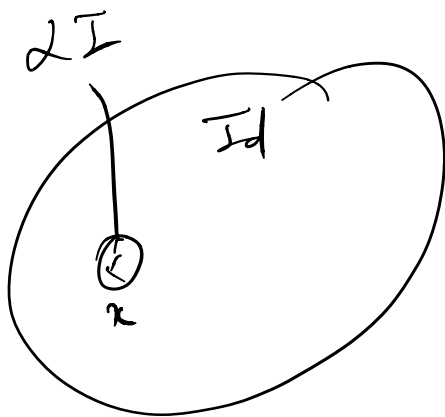
this is the **elliptical case**, that is: no divergence issue like in sedimentation **[cf. talk tomorrow]**

Indeed, energy estimate directly gives

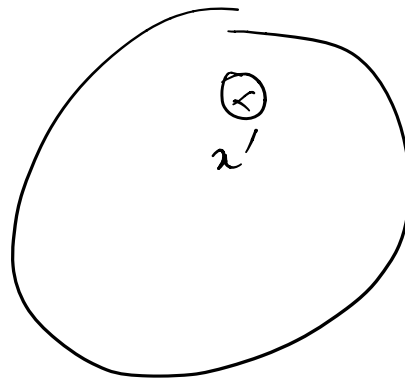
$$\int_{\Omega_L} |\nabla v_L|^2 \leq \int_{\Omega_L} |g|^2$$

Approach to annealed C τ estimates as in stochastic homogenization of divergence-free elliptic equations:

- 1) large-scale $C^{1,\alpha}$ estimates [difficulty: pressure]
- 2) Lipschitz \Rightarrow large scale quenched (weighted) C τ
- 3) Quenched C $\tau \Rightarrow$ annealed C τ



$$A' \geq A \pm SA$$



$$SA = (1-\alpha) Id \Delta_{B(r)} + (\alpha-1) Id \Delta_{B(r')}$$