

Problèmes de variance en homogénéisation stochastique

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- Classical periodic and stochastic homogenization
 - Periodic setting
 - Stochastic setting : Underlying hypotheses, ergodic results.
- Variance reduction for general stochastic homogenization
 - Where does the randomness come from ?
 - The technique of antithetic variables
 - Theoretic validation and numerical experiment
- Weakly stochastic cases
 - Hypotheses and existence of an expansion
 - Variance control for the approximate expansion

Classical periodic and stochastic homogenization

- Consider the family $(u_\varepsilon)_{\varepsilon>0}$ of solutions to the following PDEs :

$$\begin{cases} -\operatorname{div} \left(A \left(\frac{x}{\varepsilon} \right) \nabla u_\varepsilon \right) = f & \text{in } \mathcal{D}, \\ u_\varepsilon = 0 & \text{on } \partial\mathcal{D}, \end{cases}$$

with $A(y) = [a_{ij}(y)] \in (L^\infty(\mathbb{R}^d))^{d \times d}$ is a **Q-periodic** function ($Q = [0, 1]^d$) such that :

$$\exists \gamma > 0, \forall \xi \in \mathbb{R}^d, \forall y \in \mathbb{R}^d \quad \xi^T A(y) \xi \geq \gamma |\xi|^2,$$

and $f \in L^2(\mathcal{D})$.

- Up to extraction of a subsequence, the limit u_* in the weak sense of $(u_\varepsilon)_{\varepsilon>0}$ as $\varepsilon \rightarrow 0$ can be shown to satisfy the following equation :

$$\begin{cases} -\operatorname{div} (A^* \nabla u_*) = f & \text{in } \mathcal{D}, \\ u_* = 0 & \text{on } \partial\mathcal{D}. \end{cases}$$

- A^* is the **homogenized matrix**, which reflects the properties of some limit material. Its definition involves solutions w_p of the so-called **corrector problems** :

$$\begin{cases} \operatorname{div} [A(y) (p + \nabla w_p)] = 0, \\ w_p \text{ is periodic,} \end{cases}$$

- Definition of the homogenized matrix :
 - general matrix :

$$[A^*]_{ij} = \int_Q (e_i + \nabla w_{e_i})^T A (e_j + \nabla w_{e_i}) dy,$$

- symmetric matrix :

$$[A^*]_{ij} = \int_Q (e_i + \nabla w_{e_i})^T A e_j dy.$$

- Consider the family $(u_\varepsilon)_{\varepsilon>0}$ of solutions to the following SPDEs :

$$\begin{cases} -\operatorname{div} \left(A \left(\frac{x}{\varepsilon}, \omega \right) \nabla u_\varepsilon (\cdot, \omega) \right) = f & \text{in } \mathcal{D}, \\ u_\varepsilon (\cdot, \omega) = 0 & \text{on } \partial\mathcal{D}, \end{cases}$$

with $A(y, \omega)$ **stationnary** such that :

$$\exists \gamma > 0, \forall x_i \in \xi \in \mathbb{R}^d, \forall \xi \in \mathbb{R}^d \quad \xi^T A(y) \xi \geq \gamma |\xi|^2, \text{ almost surely}$$

and $f \in L^2(\mathcal{D})$.

- The **deterministic** limit in the weak sense u_\star of $(u_\varepsilon)_{\varepsilon>0}$ as $\varepsilon \rightarrow 0$ can be shown to satisfy a deterministic elliptic equation.

- The computation of A^* involves solutions w_p of the **stochastic corrector problems** :

$$\begin{cases} \operatorname{div} [A(y, \omega) (p + \nabla w_p)] = 0, \\ \nabla w_p \text{ is stationary,} \\ \mathbb{E} \left(\int_Q \nabla w_p(y, \cdot) dy \right) = 0. \end{cases}$$

- Definition of the **deterministic** homogenized matrix :

$$[A^*]_{ij} = \mathbb{E} \left(\int_Q (e_i + \nabla w_{e_i})^T A (e_j + \nabla w_{e_i}) dy \right),$$

- Numerical computation of A^* is thus quite a difficult task.
- Discretization of a SPDE posed on the whole space \mathbb{R}^d .

Variance reduction for general stochastic homogenization

- Our Goal: approximating A^*
- **First step**: the truncated corrector problem

$$\begin{cases} -\operatorname{div} (A(\cdot, \omega) (p + \nabla w_p^N(\cdot, \omega))) = 0 & \text{on } \mathbb{R}^d, \\ w_p^N(\cdot, \omega) \text{ is } Q_N\text{-periodic.} \end{cases}$$

where $Q_N = [-N, N]$.

- **Second step** : defining the approximated homogenized matrix A_N^*

$$[A_N^*]_{ij}(\omega) = \frac{1}{|Q_N|} \int_{Q_N} (e_i + \nabla w_{e_i}^N(y, \omega))^T A(y, \omega) (e_j + \nabla w_{e_j}^N(y, \omega)) dy$$

- **The randomness originates from truncation**

Variance issues: Monte Carlo methods (I)

- Underlying assumption: $\mathbb{E}(A_N^*) \approx A^*$
- **Monte Carlo** method for computations of $\mathbb{E}(A_N^*)$
 - Definition of estimators associated to $M = 2\mathcal{M}$ realizations

$$\mu_M \left([A_N^*]_{ij} \right) = \frac{1}{M} \sum_{\mathbf{m}=1}^M [A_N^{*,\mathbf{m}}]_{ij},$$

$$\sigma_M \left([A_N^*]_{ij} \right) = \frac{1}{M-1} \sum_{\mathbf{m}=1}^M \left([A_N^{*,\mathbf{m}}]_{ij} - \mu_M \left([A_N^*]_{ij} \right) \right)^2.$$

- Properties of $\mu_M \left([A_N^*]_{ij} \right)$:
 - Strong law of large numbers
 - Central Limit Theorem

$$\sqrt{M} \left(\mu_M \left([A_N^*]_{ij} \right) - \mathbb{E} \left([A_N^*]_{ij} \right) \right) \xrightarrow[M \rightarrow +\infty]{\mathcal{L}} \sqrt{\text{Var} \left([A_N^*]_{ij} \right)} \mathcal{N}(0, 1)$$

Variance issues: Monte Carlo methods (II)

- The context of numerical practice
- $\mathbb{E}(A_N^*)$ is in the **confidence interval**

$$\mu_M \left([A_N^*]_{ij} \right) \pm 1.96 \frac{\sqrt{\sigma_M \left([A_N^*]_{ij} \right)}}{\sqrt{M}}.$$

- Precision of the estimation depends on $\text{Var} \left([A_N^*]_{ij} \right)$

Our approach:

Defining new estimators of A_N^* with smaller variance than $\mu_M \left([A_N^*]_{ij} \right)$

Variance issues: antithetic variables (I)

- We give ourselves \mathcal{M} i.i.d. copies $(A^{\mathbf{m}}(x, \omega))_{1 \leq \mathbf{m} \leq \mathcal{M}}$ of $A(x, \omega)$
- Defining
 - an **antithetic initial field**

$$B^{\mathbf{m}}(x, \omega) = T(A^{\mathbf{m}}(x, \omega)), \quad 1 \leq \mathbf{m} \leq \mathcal{M},$$

- an antithetic corrector problem whose solution is denoted v_p
- We finally build an **antithetic homogenized matrix**

$$[B_N^{\star, \mathbf{m}}]_{ij}(\omega) = \frac{1}{|Q_N|} \int_{Q_N} (e_i + \nabla v_{e_i}^{N, \mathbf{m}}(\cdot, \omega))^T B^{\mathbf{m}}(\cdot, \omega) (e_j + \nabla v_{e_j}^{N, \mathbf{m}}(\cdot, \omega)).$$

Variance issues: antithetic variables (II)

- We introduce a new random variable

$$\tilde{A}_N^{*,\mathbf{m}}(\omega) := \frac{1}{2} (A_N^{*,\mathbf{m}}(\omega) + B_N^{*,\mathbf{m}}(\omega)).$$

- It is

- unbiased

$$\mathbb{E} \left(\tilde{A}_N^{*,\mathbf{m}} \right) = \mathbb{E} \left(A_N^{*,\mathbf{m}} \right)$$

- convergent

$$\tilde{A}_N^{*,\mathbf{m}} \xrightarrow[N \rightarrow +\infty]{} A^* \text{ almost surely,}$$

because B is ergodic.

\tilde{A}_N^* requires the solution of 2 correctors problems instead of 1 for the classical estimator !

- We define **new estimators** with identical properties

$$\begin{aligned}\mu_{\mathcal{M}} \left(\left[\tilde{A}_N^* \right]_{ij} \right) &= \frac{1}{\mathcal{M}} \sum_{\mathbf{m}=1}^{\mathcal{M}} \left[\tilde{A}_N^{*,\mathbf{m}} \right]_{ij}, \\ \sigma_{\mathcal{M}} \left(\left[\tilde{A}_N^* \right]_{ij} \right) &= \frac{1}{\mathcal{M}-1} \sum_{\mathbf{m}=1}^{\mathcal{M}} \left(\left[\tilde{A}_N^{*,\mathbf{m}} \right]_{ij} - \mu_{\mathcal{M}} \left(\left[\tilde{A}_N^* \right]_{ij} \right) \right)^2.\end{aligned}$$

- Efficiency criteria
 - Equal computational costs

$M = 2\mathcal{M}$ realizations of A_N^* vs \mathcal{M} of \tilde{A}_N^*

- Theoretic condition

$$\text{Cov} \left(\left[A_N^* \right]_{ij}, \left[B_N^* \right]_{ij} \right) \leq 0.$$

Results

- Hypotheses on the initial field A

- A deterministic function of **uniform** r.v on finite-size cells

There exists an integer n and a function \mathcal{A} , defined on $Q_N \times \mathbb{R}^n$, such that

$$\forall x \in Q_N, \quad A(x, \omega) = \mathcal{A}(x, X_1(\omega), \dots, X_n(\omega)) \quad \text{a.s.},$$

where $\{X_k(\omega)\}_{1 \leq k \leq n}$ are independent scalar random variables, which are all distributed according to the uniform law $\mathcal{U}[0, 1]$

- **Monotonicity** in the sense of symmetric matrices

For all $x \in Q_N$, and any vector $\xi \in \mathbb{R}^d$, the map

$$(x_1, \dots, x_n) \in \mathbb{R}^n \mapsto \xi^T \mathcal{A}(x, x_1, \dots, x_n) \xi$$

is non-decreasing with respect to each of its arguments.

Proposition 0.1 *We assume that $A(x, \omega)$ satisfies the above hypotheses. We define on Q_N the field*

$$B(x, \omega) := \mathcal{A}(x, 1 - X_1(\omega), \dots, 1 - X_n(\omega)),$$

antithetic to $A(\cdot, \omega)$. We associate to this field the antithetic corrector problem (replacing A by B), the solution of which is denoted by v_p^N , and the matrix $B_N^(\omega)$, defined from v_p^N . Then,*

$$\forall \xi \in \mathbb{R}^d, \quad \text{Cov}(\xi^T A_N^* \xi, \xi^T B_N^* \xi) \leq 0$$

Otherwise stated, \tilde{A}_N^ is an unbiased estimator of $\mathbb{E}(A_N^*)$, and*

$$\text{Var}(\xi^T \tilde{A}_N^* \xi) \leq \frac{1}{2} \text{Var}(\xi^T A_N^* \xi).$$

- This result guarantees the efficiency of the technique for
 - a **wide range of initial fields** characterized by
 1. different local behaviours
 2. different correlation structures
 3. different isotropy properties
 - a **wide range of outputs** including
 1. diagonal terms $[A_N^*]_{ii}$
 2. eigenvalues of A_N^*
 3. eigenvalues of $L_{A_N^*} = -\text{div}(A_N^* \nabla \cdot)$

- The case of an **isotropic** initial field with **no correlation**

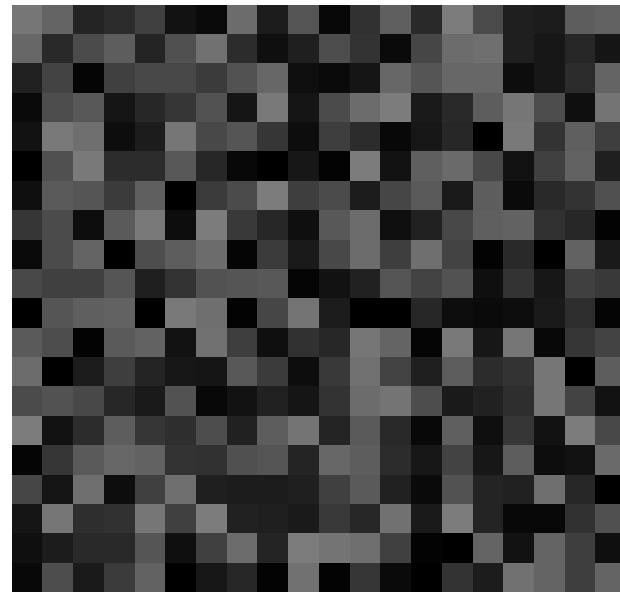
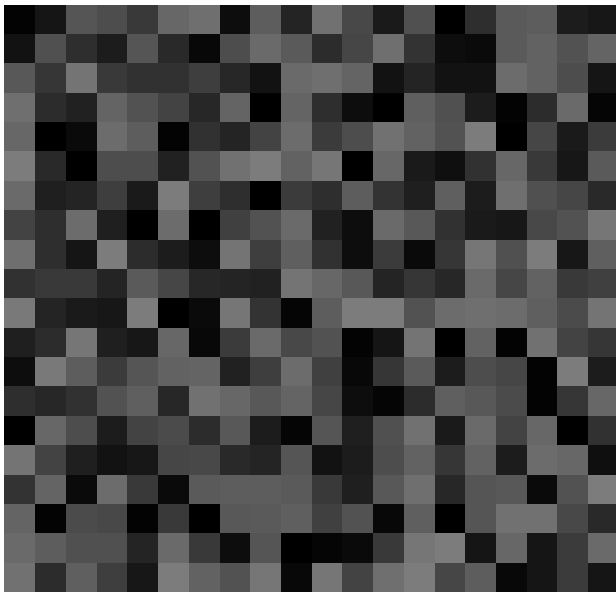
$$A(x, \omega) = \sum_{k \in \mathbb{Z}^d} \mathbf{1}_{Q+k}(x) a_k(\omega) \text{Id} = \sum_{k \in \mathbb{Z}^d} \mathbf{1}_{Q+k}(x) f(X_k(\omega)) \text{Id}.$$

- Values a_k and $a_{k'}$ of A on two distinct cells $Q + k$ and $Q + k'$ are **independent**.
- A is invariant under rotations of angle $\pi/2$
- The law of A is determined by that of a_0
 - case (i): $a_0 \sim \mathcal{U}([\alpha, \beta])$; $f(x) = \alpha + (\beta - \alpha)x$
 - case (ii): $a_0 \sim \mathcal{B}(1/2)$, $\mathbb{P}(a_0 = \alpha) = 1/2$ and $\mathbb{P}(a_0 = \beta) = 1/2$;

$$f(x) = \alpha + (\beta - \alpha) \mathbf{1}_{\{\frac{1}{2} \leq x \leq 1\}}.$$

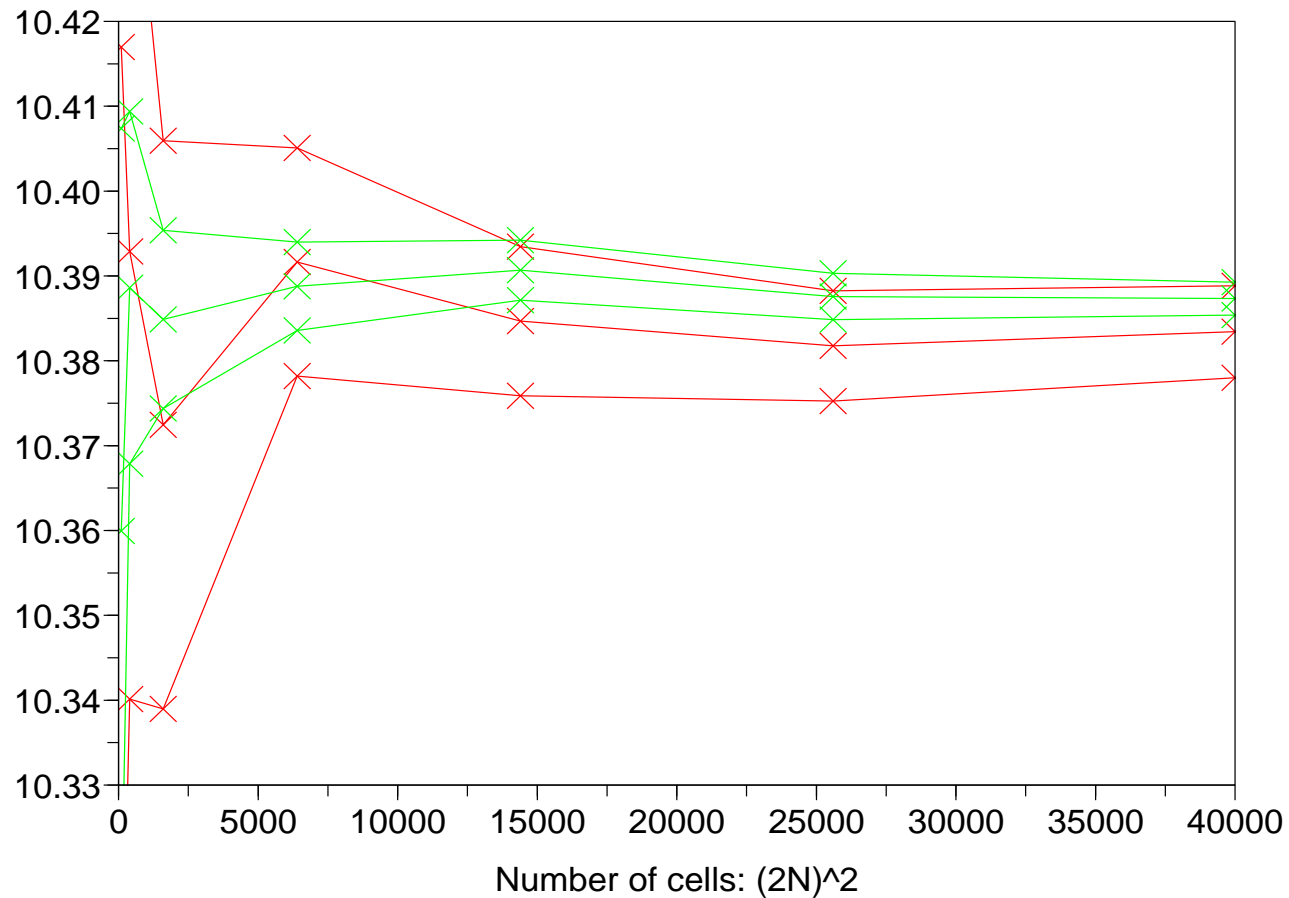
- In the sequel, $\alpha = 3, \beta = 20$.

- One realization of A and its counterpart B when $a_0 \sim \mathcal{U}([\alpha, \beta])$

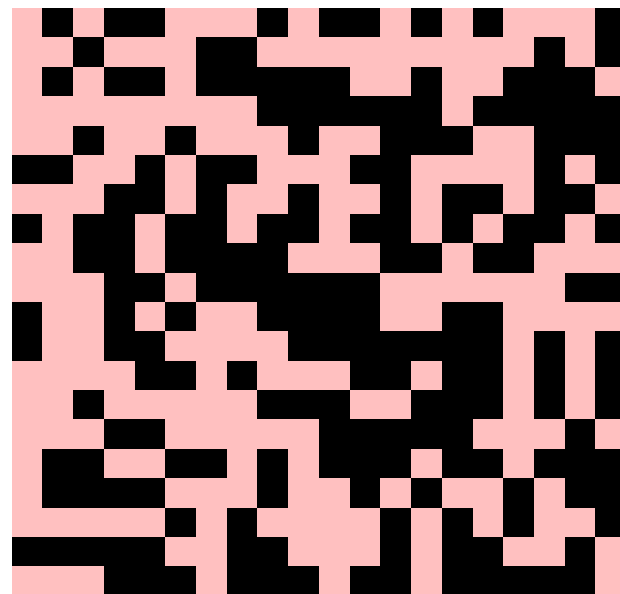


Numerical experiment: case (i) (II)

- Estimated means $\mu_{100}(A_N^*)$ and $\mu_{50}(\tilde{A}_N^*)$ and respective confidence intervals.

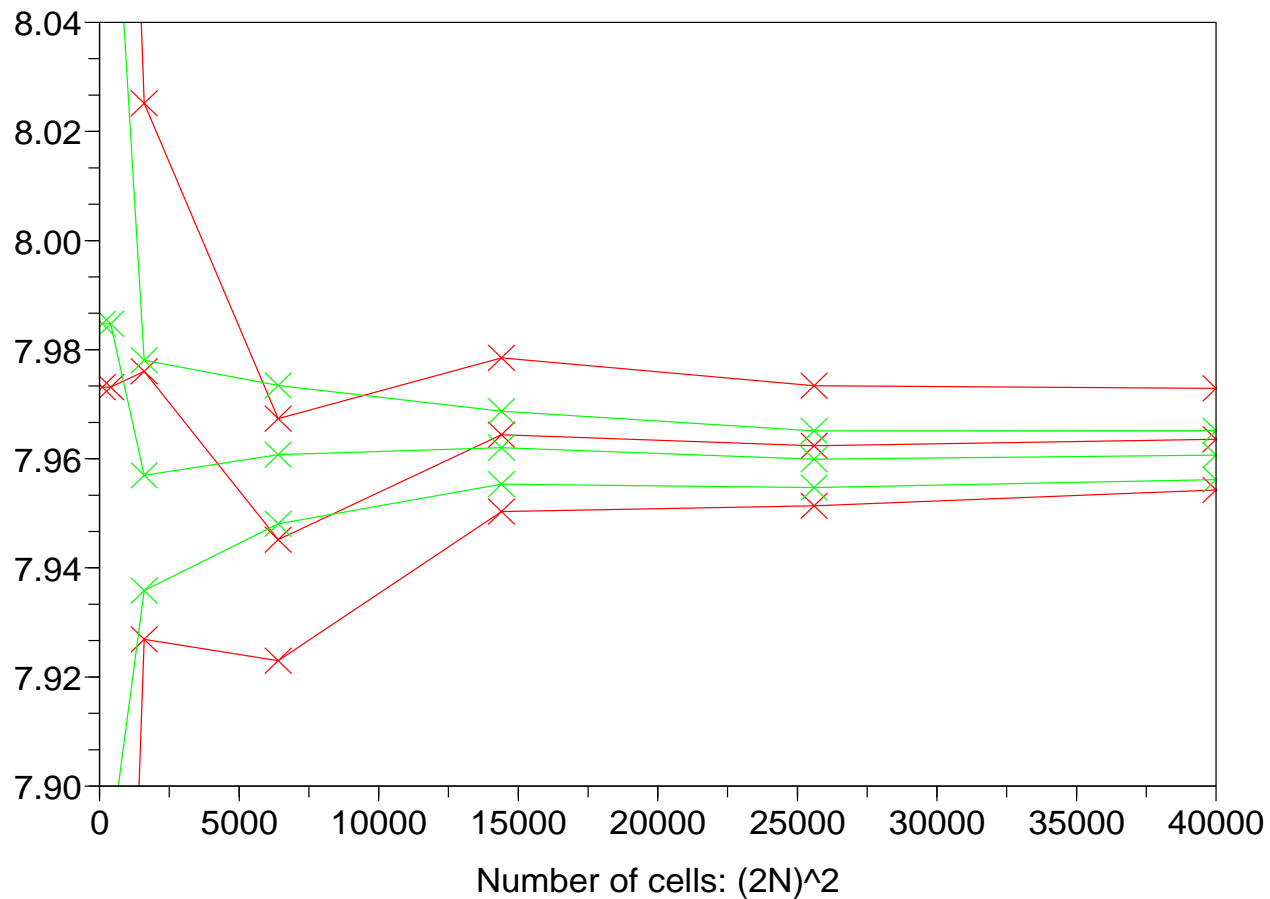


- One realization of A and its counterpart B when $a_0 \sim \mathcal{B}(\frac{1}{2})$



Numerical experiment: case (ii) (II)

- Estimated means $\mu_{100}(A_N^*)$ and $\mu_{50}(\tilde{A}_N^*)$ and respective confidence intervals.



- Ratio of the widths of confidence intervals

$$R([A_N^*]_{11}) = \frac{\sigma_{100}([A_N^*]_{11})}{2\sigma_{50}([\tilde{A}_N^*]_{11})}.$$

- Numerical values in the isotropic case

N	$a_0 \sim \mathcal{B}(1/2)$	$a_0 \sim \mathcal{B}(1/3)$	$a_0 \sim \mathcal{U}([\alpha, \beta])$
5	5.34	2.06	6.31
10	3.91	1.56	6.46
20	5.41	2.92	10.2
40	3.07	2.31	6.67
60	4.41	2.47	6.16
80	4.49	1.95	5.68
100	4.28	2.99	7.89

Weakly stochastic cases

- Linear perturbations

- First model:

$$A_\eta(x, \omega) = A_{\text{per}}(x) + \eta A_1(x, \omega) + O(\eta^2), \text{ in } (L^\infty(\Omega \times \mathbb{R}^d))^{d \times d}$$

where $A(y)$ is a **periodic** function satisfying the same conditions of boundedness and coercivity as above.

- Second model:

$$A_\eta(x, \omega) = A_{\text{per}}(x) + B_\eta(x, \omega) + O(\eta^2), \text{ in } (L^\infty(\mathbb{R}^d; L^p(\Omega)))^{d \times d}$$

- Perturbations of alternative models

$$A_\eta(x, \omega) = A_{\text{per}}(\Phi_\eta^{-1}(x, \omega))$$

The Blanc Le Bris Lions framework

- Consider the family $(u_\varepsilon)_{\varepsilon>0}$ of solutions to the following SPDEs :

$$\begin{cases} -\operatorname{div} \left(A \left(\Phi^{-1} \left(\frac{x}{\varepsilon}, \omega \right) \right) \nabla u_\varepsilon (\cdot, \omega) \right) = f & \text{in } \mathcal{D}, \\ u_\varepsilon (\cdot, \omega) = 0 & \text{on } \partial\mathcal{D}, \end{cases}$$

where $A(y)$ is a **periodic** function satisfying the same conditions of boundedness and coercivity as above.

- $\Phi(x, \omega)$ is a **stationnary stochastic diffeomorphism** :

$$\begin{cases} \operatorname{EssInf}_{\omega \in \Omega, x \in \mathbb{R}^d} [\det(\nabla \Phi(x, \omega))] = \nu > 0, \\ \operatorname{EssSup}_{\omega \in \Omega, x \in \mathbb{R}^d} (|\nabla \Phi(x, \omega)|) = M < \infty, \\ \nabla \Phi(x, \omega) \text{ is stationnary.} \end{cases}$$

- Main results proved within the seminal paper :
 - The **deterministic** limit in the weak sense u_* of $(u_\varepsilon)_{\varepsilon>0}$ as $\varepsilon \rightarrow 0$ satisfies a deterministic elliptic equation.
 - A new corrector problem :

$$\begin{cases} \operatorname{div} [A (\Phi^{-1}(y, \omega)) (p + \nabla w_p)] = 0, \\ w_p(y, \omega) = \tilde{w}_p (\Phi^{-1}(y, \omega), \omega), \quad \nabla \tilde{w}_p \text{ is stationary,} \\ \mathbb{E} \left(\int_{\Phi(Q, \cdot)} \nabla w_p(y, \cdot) dy \right) = 0, \end{cases}$$

- A new definition of the homogenized matrix :

$$[A^*]_{ij} = \det \left(\mathbb{E} \left(\int_Q \nabla \Phi(z, \cdot) dz \right) \right)^{-1} \mathbb{E} \left(\int_{\Phi(Q, \cdot)} (e_i + \nabla w_{e_i}(y, \cdot))^T A (\Phi^{-1}(y, \cdot)) e_j dy \right).$$

The perturbative result (I)

- Consider now the family $(\Phi_\eta)_{\eta \geq 0}$ of **stochastic stationary diffeomorphism**, satisfying :

$$\Phi(x, \omega) = x + \eta \Psi(x, \omega) + O(\eta^2),$$

when $\eta \rightarrow 0$ in $C^1(\mathbb{R}^d, L^\infty(\Omega))$.

- The gradient of the corrector $w_p = \tilde{w}_p(\Phi^{-1}(x, \omega))$ possesses an analogous expansion :

$$\nabla \tilde{w}_p(x, \omega) = \nabla w_p^0(x) + \eta \nabla w_p^1(x, \omega) + O(\eta^2)$$

in $L^2(Q \times \Omega)$.

- Introducing such an expansion into the formula of A^* leads to :

$$A^* = A^0 + \eta A^1 + O(\eta^2),$$

- The equations satisfied by w_p^0 and w_p^1 are known.**

- **Deterministic cell problem** satisfied by w^0 :

$$\begin{cases} -\operatorname{div} [A (p + \nabla w_p^0)] = 0, \\ w_p^0 \text{ is } Q\text{-periodic.} \end{cases}$$

- **Stochastic cell problem** satisfied by w^1 :

$$\begin{cases} -\operatorname{div} [A \nabla w_p^1] = \operatorname{div} [-A \nabla \Psi \nabla w_p^0 - (\nabla \Psi^T - (\operatorname{div} \Psi) \operatorname{Id}) A (p + \nabla w_p^0)], \\ \nabla w_p^1 \text{ is stationary and } \mathbb{E} \left(\int_Q \nabla w_p^1 \right) = 0. \end{cases}$$

- First terms of A^* 's expansion :

$$\begin{aligned} A_{ij}^0 &= \int_Q (e_i + \nabla w_{e_i}^0)^T A e_j \\ A_{ij}^1 &= -\int_Q \mathbb{E}(\operatorname{div} \Psi) A_{ij}^0 + \int_Q (e_i + \nabla w_{e_i}^0)^T A e_j \mathbb{E}(\operatorname{div} \Psi) \\ &\quad + \int_Q (\mathbb{E}(\nabla w_{e_i}^1) - \mathbb{E}(\nabla \Psi) \nabla w_{e_i}^0)^T A e_j. \end{aligned}$$

The perturbative result (III)

- In order to compute A^1 the knowledge of the deterministic $\overline{w}_{e_i}^1 = \mathbb{E}(\nabla w_{e_i}^1)$ is only required.
- $\overline{w}_{e_i}^1 = \mathbb{E}(\nabla w_{e_i}^1)$ is the unique periodic solution to the deterministic cell problem :
$$-\operatorname{div} [A \nabla \overline{w}_p^1] = \operatorname{div} [-A \mathbb{E}(\nabla \Psi) \nabla w_p^0 - (\mathbb{E}(\nabla \Psi^T) - \mathbb{E}(\operatorname{div} \Psi) \operatorname{Id}) A (p + \nabla w_p^0)]$$
- Numerical strategy : instead of tackling the very general stochastic problem, one deals with two independent deterministic problems.

Do these results still hold when dealing with approximation spaces ?

A discretized version of **BLL** results

Discretization of the corrector problem (I)

- After domain truncation and introduction of finite element spaces, one obtains the following discrete formulation

$$\begin{cases} \text{Find } \tilde{w}_p^{h,N}(\cdot, \omega) \in V_h^{\text{per}}(Q_N) \text{ such that, for all } \tilde{v}_h \in V_h^{\text{per}}(Q_N), \\ \int_{Q_N} \det(\nabla\Phi) (\nabla\tilde{v}_h)^T (\nabla\Phi)^{-T} A \left(p + (\nabla\Phi)^{-1} \nabla\tilde{w}_p^{h,N}(\cdot, \omega) \right) = 0, \text{ a.s.} \end{cases}$$

- The chosen method is thus a Monte-Carlo Method Finite Elements Method.
 - At each realization ω corresponds a divergence-form deterministic elliptic equation on the bounded domain Q_N , whose solution existence is guaranteed by Lax-Milgram Lemma.
 - No discretization of the random space Ω (Galerkin Method)

Discretization of the corrector problem (II)

- Thanks to the approximated corrector $\tilde{w}_p^{h,N}(\cdot, \omega)$, one can compute an **approximated random tensor**

$$(A_{\star}^{h,N})_{ij}(\omega) = \frac{\frac{1}{|Q_N|} \int_{Q_N} \det(\nabla\Phi) (e_i + (\nabla\Phi)^{-1} \nabla \tilde{w}_{e_i}^{h,N})^T A e_j}{\det\left(\frac{1}{|Q_N|} \int_{Q_N} \nabla\Phi\right)}$$

- $A_{\star}^{h,N}(\omega)$ is random due to the **loss of ergodicity** resulting from the discretization procedure.
- **What do the former results become within this discretized framework ?**
- One introduces the formal expansion : $\tilde{w}_p^{h,N} = w_p^{0,h,N} + \eta w_p^{1,h,N} + O(\eta^2)$

Discretization of the corrector problem (III)

- Introducing such a development within the discrete corrector problem
 - $w_p^{0,h,N} = w_p^{0,h}$ is **deterministic**. It does not depend on N .

$$\begin{cases} \text{Find } w_p^{0,h} \in V_h^{\text{per}}(Q) \text{ such that, for all } v_h \in V_h^{\text{per}}(Q), \\ \int_Q (\nabla v_h)^T A (p + \nabla w_p^{0,h}) = 0, \end{cases}$$

- $w_p^{1,h,N}$ is **stochastic**. It does depend on N :

$$\begin{cases} \text{Find } w_p^{1,h,N}(\cdot, \omega) \in V_h^{\text{per}}(Q_N) \text{ such that, for all } v_h \in V_h^{\text{per}}(Q_N), \\ \int_{Q_N} (\nabla v_h)^T A \nabla w_p^{1,h,N} = \int_{Q_N} (\nabla v_h)^T A \nabla \Psi \nabla w_p^{0,h} \\ + \int_{Q_N} (\nabla \Psi^T - (\text{div } \Psi) \text{Id}) A (p + \nabla w_p^{0,h}). \end{cases}$$

- First terms of the expansion

$$\begin{aligned} [A^{0,h}]_{ij} &= \int_Q (e_i + \nabla w_{e_i}^{0,h})^T A \\ [A^{1,h,N}]_{ij} &= -(A^{0,h})_{ij} \frac{1}{|Q_N|} \int_{Q_N} \operatorname{div} \Psi \\ &+ \frac{1}{|Q_N|} \int_{Q_N} (e_i + \nabla w_{e_i}^{0,h})^T A e_j \operatorname{div} \Psi \\ &+ \frac{1}{|Q_N|} \int_{Q_N} (\nabla w_{e_i}^{1,h,N} - \nabla \Psi \nabla w_{e_i}^{0,h})^T A e_j. \end{aligned}$$

- **Does the expansion $(A_{\star}^{h,N})_{ij}(\omega) = (A^{0,h})_{ij} + \eta(A^{1,h,N})_{ij}$ hold ?**

- The discrete expansion

Proposition 0.2 *Suppose $\Phi(x, \omega) = x + \eta\Psi(x, \omega) + O(\eta^2)$ when $\eta \rightarrow 0$, in $C^1(\mathbb{R}^d, L^\infty(\Omega))$, with $\nabla\Psi$ stationary. A constant C exists such that, for small values of parameter η ,*

$$\eta^{-2} \left| \nabla \tilde{w}_p^{h,N}(x, \omega) - \nabla w_p^{0,h}(x) - \eta \nabla w_p^{1,h,N}(x, \omega) \right| \leq C,$$

where $\tilde{w}_p^{h,N}$, $w_p^{0,h}$ et $w_p^{1,h,N}$ are respectively solutions of the discrete, order 0 and order 1 corrector problems. In addition we have

$$\eta^{-2} \left| A_{\star}^{h,N}(\omega) - A^{0,h} - \eta A^{1,h,N}(\omega) \right| \leq C,$$

- Sketch of the proof

- uniform bounds in η thanks to the domain boundedness.
- passing to the limit into the variational equation which corresponds to the discrete corrector problem.

Numerical experiments

- Reference periodic structure

$$\forall x \in Q, A(x) = a_{\text{per}}(x) \text{Id}_2, \quad a_{\text{per}}(x_1, x_2) = \beta + (\alpha - \beta) \sin^2(\pi x_1) \sin^2(\pi x_2).$$

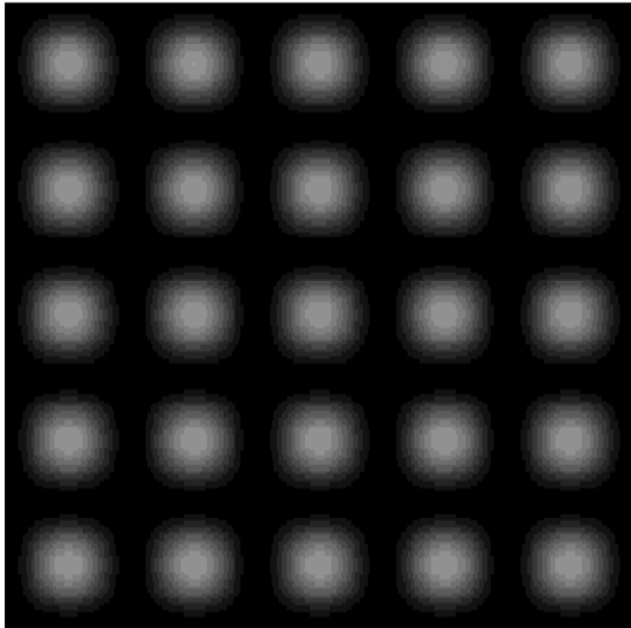
- Stochastic diffeomorphism :

- Two independant family of i.i.d random variables $(X_k)_{k \in \mathbb{Z}}$ et $(Y_k)_{k \in \mathbb{Z}}$, who follow a uniform law $\mathcal{U}([a, b])$
- $\Phi_\eta(x) = x + \eta \Psi(x, \omega)$, with $\Psi(x, \omega) = (\psi_X(x_1, \omega), \psi_Y(x_2, \omega))$ and

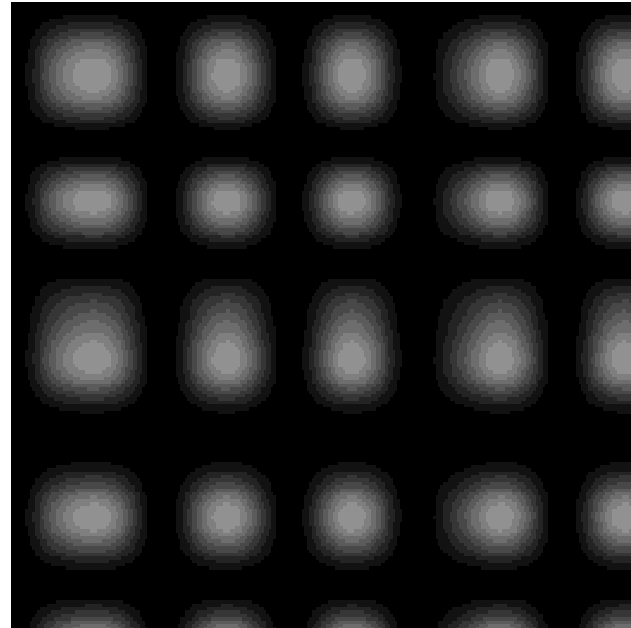
$$\psi_X(x_1, \omega) = \sum_{k \in \mathbb{Z}} 1_{[k, k+1[}(x_1) \left(\sum_{q=0}^{k-1} X_q(\omega) + 2X_q(\omega) \int_k^{x_1} \sin^2(2\pi t) dt \right),$$

- Parameters

$$\alpha = 10 \quad \beta = 1 \quad a = 5.75 \quad b = -2.25$$



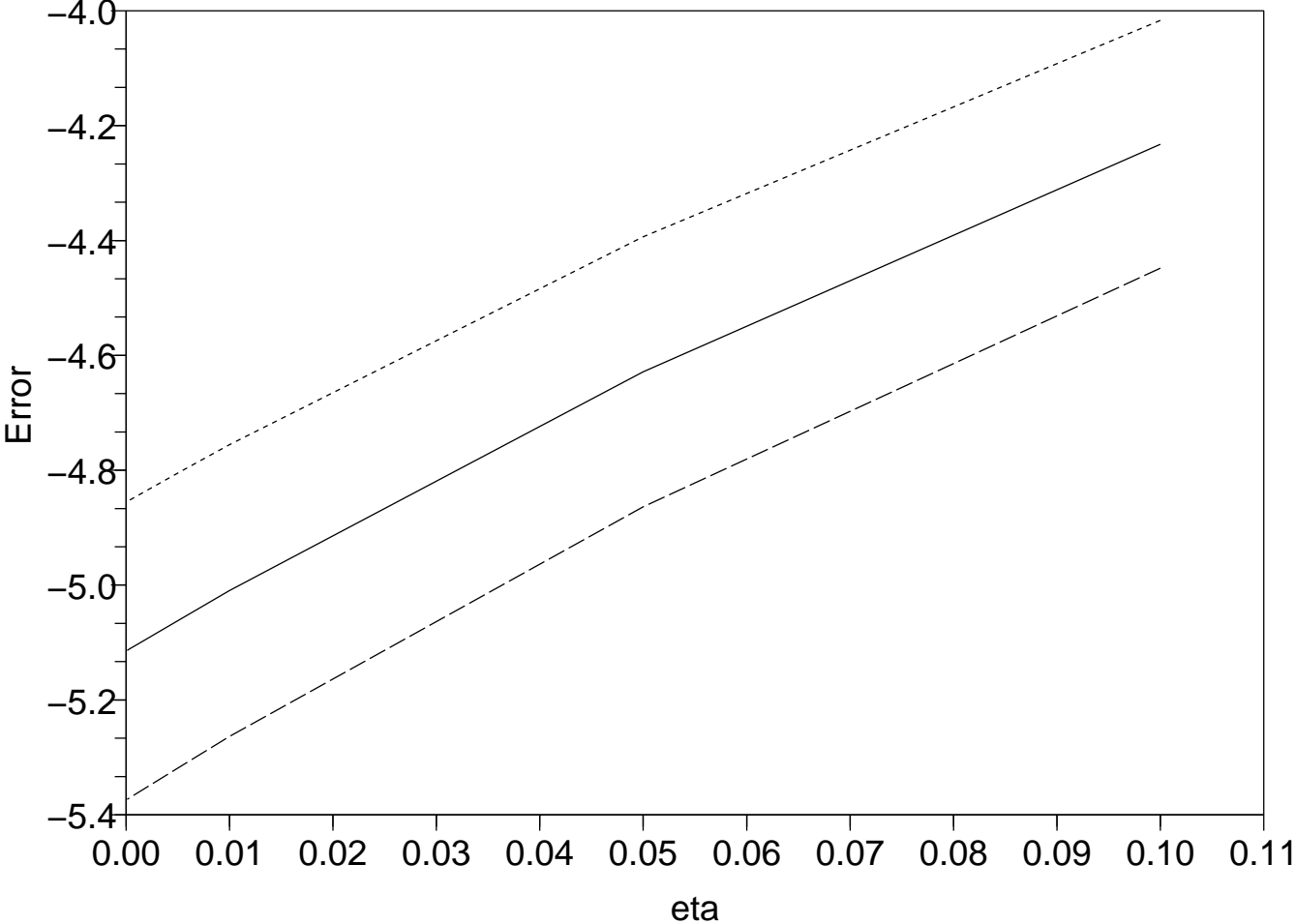
(a) Periodic structure



(b) Periodic structure through diffeomorphism

$$\Phi_{\eta}(x, \omega), \eta = 0.05$$

Expansion Error and A_2 Estimate



- Variance reduction
 - R. Costeaouec, C. Le Bris et F. Legoll, *Variance reduction in stochastic homogenization: proof of concept, using antithetic variables*, Boletin Soc. Esp. Mat. Apl., vol. 50, 9-27 (2010).
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