

# Modeling of random media and stochastic homogenization for elliptic equations

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Sept. 2020

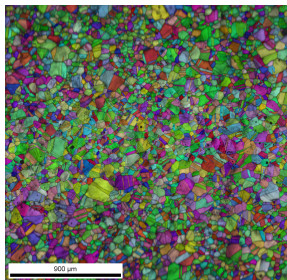
# Outline

- 1 Modeling of random media
  - Examples of disordered media
  - Generation of samples of disordered media
- 2 Periodic homogenization (elliptic equations)
  - General principle
  - Multi-scale homogenization
  - Numerical approximation and bounds
- 3 Stochastic homogenization (elliptic equations)
  - Energy method
  - Periodic-Stochastic comparison
  - Numerical simulation of homogenized coefficients
  - Quasi-periodic media

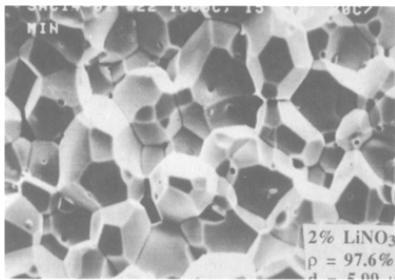
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## Examples of discrete disordered media



(a) Polycrystalline Material<sup>1</sup>



(b) Cell material<sup>2</sup>

- The information is very rich : what parameter (connectivity, "grain" size, correlation of properties between two grains, ...) do you want your model to enforce ?
- When (many) different samples are available, stationarity and ergodicity hypotheses can be checked
- Experimentally, one can measure the statistics of the different possible quantities of interest, often assuming ergodicity

$$\bar{\kappa} = \int_D \kappa(\mathbf{x}) d\mathbf{x}, \quad C_\kappa(\mathbf{y}) = \int_D \kappa(\mathbf{x}) \kappa(\mathbf{y} - \mathbf{x}) d\mathbf{x}$$

1. P. MU. "Study of crack initiation in low-cycle fatigue of an austenitic stainless steel". Thèse de doct. France : École Centrale de Lille, 2011

2. S. KUMAR et al. "Properties of a three-dimensional Poisson-Voronoi tessellation : a Monte Carlo study". In : *J. Stat. Phys.* 67.3-4 (1992), p. 523-551.

DOI : 10.1007/BF01049719



# Examples of continuous complex recordings

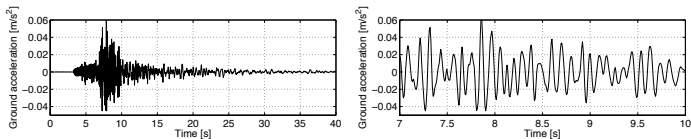


FIGURE – Example of a seismic record (right : zoom on the coda).

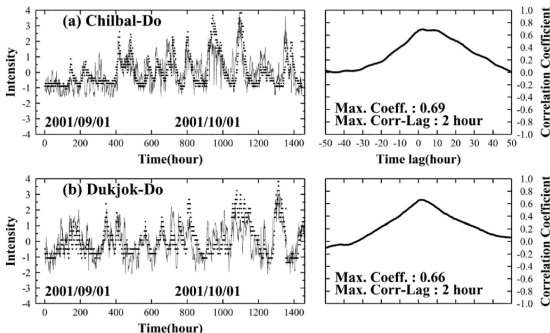
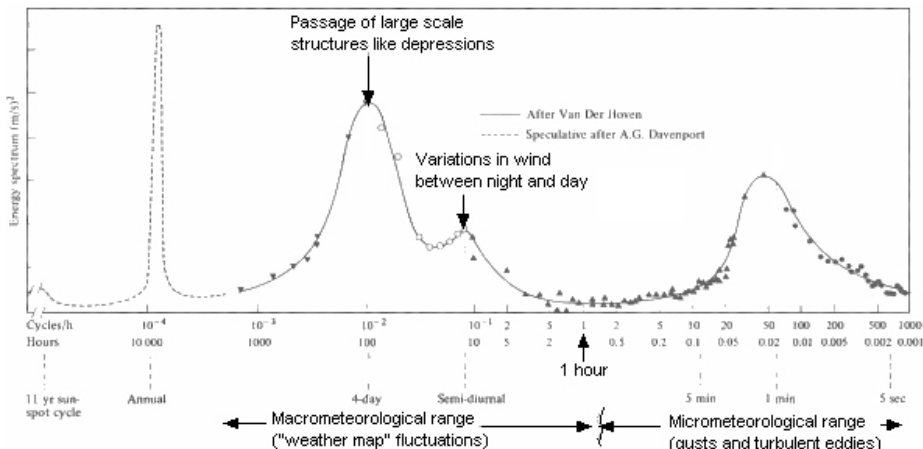


FIGURE – Correlation between wind speed and wave height in the Yellow Sea

## Examples of power spectral density



Energy spectrum of wind speed at 100 metres above the ground.

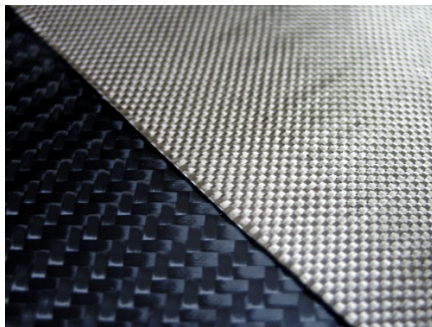
(Van der Hoven at Brooklyn, NY, USA)

From *Wind Forces on Buildings and Structures* by E.L.Houghton and N.B.Carruthers  
Edward Arnold 1976

## Examples of quasi-periodic structures



(a) Silicon carbide fibres in a titanium matrix (© ONERA)



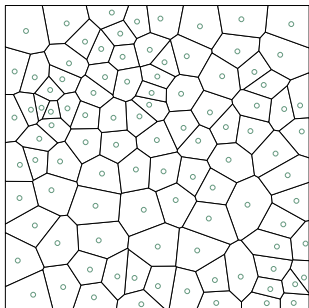
(b) Carbon fibres woven material

- Man-made material can potentially be "almost" periodic
- How much does the influence of small deviations from periodicity influence the final design ?

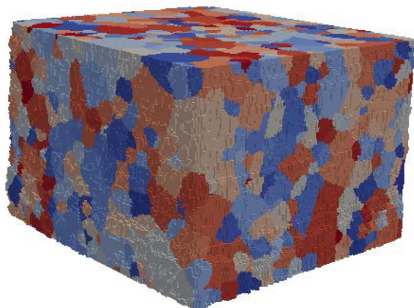
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# Generation of samples of discrete random media



(a) 2D model of a polycrystalline material<sup>3</sup>



(b) 3D model of a polycrystalline material<sup>4</sup>

1 Draw  $n$  points (for instance uniformly in space)  $(\mathbf{x}_k)_{1 \leq k \leq n} \in \mathbb{R}^d$

2 Generate the Voronoi tessellation, that is, for each  $1 \leq k \leq n$ ,

$$V(\mathbf{x}_k) = \left\{ \mathbf{x} \in \mathbb{R}^d, |\mathbf{x} - \mathbf{x}_k| < |\mathbf{x} - \mathbf{x}_j|, j \neq k \right\}$$

3 Draw (following appropriate probability law) realizations of the mechanical parameter associated to each  $V(\mathbf{x}_k)$ .

3. C. TALISCHI et al. "PolyMesher : a general-purpose mesh generator for polygonal elements written in Matlab". In : *Struct. Multidisc. Optim.* 45 (2012), p. 309-328. DOI : 10.1007/s00158-011-0706-z

4. K. TEFERRA et L. GRAHAM-BRADY. "Tessellation growth models for polycrystalline microstructures". In : *Comp. Mat. Sci.* 102 (2015), p. 57-67. DOI : 10.1016/j.commatsci.2015.02.006

# Generation of samples of discrete random media

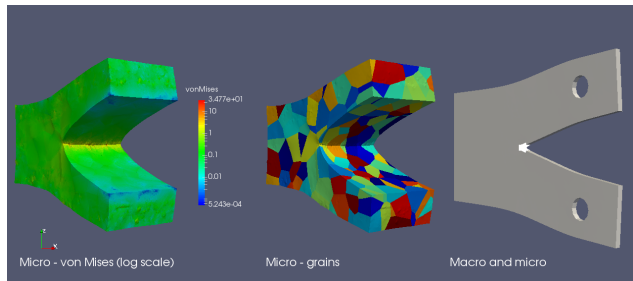


FIGURE – Simulation of the influence of grains at the tip of a fracture<sup>5</sup>

- Once generated (for instance using point processes), it is possible to obtain (at least numerically) statistics of the continuous fields of mechanical parameters

$$(\kappa^*)^{-1} = \frac{1}{N} \sum_{j=1}^N \int_D \kappa_j^{-1}(\mathbf{x}) d\mathbf{x}, \quad C_\kappa(\mathbf{y}) = \frac{1}{N} \sum_{j=1}^N \int_D \kappa(\mathbf{x}) \kappa_j(\mathbf{y} - \mathbf{x}) d\mathbf{x}$$

- Samples can be generated (contrarily to experiments), so stationarity or ergodicity can be checked.

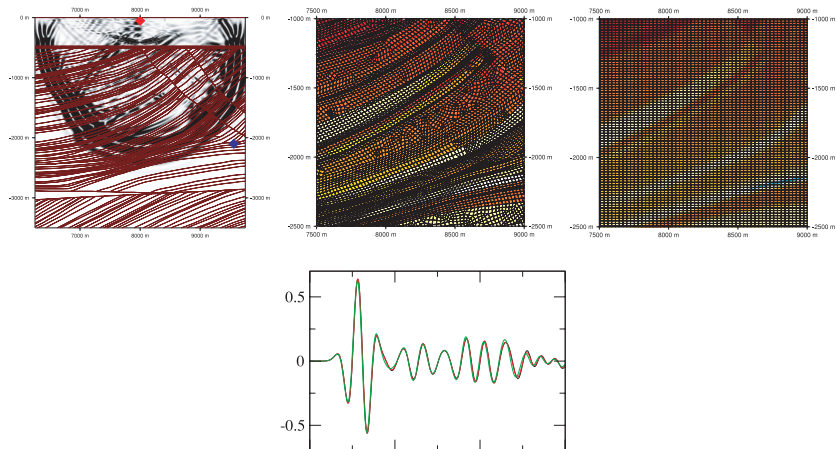
5. T. MILANETTO-SCHLITTLER et R. COTTAREAU. "Fully scalable implementation of a volume coupling scheme for the modeling of polycrystalline materials". In : *Comp. Mech.* 60.5 (2017), p. 827-844. DOI : 10.1007/s00466-017-1445-9

## Raining technique



- Check the statistics !!!

## Discrete or continuous random fields ?



- Discrete or continuous heterogeneity is often a modeling assumption more than a "fact"
- \*wink\* : homogenization is useful !!



## Generation of samples of a continuous random field

Realizations of a continuous Gaussian random field with expectation  $a_0$  and power spectral density  $S(k)$  can be generated using the following spectral representation formula<sup>6</sup> :

$$A(x) = a_0 + 2 \sum_{j=1}^N \sqrt{S(k_j)} \delta k \zeta_j \cos(k_j x + \phi_j)$$

where  $k_j = j\delta k$ , and the  $\{\zeta_j\}_{1 \leq j \leq N}$  et  $\{\phi_j\}_{1 \leq j \leq N}$  are families of independent random variables, respectively with average 0 and variance 1 (any probability density), and uniform on  $[-\pi, \pi]$ .

- The only control on the field is on its expectation and covariance (in particular, all the marginal laws are Gaussian).
- Can be easily (but not necessarily cheaply) generalized to multidimensional fields
- Very efficient FFT scheme.
- Other techniques exist (see Olivier's talk ?)

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6. M. SHINOZUKA et G. DEODATIS. "Simulation of stochastic processes by spectral representation". In : *Appl. Mech. Rev.* 44.4 (1991), p. 191-205. DOI : 10.1115/1.3119501

# Large scale generation of random fields : overlap and merge<sup>7</sup>

## Issue

- $N$  increases with the total size of the generation domain even for fixed correlation length
- We are performing a sum of functions correlated over the entire domain to retrieve a function correlated only over a correlation length

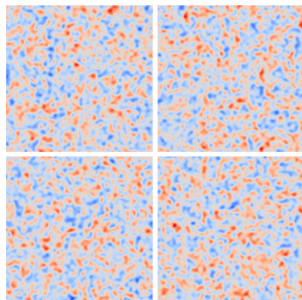
## Idea

- Generate independent random fields  $u_j(x)$  over each domain for an overlapping partition  $P$
- Assemble the field through partition of unity functions  $\psi_j(x)$

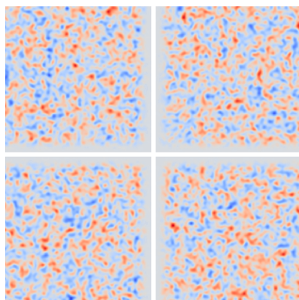
$$u(x) = \sum_{j \in P} \sqrt{\psi_j(x)} u_j(x)$$

7. L. DE CARVALHO PALUDO, V. BOUVIER et R. COTTHEREAU. "Scalable parallel scheme for sampling of Gaussian random fields over large domains". In : *Int. J. Numer. Meth. Engr.* 117.8 (2019), p. 845-859. DOI : 10.1002/nme.5981

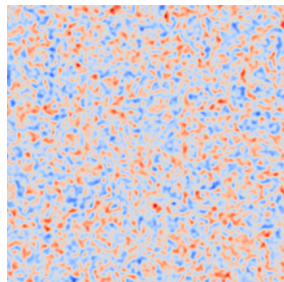
## Illustration in 2D



Generate independent fields  
( $u_i$ )



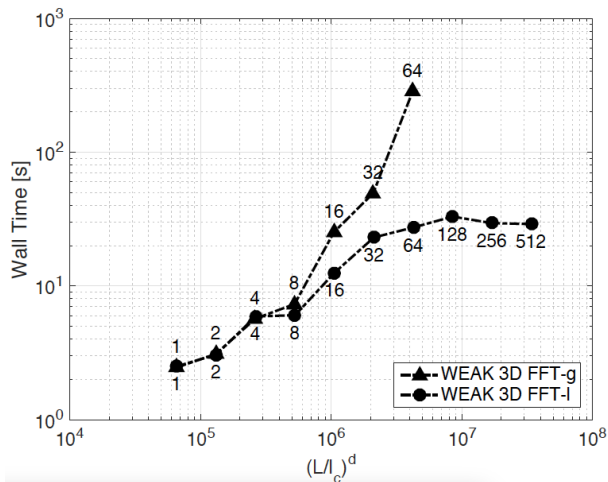
Multiply each field by the  
square root of a partition of  
unity  
( $u_i\sqrt{\psi_i}$ )



Sum all the fields  
( $\sum u_i\sqrt{\psi_i}$ )

# Weak scalability of local scheme (versus global scheme)

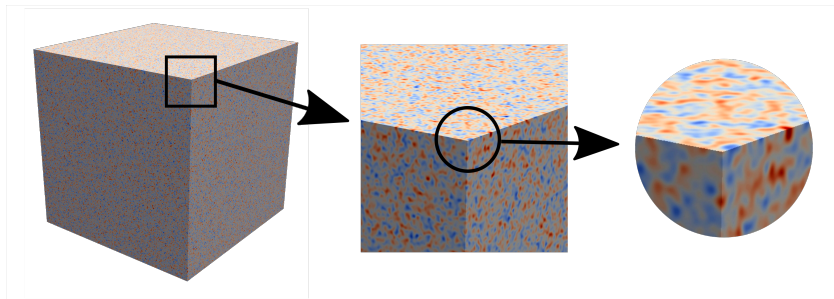
on Igloo (SGI machine with 800 Intel Xeon X5650 cores at Châtenay-Malabry, France)



on each processor :  $8 \times 10^6$  DOFs,  $30 \times 30 \times 60 \ell_c^3$   
overlap =  $5\ell_c$  (in overlapping technique)

# Random field generation over large cluster

Generation over  $300\ell_c \times 300\ell_c \times 300\ell_c$  in 116 s (walltime) over 512 processors



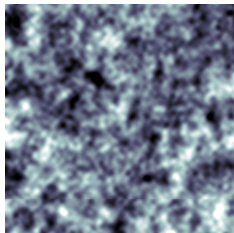
Development of an open-source library for large-scale generation of random fields

<https://github.com/cottureau/randomField>

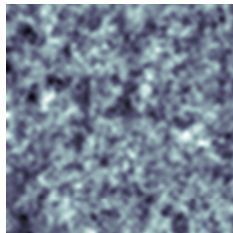
(based on MPI, pHDF5, FFTW)

# Comparison of power spectral density models

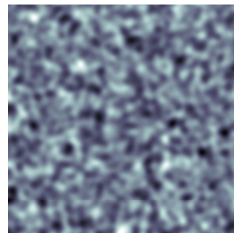
Centered Gaussian fields with  $\tau_c = 1$  et  $\sigma^2 = 1$  (in 2D)



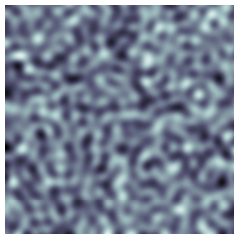
(a) Exponential



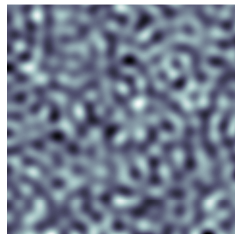
(b) Power law



(c) Gaussian



(d) Triangular



(e) Truncated white noise

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# Objectives of homogenization

Example of a diffusion problem (elliptic equation)

## Micro-scale problem

$$\begin{aligned} -\nabla \cdot A\left(\frac{x}{\epsilon}\right) \nabla u_\epsilon\left(x, \frac{x}{\epsilon}\right) &= f(x), & \text{in } D \\ u_\epsilon &= 0, & \text{on } \partial D \end{aligned}$$

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## Macro-scale problem

$$\begin{aligned} \mathcal{L}(u^*(x), x; f(x)) &= 0, & \text{in } D \\ u^* &= 0, & \text{on } \partial D \end{aligned}$$

How to find the homogenized equation (and does it exist?) and its coefficients?

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# General principle of multi-scale homogenization

## Description of the micro-scale problem

- Sequence of problems indexed on  $\epsilon$  : Find  $u_\epsilon \in H_0^1(D)$  such that

$$-\nabla \cdot A \left( \frac{x}{\epsilon} \right) \nabla u_\epsilon \left( x, \frac{x}{\epsilon} \right) = f(x), \quad \text{in } D$$

$$u_\epsilon = 0, \quad \text{on } \partial D$$

- If  $f \in L^2(D)$ , then the problem is well posed : there exists a unique solution  $u_\epsilon \in H_0^1(D)$  and

$$\|u_\epsilon\|_{H_0^1(D)} \leq C \|f\|_{L^2(D)}$$

- Study of the convergence of  $u_\epsilon \rightarrow u^*$

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# Multi-scale homogenization

## General principle

- Expanding the solution using a separation of scales assumption  $\epsilon \ll 1$

$$u_\epsilon(x) = \sum_{k=0}^{+\infty} \epsilon^k u_k \left( x, y = \frac{x}{\epsilon} \right)$$

- Introduction of the *ansatz* in the diffusion equation (micro)
- Modification of the derivation operator

$$\nabla = \epsilon^{-1} \nabla_y + \nabla_x$$

using the relation, valid for functions  $v_\epsilon(x) = v(x, x/\epsilon)$ ,

$$\nabla v_\epsilon(x) = \left[ \frac{1}{\epsilon} \nabla_y v(x, y) + \nabla_x v(x, y) \right]_{y=\frac{x}{\epsilon}}$$

- A sequence of problems is obtained for each power of  $\epsilon$

This part follows G. Allaire's course notes<sup>8</sup>. The classical book<sup>9</sup> is extremely clear, and treats both periodic and stochastic cases.

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8. G. ALLAIRE. "Introduction to homogenization theory". In : *Multiscale methods – 32nd Computational Fluid Dynamics*. Sous la dir. de H. DECONINCK. Von Karman Institute Lecture Series. Von Karman Institute, 2002

9. A. BENSOUSSAN, J.-L. LIONS et G. PAPANICOLAOU. *Asymptotic analysis for periodic structures*. Studies in Mathematics and its Applications 5. North Holland, 1978

# Two-scale homogenization

General expansion

$$\begin{aligned} & -\epsilon^{-2} [\nabla_y \cdot \mathbf{A} \nabla_y u_0] \left( x, \frac{x}{\epsilon} \right) \\ & -\epsilon^{-1} [\nabla_y \cdot \mathbf{A} (\nabla_x u_0 + \nabla_y u_1) + \nabla_x \cdot \mathbf{A} \nabla_y u_0] \left( x, \frac{x}{\epsilon} \right) \\ & -\epsilon^0 [\nabla_y \cdot \mathbf{A} (\nabla_x u_1 + \nabla_y u_2) + \nabla_x \cdot \mathbf{A} (\nabla_x u_0 + \nabla_y u_1)] \left( x, \frac{x}{\epsilon} \right) \\ & - \sum_{k=1}^{+\infty} \epsilon^k [\nabla_x \cdot \mathbf{A} (\nabla_x u_k + \nabla_y u_{k+1}) + \nabla_y \cdot \mathbf{A} (\nabla_x u_{k+1} + \nabla_y u_{k+2})] \left( x, \frac{x}{\epsilon} \right) \\ & = f(x) \end{aligned}$$

# Two-scale homogenization

- Cancellation of  $\epsilon^{-2}$  order term

$$-\nabla_y \cdot A(y) \nabla_y u_0(x, y) = 0$$

- The leading-order term is the solution of the homogenized problem

$$u_0(x, y) = u(x)$$



## Two-scale homogenization

- Cancellation of  $\epsilon^{-1}$  order term

$$-\nabla_y \cdot \mathbf{A}(y) \nabla_y u_1(x, y) = \nabla_y \cdot \mathbf{A}(y) \nabla_x u(x)$$

- Proportionality of  $u_1(x, y)$  and  $\nabla_x u(x)$  :

$$u_1(x, y) = \mathbf{w}(y) \cdot \nabla_x u(x) = \sum_{k=1}^d (w_k(y) \mathbf{e}_k) \cdot \nabla_x u(x)$$

- Defines the corrector problems with periodic boundary conditions :

$$-\nabla_y \cdot \mathbf{A}(y) \nabla_y w_k(y) = \nabla_y \cdot \mathbf{A}(y) \mathbf{e}_k$$

## Two-scale homogenization

- Cancellation of  $\epsilon^0$  order term

$$-\nabla_y \cdot A(y) \nabla_y u_2(x, y) = \nabla_y \cdot A(y) \nabla_x u_1 + \nabla_x \cdot A(y) (\nabla_y u_1 + \nabla_x u) + f(x)$$

### Theorem 1 (Fredholm alternative)

Let  $f(y) \in L^2_{\#}(Y)$  be a periodic function. There exists a solution in  $H^1_{\#}(Y)$  of

$$-\nabla_y \cdot (A(y) \nabla_y w(y)) = f(y)$$

if and only if  $\int_Y f(y) dy = 0$ . The solution is unique up to an additive function.

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- Using  $\int_Y \nabla_y \cdot \alpha(x, y) dy = 0$  for  $\alpha$   $Y$ -periodic, this yields :

$$\begin{aligned} \left( \frac{1}{|Y|} \int_Y -\nabla_x \cdot A(y) (\nabla_y u_1 + \nabla_x u) dy \right) &= f(x) \\ -\nabla_x \cdot \left( \frac{1}{|Y|} \int_Y A(y) (I + \nabla_y \mathbf{w}(y)) dy \right) \nabla_x u &= f(x) \end{aligned}$$

- **This equation is the homogenized equation** and

$$A^* = \frac{1}{|Y|} \int_Y A(y) (I + \nabla_y \mathbf{w}(y)) dy = \frac{1}{|Y|} \int_Y (I + \nabla_y \mathbf{w}(y))^T A(y) (I + \nabla_y \mathbf{w}(y)) dy$$

# Two-scale homogenization

## Summary

### Corrector problem

For each basis vector  $\mathbf{e}_i$  (canonical basis of  $\mathbb{R}^d$ ), find  $w_i \in H_{\#}^1(Y)$  such that

$$-\nabla_y \cdot \mathbf{A}(y)(\mathbf{e}_i + \nabla_y w_i(y)) = 0, \quad \text{in } Y.$$

### Macro-scale problem

Find  $u^* \in H_0^1(D)$  such that

$$-\nabla \cdot \mathbf{A}^*(x) \nabla u^*(x) = f(x), \text{ in } D, \text{ et } u^* = 0, \text{ on } \partial D$$

with

$$\mathbf{A}^* = \frac{1}{|Y|} \int_Y (\mathbf{I} + \nabla_y \mathbf{w}(y))^T \mathbf{A}(y) (\mathbf{I} + \nabla_y \mathbf{w}(y)) dy.$$

### Homogenized solution

$$u_\epsilon \approx u(x) + \epsilon \mathbf{w}(y) \cdot \nabla_x u(x)$$

The approach described before is very efficient at providing the "correct" homogenized solution but other methods are often needed to prove that it is indeed correct (energy method, ...).

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## Corrector problem

For each basis vector  $\mathbf{e}_i$ , find  $w_i \in H_{\#}^1(Y)$  such that

$$-\nabla_y \cdot \mathbf{A}(y)(\mathbf{e}_i + \nabla_y w_i(y)) = 0, \quad \text{in } Y$$

- provides micro-scale information on the reaction of the medium to macro-scale excitations
- must (in general) be approximated numerically
- the problem is set on a cell (much cheaper numerically than the micro-scale problem) ...  
NB : this will not be true anymore in stochastics ...

# Functional characterization of the homogenized tensor

- Homogenized tensor

$$A^* = \frac{1}{|Y|} \int_Y (I + \nabla_y \mathbf{w}(y)) \cdot A(y) (I + \nabla_y \mathbf{w}(y)) dy$$

- Redefining the homogenized problem as a minimization problem

$$A_{jj}^* = \frac{1}{|Y|} \min_{w(y) \in H_{\#}^1(Y)} \int_Y (\mathbf{e}_j + \nabla_y w_j) \cdot A(\mathbf{e}_j + \nabla_y w_j) dy$$

# Voigt and Reuss bounds

## Voigt (upper) bound

For  $w(y) = \text{cst}$

$$A_{jj}^* \leq \mathbf{e}_j \cdot \left( \frac{1}{|\mathcal{Y}|} \int_{\mathcal{Y}} A \, dy \right) \mathbf{e}_j$$

## Reuss (lower) bound

By increasing the space over which minimization is performed

$$A_{jj}^* \geq \min_{\zeta(y) \in (L^2_{\#}(\mathcal{Y}))^d, \int_{\mathcal{Y}} \zeta \, dy = 0} \int_{\mathcal{Y}} (\mathbf{e}_j + \zeta_j) \cdot A (\mathbf{e}_j + \zeta_j) \, dy = \mathbf{e}_j \cdot \left( \frac{1}{|\mathcal{Y}|} \int_{\mathcal{Y}} A^{-1} \, dy \right)^{-1} \mathbf{e}_j$$



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## Some preliminary notations

- $\mathcal{H} = L^2(\Theta, \mathcal{T}, \mathbb{P})$  is the Hilbert space of square integrable functions on  $\Theta$  with

$$\mathbb{E}[gh] = \int_{\Theta} g(\theta)h(\theta)\mathbb{P}(d\theta)$$

- $H = L^2(D, \mathcal{H})$  is the Hilbert space of square integrable functions on  $D$  with values in  $\mathcal{H}$  with

$$(u, v) = \int_D \mathbb{E}[u(x)v(x)]dx = \int_D \int_{\Theta} u(x, \theta)v(x, \theta)\mathbb{P}(d\theta)dx$$

- $H^1 = H^1(D, \mathcal{H})$  is the Hilbert space of  $\mathcal{H}$ -valued functions over  $D$  whose derivatives are square integrable with

$$(u, v)_1 = (u, v) + \sum_{k=1}^d \left( \frac{\partial u}{\partial x_k}, \frac{\partial v}{\partial x_k} \right)$$

- $H_0^1 = H_0^1(D, \mathcal{H})$  is the Hilbert space of  $\mathcal{H}$ -valued functions with square integrable derivatives and such that  $u = 0$  on  $\partial D$

This part follows Papanicolaou and Varadhan's paper<sup>10</sup>

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10. G. C. PAPANICOLAOU et S. R. S. VARADHAN. "Boundary value problems with rapidly oscillating random coefficients". In : *Proceedings of the Conference on Random Fields*. Sous la dir. de J. FRITZ et J. L. LEBOWITZ. T. 2. Seria Colloquia Mathematica Societatis Janos Bolyai 27. North Holland, 1981, p. 835-873

# Hypotheses on the diffusion tensor

## Diffusion tensor

We consider a **strictly stationary** matrix-valued **continuous** random field of diffusion tensor  $A(y, \theta) = a_{ij}(y, \theta) \mathbf{e}_i \otimes \mathbf{e}_j$  **bounded and coercive**

- the probability measure  $\mathbb{P}$  is invariant with respect to the translation group  $\tau_x : \Theta \rightarrow \Theta$  defined by

$$(\tau_x \theta)(y) = \theta(y - x), \quad \forall x, y \in \mathbb{R}^d.$$

For stationary random fields, we introduce the notation

$$\tilde{A}_{ij}(\theta) = A_{ij}(0, \theta), \quad \text{such that} \quad A_{ij}(x, \theta) = \tilde{A}_{ij}(\tau_x \theta)$$

The translation group is also assumed ergodic : the only sets that are invariant are such that  $\mathbb{P}(A) = 0$  or  $1$ .

- the continuity is such that

$$\lim_{h \rightarrow 0} \mathbb{P}[|A(y+h, \theta) - A(y, \theta)| > \delta] = 0, \quad \forall \delta > 0, y \in \mathbb{R}^d$$

- $\exists a_0 > 0$  such that

$$a_0 |\xi|^2 \leq \xi \cdot A(y, \theta) \cdot \xi \leq a_0^{-1} |\xi|^2, \quad \forall y, \xi \in \mathbb{R}^d, \forall \theta \in \Theta$$

## General setting

- We consider a strictly stationary continuous random field  $A(y)$  bounded and coercive ...
- ... and a sequence of problems parameterized by  $\epsilon$  : Find  $u_\epsilon(x, \theta) \in H_0^1$  such that

$$\int_D \int_\Theta \left( A \left( \frac{x}{\epsilon}, \theta \right) \nabla u_\epsilon(x, \theta) \right) \cdot \nabla \phi(x, \theta) \mathbb{P}(d\theta) dx = \int_D \int_\Theta f(x) \phi(x, \theta) \mathbb{P}(d\theta) dx, \quad \forall \phi \in H_0^1 \quad (1)$$

where  $f(x) \in L^2(D, \mathbb{R})$  is a given deterministic function

- Each problem has a unique solution (by Lax-Milgram), which is bounded

$$(u_\epsilon, u_\epsilon)_1 \leq C \int_D |f(x)|^2 dx$$

# Weak convergence theorem

## Theorem 2 (Weak convergence)

The solution  $u_\epsilon \in H_0^1$  converges weakly in  $H_0^1$  to the solution  $u(x) \in H_0^1(D, \mathbb{R})$  of the deterministic variational problem

$$\int_D (A^* \nabla u(x)) \cdot \nabla \phi(x) dx = \int_D f(x) \phi(x) dx, \quad \forall \phi(x) \in H_0^1(D, \mathbb{R})$$

Here, the homogenized matrix  $A^*$  is a constant matrix defined by

$$A^* = \mathbb{E} \left[ \tilde{A} \left( I + \tilde{\psi} \right) \right],$$

where the corrector strains  $\tilde{\psi}$  are stationary and defined below (Theorem 3)

- Weak convergence means

$$\lim_{\epsilon \rightarrow 0} (u_\epsilon, \phi)_1 = (u, \phi)_1, \quad \forall \phi \in H_0^1$$

- The homogenized problem is no longer stochastic

## Theorem 3 (Correctors)

There exist functions  $\tilde{\psi}^k \in \mathcal{H}^d$ ,  $1 \leq k \leq d$ , such that

$$\mathbb{E} \left[ \tilde{A}(\mathbf{e}_k + \tilde{\psi}^k) \cdot \nabla \tilde{\phi} \right] = 0, \quad \forall \tilde{\phi} \in \mathcal{H}^1$$

Furthermore, there exist uniquely defined random fields  $\mathbf{w} = (w_k(x, \theta))_{1 \leq k \leq d}$  that are not stationary, but whose gradients are stationary, and such that  $\mathbf{w}(0, \theta) = 0$  and

$$\nabla w_k(x, \theta) = \psi^k(x, \theta) = \tilde{\psi}^k(\tau_{-x}\theta).$$

We denote  $\tilde{\psi} = \{\tilde{\psi}^k\}_{1 \leq k \leq d}$ .

Proof : see <sup>11</sup>

11. G. C. PAPANICOLAOU et S. R. S. VARADHAN. "Boundary value problems with rapidly oscillating random coefficients". In : *Proceedings of the Conference on Random Fields*. Sous la dir. de J. FRITZ et J. L. LEBOWITZ. T. 2. Seria Colloquia Mathematica Societatis Janos Bolyai 27. North Holland, 1981, p. 835-873

## Steps of the proof of Theorem 2 (energy method)

- Issue : the product of two weakly converging sequences  $A(x/\epsilon)$  and  $\nabla u_\epsilon$  does not necessarily converge to the product of the limit (but  $\xi^\epsilon = A(x/\epsilon)\nabla u^\epsilon(x)$  does)
- Solution : use a particular test function (oscillating function)  $\phi_\epsilon$  that allows to pass to the limit

$$\phi_\epsilon(x) = \phi(x) + \epsilon \sum_{j=1}^d w_j^*(x/\epsilon) \frac{\partial \phi(x)}{\partial x_j}$$

where  $w_k^*$  is solution to the adjoint corrector problem (with diffusion matrix  $A^T$  loaded in direction  $\mathbf{e}_k$ ). Note that

$$\nabla \phi_\epsilon(x) = \sum_{j=1}^d \frac{\partial \phi(x)}{\partial x_j} \left( \mathbf{e}_j + \nabla_y w_j^*(x/\epsilon) \right) + \epsilon \sum_{j=1}^d w_j^*(x/\epsilon) \nabla \frac{\partial \phi(x)}{\partial x_j}$$

- In the weak formulation

$$\begin{aligned} & (A(x/\epsilon)\nabla u_\epsilon, \nabla \phi_\epsilon) = (f, \nabla \phi_\epsilon) \\ & - \left( u_\epsilon, \nabla \cdot \left( A^T(x/\epsilon) \sum_{j=1}^d \frac{\partial \phi(x)}{\partial x_j} \left( \mathbf{e}_j + \nabla_y w_j^*(x/\epsilon) \right) \right) \right) + \epsilon C = (f, \nabla \phi_\epsilon) \\ & - \sum_{j=1}^d \left( u_\epsilon, \left( \nabla \frac{\partial \phi(x)}{\partial x_j} \right) \left( A^T(x/\epsilon) \left( \mathbf{e}_j + \nabla_y w_j^*(x/\epsilon) \right) \right) \right) + \epsilon C = (f, \nabla \phi_\epsilon) \end{aligned}$$



# Strong convergence theorem

## Theorem 4 (Strong convergence)

Let  $u_\epsilon \in H^1$  be the solution of (1) and  $u \in H^1(\mathbb{R}^d, \mathbb{R})$  be the solution of the deterministic variational problem

$$\int_D (A^* \nabla u(x)) \cdot \nabla \phi(x) dx = \int_D f(x) \phi(x) dx, \quad \forall \phi(x) \in H_0^1(\mathbb{R}^d, \mathbb{R}),$$

where the homogenized matrix  $A^*$  is defined as earlier. Then

$$\lim_{\epsilon \rightarrow 0} \int_{\mathbb{R}^d} \int_{\Theta} |u_\epsilon(x, \theta) - u(x)|^2 \mathbb{P}(d\theta) dx = 0$$

and

$$\lim_{\epsilon \rightarrow 0} \int_{\mathbb{R}^d} \int_{\Theta} \left\| \nabla u_\epsilon(x, \theta) - \nabla u(x) - \sum_{k=1}^d \psi^k \left( \frac{x}{\epsilon}, \theta \right) \frac{\partial u(x)}{\partial x_k} \right\|^2 \mathbb{P}(d\theta) dx = 0$$

where the corrector strains  $\psi^k$  were defined earlier (Theorem 3)

Proof : see <sup>12</sup>

12. G. C. PAPANICOLAOU et S. R. S. VARADHAN. "Boundary value problems with rapidly oscillating random coefficients". In : *Proceedings of the Conference on Random Fields*. Sous la dir. de J. FRITZ et J. L. LEBOWITZ. T. 2. Seria Colloquia Mathematica Societatis Janos Bolyai 27. North Holland, 1981, p. 835-873

# Outline

- 1 Modeling of random media
  - Examples of disordered media
  - Generation of samples of disordered media
- 2 Periodic homogenization (elliptic equations)
  - General principle
  - Multi-scale homogenization
  - Numerical approximation and bounds
- 3 Stochastic homogenization (elliptic equations)
  - Energy method
  - **Periodic-Stochastic comparison**
  - Numerical simulation of homogenized coefficients
  - Quasi-periodic media

## Periodic-Stochastic comparison

- Stationarity of the material random field replaces periodicity in the deterministic case.
- The physical "averaging" phenomena are the same and the ingredients of the theoretical proofs are (mostly) the same
- In the stochastic case, the corrector problem is posed on the entire (space) domain rather than on the periodic cell. The corrector problem is therefore *a priori* as expensive to approximate as the full-scale problem.
- When using the alternative formulation of the homogenized tensor, convergence is obtained for very large sizes  $N \rightarrow +\infty$ , even when multiple Monte Carlo realizations are considered. Approaches can be developed to improve that convergence

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# Numerical simulation of homogenized coefficients

## Alternative formulation of the corrector problem

There exist uniquely defined functions  $w_k(x, \theta)$ ,  $1 \leq k \leq d$ , whose gradients are stationary, and such that, if  $B_N$  is a cube of side  $2N$  in  $\mathbb{R}^d$ , centered at the origin,

$$\lim_{N \rightarrow +\infty} \frac{1}{(2N)^d} \int_{B_N} (A(x, \theta)(\mathbf{e}_k + \nabla w_k(x, \theta))) \cdot \nabla \phi(x, \theta) dx = 0,$$

$$\forall \phi \in H^1_{\mathcal{G}}(\mathbb{R}^d, \mathcal{H}), \quad \mathbb{P} - \text{almost everywhere}$$

The gradients of  $w_k$  have mean zero and

$$A^* = \lim_{N \rightarrow +\infty} \frac{1}{(2N)^d} \int_{B_N} A(x, \theta)(I + \nabla \mathbf{w}(x, \theta)) dx, \quad \mathbb{P} - \text{almost everywhere}$$

# Numerical simulation of homogenized coefficients

## Neumann and Dirichlet homogenized tensors

$$\check{a}_{N,M} = \frac{1}{M(2N)^d} \sum_{i=1}^M \int_{B_N} A(x, \theta_i) \left( I + \nabla w_N^i(x) \right) dx, \quad w_N^i(x) \text{ computed with KUBC}$$

$$\hat{a}_{N,M} = \left( \frac{1}{M} \sum_{i=1}^M \left( \frac{1}{(2N)^d} \int_{B_N} A(x, \theta_i) \left( I + \nabla w_N^i(x) \right) dx \right)^{-1} \right)^{-1}, \quad w_N^i(x) \text{ computed with SUBC}$$

## Theorem 5 (Convergence of approximate homogenized tensors<sup>13</sup>)

$$\lim_{N \rightarrow \infty} \check{a}_{N,M} = \lim_{N \rightarrow \infty} \hat{a}_{N,M} = a^*$$

The following hierarchy can be shown<sup>14</sup>,

$$\hat{a}_{N,M} \leq a^* \leq \check{a}_{N,M}$$

Except for particular situations, for finite size  $N$ ,

$$\lim_{M \rightarrow \infty} \hat{a}_{N,M} \neq \lim_{M \rightarrow \infty} \check{a}_{N,M} \neq a^*$$

13. A. BOURGEAT et A. PIATNITSKI. "Approximations of effective coefficients in stochastic homogenization". In : *Ann. Inst. Henri Poincaré* 40 (2004), p. 153-165. DOI : 10.1016/j.anihpb.2003.07.003

14. C. HUET. "Application of variational concepts to size effects in elastic heterogeneous bodies". In : *J. Mech. Phys. Solids* 38.6 (1990), p. 813-841. DOI : 10.1016/0022-5096(90)90041-2

# Numerical illustration of convergence of estimates of homogenized tensors <sup>15</sup>

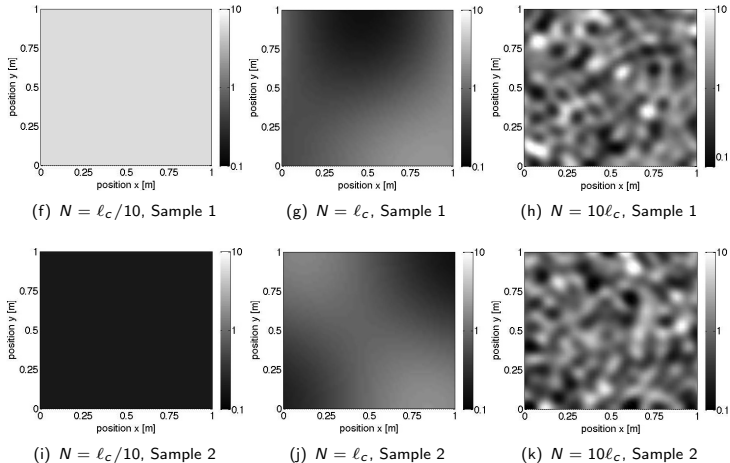


FIGURE – Map of parameter  $A(x, \theta_i)$  (in logarithmic scale) for two realizations of each of the box sizes.

# Numerical illustration of convergence of estimates of homogenized tensors

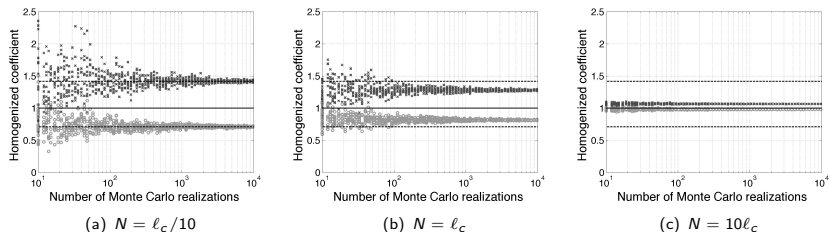


FIGURE – Convergence of the homogenized coefficients  $\check{\alpha}_{N,M}$  (dark grey crosses) and  $\hat{\alpha}_{N,M}$  (light grey circles) for different box sizes  $N$  as a function of the numbers of Monte Carlo trials  $M$ . The dashed lines indicate the values of the arithmetic average  $E[\mathbf{k}_\epsilon]$  and of the harmonic average  $E[\mathbf{k}_\epsilon^{-1}]^{-1}$ . The solid lines indicate the value of  $a^* = 1$ .



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FIGURE – Silicon carbide fibres in a titanium matrix (© ONERA)

we consider a particular medium, of the type

$$A\left(\frac{x}{\epsilon}, \theta\right) = A_{\text{per}}\left(\Phi^{-1}\left(\frac{x}{\epsilon}, \theta\right)\right)$$

where  $A_{\text{per}}(x) \in L^2_{\#}([0, 1]^d)$  is a deterministic periodic material parameter field, and  $\Phi(x, \theta)$  is a random diffeomorphism, whose gradient  $\nabla\Phi$  is assumed stationary in the sense that there exists an ergodic group action  $\tau$  such that,  $\forall k \in \mathbb{Z}^d$ ,

$$\nabla\Phi(x + k, \theta) = \nabla\Phi(x, \tau_k\theta), \quad \text{almost surely, almost everywhere in } x$$

16. X. BLANC, C. LE BRIS et P.-L. LIONS. "Stochastic homogenization and random lattices". In : *Journal Mathématiques Pures Appliquées* 88.1 (2007), p. 34-63. DOI : 10.1016/j.matpur.2007.04.006

## The particular case of perturbations of the identity

- we are particularly interested in the case when (for a "small"  $\epsilon$ )

$$\Phi(y, \theta) = y + \epsilon \Psi(y, \theta) + \mathcal{O}(\epsilon^2)$$

In that case the matrix  $A_{\text{per}}(\Phi^{-1}(y, \theta))$  is formally close to a periodic matrix  $A_{\text{per}}(y)$ .

- The corrector is expanded  $\mathbf{w} = \mathbf{w}_0 + \epsilon \mathbf{w}_1 + \mathcal{O}(\epsilon^2)$  and both  $\mathbf{w}_0$  and  $\bar{\mathbf{w}}_1 = \mathbb{E}[\mathbf{w}_1]$  are shown to be solutions to deterministic periodic corrector problems (NB : tedious computations)

$$-\nabla \cdot (A_{\text{per}}(y) (\mathbf{I}_d + \nabla \mathbf{w}_0)) = 0$$

and

$$\begin{aligned} -\nabla \cdot (A_{\text{per}}(y) \nabla \bar{\mathbf{w}}_1) &= -\nabla \cdot (A_{\text{per}}(y) (\mathbb{E}[\nabla \Psi] \nabla \mathbf{w}_0)) \\ &\quad - \nabla \cdot \left( \left( \mathbb{E}[\nabla \cdot \Psi] \mathbf{I}_d - \mathbb{E}[\nabla \Psi]^T \right) A_{\text{per}}(y) (\mathbf{I}_d + \nabla \mathbf{w}_0) \right) \end{aligned}$$

- The approximation of the homogenized matrix  $A^* = A_0^* + \epsilon A_1^* + \mathcal{O}(\epsilon^2)$  only requires the knowledge of  $\mathbf{w}_0$  and  $\bar{\mathbf{w}}_1$  (NB : tedious computations)

$$A_0^* = \int_Y A_{\text{per}}(y) (\mathbf{I}_d + \nabla \mathbf{w}_0) dy,$$

and

$$\begin{aligned} A_1^* &= -A_0^* \int_Y \mathbb{E}[\nabla \cdot \Psi] dy + \int_Y \mathbb{E}[\nabla \cdot \Psi] A_{\text{per}}(y) (\mathbf{I}_d + \nabla \mathbf{w}_0) dy \\ &\quad + \int_Y A_{\text{per}}(y) (\nabla \bar{\mathbf{w}}_1 - \mathbb{E}[\nabla \Psi] \nabla \mathbf{w}_0) dy \end{aligned}$$

# Modeling of random media and stochastic homogenization for elliptic equations

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Sept. 2020