





Construction of coarse approximations for problems with highly oscillating coefficients

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An Inverse Multiscale Problem

Our study focuses on multiscale systems (e.g. composite materials).

Determining the fine-scale structure from measurements is an ill-posed inverse problem (see [Lions78]¹) but identifying effective parameters is possible!

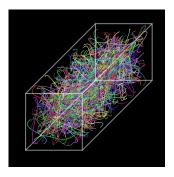


Figure: Scheme of a composite material composed of microfibers randomly intertwined.

J.-L. Lions, Some aspects of modelling problems in distributed parameter systems, 1978.

Multiscale Context: ill-posed inverse problem

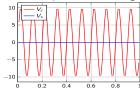
Consider the prototypical linear equation with caracteristic length scall arepsilon

$$\mathcal{L}_{\varepsilon}u_{\varepsilon}=(-\Delta+\varepsilon^{-1}V(\varepsilon^{-1}\cdot))u_{\varepsilon}=f \text{ in } \Omega, \qquad u_{\varepsilon}=0 \text{ on } \partial\Omega,$$

with **periodic** coefficient V such that $\langle V \rangle = 0$, and RHS $f \in L^2(\Omega)$. Homogenization theory² assesses the existence of a limit equation when $\varepsilon \to 0$:

$$\mathcal{L}_{\star}u_{\star}=(-\Delta+rac{V_{\star}}{})u_{\star}=f \ ext{in} \ \Omega, \qquad u_{\star}=0 \ ext{on} \ \partial\Omega,$$

and we have $u_{\varepsilon} \to u_{\star}$ strongly in $L^{2}(\Omega)$ and weakly in $H^{1}(\Omega)$ when $\varepsilon \to 0$.



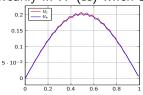


Figure: Two similar solutions associated to two very distinct coefficients.

² A. Bensoussan, J.-L. Lions, G. Papanicolaou, *Asymptotic analysis for periodic structures*, 1978

Our Approach

Consider the problem (1) involving a periodic coefficient V:

$$\mathcal{L}_{\varepsilon}u_{\varepsilon}=\left(-\Delta+\varepsilon^{-1}V\left(\varepsilon^{-1}\cdot\right)\right)u_{\varepsilon}=f\text{ in }\Omega,\qquad u_{\varepsilon}=0\text{ on }\partial\Omega.\tag{1}$$

From the knowledge of solutions u_{ε} for various RHS f, our aim is to propose a numerical methodology to build an effective operator $\overline{\mathcal{L}}$ approaching $\mathcal{L}_{\varepsilon}$ with satisfying L^2 error on the solutions.

Our strategy

- is inspired by homogenization theory,
- does not rely at its core on classical hypothesis for homogenization (such as periodicity), which may be too restrictive in practical situations,
- is valid outside the regime of homogenization (i.e. $\varepsilon \to 0$),
- can be adapted to a wide range of other equations (see [LeBris18]³).

³C. Le Bris, F. Legoll, and S. Lemaire, *On the best constant matrix approximating an oscillatory matrix-valued coefficient in divergence-form operators*, 2018.

Recovering an effective coefficient

Let $\overline{V} \in \mathbb{R}$ be a *constant* coefficient, and $\overline{u} = u(\overline{V}, f)$ be the solution to (2) with RHS $f \in L^2(\Omega)$.

$$\overline{\mathcal{L}}\overline{u} = \left(-\Delta + \overline{V}\right)\overline{u} = f \text{ in } \Omega, \qquad \overline{u} = 0 \text{ on } \partial\Omega. \tag{2}$$

The quality of the approximation of $\mathcal{L}_{\varepsilon}$ by $\overline{\mathcal{L}}$ can be quantified through the functional

$$\sup_{\|f\|_{L^2(\Omega)}=1} \left\|u_{\varepsilon}(f)-u(\overline{V},f)\right\|_{L^2(\Omega)}^2$$

Hence we can minimize the worst case scenario with the optimization problem

$$I_{\varepsilon} = \inf_{\overline{V} \in \mathbb{R}} \sup_{\|f\|_{L^{2}(\Omega)} = 1} \|u_{\varepsilon}(f) - u(\overline{V}, f)\|_{L^{2}(\Omega)}^{2}$$
(3)

The choice of an $L^2(\Omega)$ norm is reminescent of the fact that $\|u_{\varepsilon} - u_{\star}\|_{L^2(\Omega)}$ tends to 0 with ε . (In practice, we slighty modify (3) to recover a quadratic problem.)

Consistency with homogenization in the regime of separated scales

The problem

$$I_{\varepsilon} = \inf_{\overline{V} \in \mathbb{R}} \sup_{\|f\|_{L^{2}(\Omega)} = 1} \|u_{\varepsilon}(f) - u(\overline{V}, f)\|_{L^{2}(\Omega)}^{2}$$
(3)

is consistent with homogenization theory in the sense that :

$$\lim_{\varepsilon \to 0} I_{\varepsilon} = 0. \tag{4}$$

Furthermore, for $\varepsilon>0$ fixed (sufficiently small), there exists a unique minimizer $\overline{V}^0_\varepsilon\in\mathbb{R}$. The following convergence holds :

$$\lim_{\varepsilon \to 0} \overline{V}_{\varepsilon}^{0} = V_{\star}, \tag{5}$$

where V_{\star} is the homogenized coefficient.

Numerical results

To solve

$$I_{\varepsilon} = \inf_{\overline{V} \in \mathbb{R}} \sup_{\|f\|_{L^{2}(\Omega)} = 1} \|u_{\varepsilon}(f) - u(\overline{V}, f)\|_{L^{2}(\Omega)}^{2},$$

we use an Uzawa algorithm (i.e. $\min_{\overline{V}}$ intertwinned with \max_f) in 2D $(\Omega=[0,1]^2)$ using the coefficient

$$V(x,y) = c \left(\sin(2\pi x) + \sin(2\pi y) \right).$$

We approximate the supremum by a maximization over the first eigenmodes of $(-\Delta)$ -operator. In practice, a *single* mode is often sufficient in order to find the *single* coefficient \overline{V} .

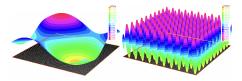


Figure: Coefficient V and Oscillating Coefficient $V_{\varepsilon} = \varepsilon^{-1} V(\varepsilon^{-1})$.

Relationship with homogenized potential

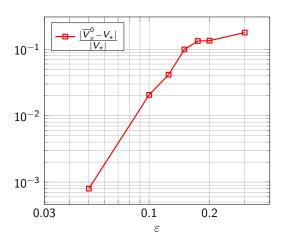


Figure: Error between the homogenized coefficient V_{\star} and the effective coefficient $\overline{V}_{\varepsilon}^{0}$ as a function of ε .

Beyond the regime of separated scales

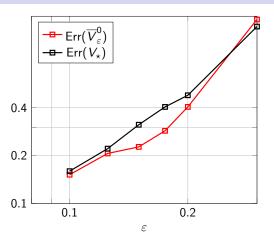


Figure: Error
$$\operatorname{Err}(\mathbf{V}) = \sup_{f \in \operatorname{Span}_{1 \leq \rho \leq 10}(f_{\rho})} \left(\frac{\|u_{\varepsilon}(f) - u(\mathbf{V}, f)\|_{L^{2}(\Omega)}}{\|u_{\varepsilon}(f)\|_{L^{2}(\Omega)}} \right)$$
 as a function of ε for $\mathbf{V} = V_{\star}$ and $\mathbf{V} = \overline{V}_{\varepsilon}^{0}$.

Future Work

In progress:

- Exploring other type of data : boundary measurements, macroscopic data (e.g. energy $\int_{\Omega} |\nabla u_{\varepsilon}|^2 + V_{\varepsilon} u_{\varepsilon}^2$) instead of microscopic ones, ...
- Robustness analysis: what happens if the data is blurred/perturbed/deteriorated/...?
- Exploring pertubative setting.
- Performing classification rather than identification of the effective coefficient.

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An Inverse Multiscale Problem

How to tackle such problems?

General approach:

- Direct Inversion (see [Uhlmann13]),
- Perturbation (see [Ammari08]),

Approach based on homogenization (see [Nolen12]):

- Features Determination (see [Engquist14]),
- Regularization at order 0, identifying effective quantity (see [Ammari16, Caiazzo20]),
- Regularization at order 1, beyond effective quantity: H¹ reconstruction (see [Garnier23, LeBris18]).
- Inverse Homogenization (see [Cherkaev01])