

High-order homogenization in optimal control by the Bloch wave method

MATHLEC-2021

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Classical homogenization problem

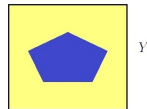
Heterogeneous equation

$$-\nabla \cdot (A^\varepsilon(x) \nabla u_\varepsilon(x)) + u_\varepsilon(x) = f(x) \text{ in } \mathbb{R}^n$$

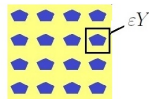
with $f \in L^2(\mathbb{R}^n)$ and $A^\varepsilon(x) := a_Y\left(\frac{x}{\varepsilon}\right)$.

► $a_Y \in L^\infty(\mathbb{R}^n, \mathbb{R}^{n \times n})$ is Y -periodic and elliptic

Unit cell $Y = [0, 1]^n$



Periodic medium



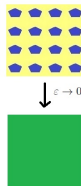
Classical question: What is the effective behavior of u^ε as $\varepsilon \rightarrow 0$?

Classical homogenization theorem

The weak limit u with $u_\varepsilon \rightharpoonup u$ in $H^1(\mathbb{R}^n)$ solves

$$-\nabla \cdot (A^* \nabla u(x)) + u(x) = f(x) \text{ in } \mathbb{R}^n$$

with constant **effective coefficient** $A^* \in \mathbb{R}^{n \times n}$.



A classical corrector result

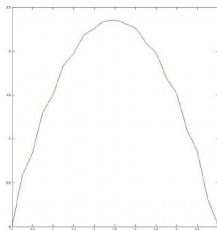
A^* is given in terms of a_Y and periodic **first order correctors** $\chi_j^{(1)}$.

Formula for A^*

$$A_{ij}^* = \int_Y [a_Y(y)(e_j + \nabla \chi_j^{(1)}(y))] \cdot e_i dy$$

Cell problem for the corrector

$$\nabla \cdot (a_Y(y)(e_j + \nabla \chi_j^{(1)}(y))) = 0 \text{ in } Y$$



One can **not** expect strong convergence in H^1 or convergence rates in L^2 better than ε^α with $\alpha < 1$!

Corrector result

$$\|u_\varepsilon - (u + \varepsilon \chi^{(1)}(\frac{\cdot}{\varepsilon}) \cdot \nabla u)\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^2$$

Problem: Adding a second order corrector $\varepsilon^2 \chi^{(2)}(\frac{x}{\varepsilon}) \cdot \nabla^2 u(x)$ will not improve the L^2 -rate!

Higher order estimates

Classical two-scale expansion: $u_\varepsilon(x) \sim \sum_{n \geq 0} \varepsilon^n u_n(x, \frac{x}{\varepsilon})$

Bakhvalov, Panasenko '89

1) $u_0(x, y) = u(x)$ with homogenized solution u .

2) Define $\tilde{u}_k(x) := \int_Y u_k(x, y) dy$. Then

$$-\nabla \cdot (A^* \nabla \tilde{u}_n) = \mathbb{B}_{n+2}^* \nabla^{n+2} u + \sum_{k=1}^{n-1} \mathbb{B}_{n+2-k}^* \nabla^{n+2-k} \tilde{u}_k$$

3) Define $U_\varepsilon(x) := u(x) + \sum_{n \geq 2} \varepsilon^n \tilde{u}_n(x)$. Then

$$u_\varepsilon(x) \sim \sum_{n \geq 0} \varepsilon^n \chi^{(n)}(\frac{x}{\varepsilon}) \cdot \nabla^n U_\varepsilon(x)$$

Separation of

- ▶ Macroscopic profile U_ε
- ▶ Highly oscillatory correctors $\chi^{(n)}$

Higher order estimates: Truncate the series in 3) at $n = N$ to obtain L^2 -error of order ε^{N+1} .

Optimal control

General form

Energy: $\min J^\varepsilon(\theta_\varepsilon, u_\varepsilon)$

Admissible set: $\theta_\varepsilon \in U_{ad} \subset L^2(\mathbb{R}^n)$

Constraint: $-\nabla \cdot (A^\varepsilon(x) \nabla u_\varepsilon(x)) + u_\varepsilon(x) = \theta_\varepsilon(x)$ in $\mathbb{R}^n$

- ▶ J^ε convex and continuous
- ▶ $U_{ad} \subset L^2(\mathbb{R}^d)$ convex and closed
- ▶ θ_ε^* optimal control with solution u_ε^*

Homogenization result. Weak convergences methods provide

- 1) $\theta_\varepsilon^* \rightharpoonup \theta^* \in U_{ad}$ and $u_\varepsilon^* \rightharpoonup u^*$ in $H^1(\mathbb{R}^n)$.
- 2) (θ^*, u^*) solves effective optimal control problem.

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- 2) (θ^*, u^*) solves effective optimal control problem.

Aim: Find (θ_M^*, u_M^*) with $\|\theta_\varepsilon^* - \theta_M^*\|_{L^2(\mathbb{R}^n)} + \|u_\varepsilon^* - u_M^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$.

Problem: Truncating the asymptotic expansion of θ_ε^* and u_ε^* is a bad idea.

Example: $U_{ad} := \{\theta_\varepsilon \in L^2(\mathbb{R}^n) \mid \|u_\varepsilon\|_{L^2(\mathbb{R}^n)} \leq L\}$

$$u_M^* := u^* + \sum_{n=1}^{2M-1} u_n(x, \frac{x}{\varepsilon}) \not\Rightarrow \|u_M^*\|_{L^2(\mathbb{R}^n)} \leq L$$

Bloch expansions in a nutshell

1.) Fourier transform of
 $u : \mathbb{R}^n \rightarrow \mathbb{C}$:

$$u(x) = \int_{\mathbb{R}^n} \hat{u}(\xi) e^{2\pi i \xi \cdot x} d\xi$$

2.) Write ξ as $\xi = k + \eta$ with $k \in \mathbb{Z}^n$ and
 $\eta \in Z := [-\frac{1}{2}, \frac{1}{2})^n$:

$$u(x) = \int_Z \underbrace{\sum_{k \in \mathbb{Z}^n} \hat{u}(k + \eta)}_{=: F = F(x; \eta)} e^{2\pi i k \cdot x} e^{2\pi i \eta \cdot x} d\eta$$

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3.) Expand F in periodic eigenfunctions $\Phi_m(x; \eta)$ of
 $-(\nabla + 2\pi i \eta) \cdot (a_Y(\cdot)(\nabla + 2\pi i \eta))$:

$$F(x; \eta) = \sum_{m \in \mathbb{N}_0} \hat{u}_m(\eta) \Phi_m(x; \eta)$$

$w_m(x; \eta) = \Phi_m(x; \eta) e^{2\pi i \eta \cdot x}$ solves

$$-\nabla \cdot (a_Y(x) \nabla w_m(x; \eta)) = \lambda_m(\eta) w_m(x; \eta)$$

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$$w_m(x; \eta) = \Phi_m(x; \eta) e^{2\pi i \eta \cdot x} \text{ solves } -\nabla \cdot (a_Y(x) \nabla w_m(x; \eta)) = \lambda_m(\eta) w_m(x; \eta)$$

Result: The operator $\mathcal{L} = -\nabla \cdot (a_Y(\cdot) \nabla)$ acts as a multiplier:

$$\mathcal{L}u = \mathcal{L} \int_Z \sum_{m \in \mathbb{N}_0} \hat{u}_m(\eta) w_m(x; \eta) d\eta = \int_Z \sum_{m \in \mathbb{N}_0} \hat{u}_m(\eta) \lambda_m(\eta) w_m(x; \eta) d\eta$$

Bloch adaption

Rescaled Bloch expansion

$$u(x) = \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{u}_m^\varepsilon(\eta) \omega_m^\varepsilon(x; \eta) d\eta$$

$$\begin{aligned} w_m^\varepsilon(x, \eta) &:= w_m(x/\varepsilon, \varepsilon\eta) \\ \lambda_m^\varepsilon(\eta) &:= \frac{1}{\varepsilon^2} \lambda_m(\varepsilon\eta) \end{aligned}$$

The oscillatory operator $\mathcal{L}^\varepsilon := -\nabla \cdot (A^\varepsilon(\cdot) \nabla)$ acts as a multiplier:

$$\mathcal{L}^\varepsilon u = \mathcal{L}^\varepsilon \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{u}_m^\varepsilon(\eta) \omega_m^\varepsilon(x; \eta) d\eta = \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{u}_m^\varepsilon(\eta) \lambda_m^\varepsilon(\eta) \omega_m^\varepsilon(x; \eta) d\eta$$

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Question: Analogue for adding highly oscillatory correctors?

Bloch adaption (Conca, Orive, Vanninathan '02)

Let $u \in L^2(\mathbb{R}^n; \mathbb{C})$ with Fourier transform $\hat{u}$.

$$\mathcal{A}^\varepsilon(u) := \int_{Z/\varepsilon} \hat{u}(\eta) \omega_0^\varepsilon(\cdot; \eta) d\eta$$

For u sufficiently smooth:

$$\begin{aligned} \mathcal{A}^\varepsilon(u) &= u + \varepsilon \chi^{(1)}\left(\frac{\cdot}{\varepsilon}\right) \cdot \nabla u \\ &\quad + \varepsilon^2 \chi^{(2)}\left(\frac{\cdot}{\varepsilon}\right) \cdot \nabla^2 u + O(\varepsilon^3) \end{aligned}$$

Note: For $\text{supp}(\hat{u}) \subset K \subset Z/\varepsilon$ one has

$$\|\mathcal{A}^\varepsilon(u)\|_{L^2(\mathbb{R}^n)} = \|\hat{u}\|_{L^2(Z/\varepsilon)} = \|\hat{u}\|_{L^2(K)} = \|u\|_{L^2(\mathbb{R}^n)}$$

Homogenization problem

Oscillatory optimal control problem

Energy: $\min J^\varepsilon(\theta_\varepsilon, u_\varepsilon)$

Admissible set: $\theta_\varepsilon \in U_{ad}^\varepsilon \subset L^2(\mathbb{R}^n)$

Constraint: $-\nabla \cdot (A^\varepsilon(x) \nabla u_\varepsilon(x)) + u_\varepsilon(x) = \theta_\varepsilon(x)$ in $\mathbb{R}^n$

► Admissible set U_{ad}^ε may depend on $\varepsilon \rightarrow$ next slide

- 1) For fixed constants $\mu_1, \mu_2 \geq 0$ and $\kappa > 0$

$$J^\varepsilon(\theta_\varepsilon, u_\varepsilon) = \frac{\mu_1}{2} E^\varepsilon(u_\varepsilon - u_{d,1}^\varepsilon) + \frac{\mu_2}{2} \|u_\varepsilon - u_{d,2}^\varepsilon\|_{L^2(\mathbb{R}^n)}^2 + \frac{\kappa}{2} \|\theta^\varepsilon\|_{L^2(\mathbb{R}^n)}^2$$

- 2) E^ε is the energy corresponding to the oscillatory PDE-constraint,

$$E^\varepsilon(u) = \int_{\mathbb{R}^n} \nabla u(x) \cdot A^\varepsilon(x) \nabla u(x) + |u(x)|^2 dx$$

- 3) For $u_{d,1}, u_{d,2}$ with **compact support** K in Fourier space

$$u_{d,1}^\varepsilon = \mathcal{A}^\varepsilon(u_{d,1}) \text{ and } u_{d,2}^\varepsilon = \mathcal{A}^\varepsilon(u_{d,2})$$

Remark: Well-prepared data are important for corrector results!

Admissible set

Assumptions

- 1) $U_{ad}^\varepsilon \subset L^2(\mathbb{R}^n)$ convex and closed.
- 2) There exists $\theta_0 \in L^2(\mathbb{R}^n)$ such that $\mathcal{A}^\varepsilon(\theta_0) \in U_{ad}^\varepsilon$ for all $\varepsilon > 0$.
- 3) For every $\theta \in U_{ad}^\varepsilon$ with $\theta(x) = \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{\theta}_m^\varepsilon(\eta) \omega_m^\varepsilon(x; \eta) d\eta$

$$\theta_K := \int_K \hat{\theta}_0^\varepsilon(\eta) \omega_0^\varepsilon(\cdot; \eta) d\eta \in U_{ad}^\varepsilon.$$

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$$\theta_K := \int_K \hat{\theta}_0^\varepsilon(\eta) \omega_0^\varepsilon(\cdot; \eta) d\eta \in U_{ad}^\varepsilon.$$

Example: $U_{ad}^\varepsilon = \{\theta \in L^2(\mathbb{R}^n) \mid \|\theta\|_{L^2(\mathbb{R}^n)} \leq L\}$

- 2) Choose $\theta_0 = 0$.
- 3) $\|\theta_K\|_{L^2(\mathbb{R}^n)} = \|\hat{\theta}_0^\varepsilon\|_{L^2(K)} \leq \sum_{m \in \mathbb{N}_0} \|\hat{\theta}_m^\varepsilon\|_{L^2(Z/\varepsilon)} = \|\theta\|_{L^2(\mathbb{R}^n)} \leq L$

Other examples:

$$U_{ad}^\varepsilon = \{\theta \in L^2(\mathbb{R}^n) \mid \|u_\varepsilon\|_{L^2(\mathbb{R}^n)} \leq L\} \text{ or } U_{ad}^\varepsilon = \{\theta \in L^2(\mathbb{R}^n) \mid E^\varepsilon(u_\varepsilon) \leq L\}$$

M -th order effective problem and main result

1) M -th order effective energy

$$J_M^*(\theta, u) = \frac{\mu_1}{2} E_M^*(u - u_{d,1}) + \frac{\mu_2}{2} \|u - u_{d,2}\|_{L^2(\mathbb{R}^n)}^2 + \frac{\kappa}{2} \|\theta\|_{L^2(\mathbb{R}^n)}^2$$

with $E_M^*(u) = \int_{\mathbb{R}^n} \left(\sum_{k=1}^M \varepsilon^{2k-2} \langle \nabla^k u, \nabla^k u \rangle_{\mathbb{A}_{2k}^*} \right) + |u|^2$

$$u_{d,1}^\varepsilon = \mathcal{A}^\varepsilon(u_{d,1})$$

$$u_{d,2}^\varepsilon = \mathcal{A}^\varepsilon(u_{d,2})$$

- Constant tensors $\mathbb{A}_{2k}^*$ of order $2k$
- $\mathbb{A}_2^* = A^*$ classical effective matrix

2) Effective PDE-constraint

$$\left(\sum_{k=1}^M \varepsilon^{2k-2} (-1)^k \mathbb{A}_{2k}^* \nabla^{2k} u \right) + u = \theta$$

3) Effective admissible set

$$U_{ad}^* = \left\{ \theta \in (\mathcal{A}^\varepsilon)^{-1}(U_{ad}^\varepsilon) \mid \text{supp}(\hat{\theta}) \subset K \right\}$$

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$$\text{with } E_M^*(u) = \int_{\mathbb{R}^n} \left(\sum_{k=1}^M \varepsilon^{2k-2} \langle \nabla^k u, \nabla^k u \rangle_{\mathbb{A}_{2k}^*} \right) + |u|^2$$

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3) Effective admissible set

$$U_{ad}^* = \left\{ \theta \in (\mathcal{A}^\varepsilon)^{-1}(U_{ad}^\varepsilon) \mid \text{supp}(\hat{\theta}) \subset K \right\}$$

Error estimate: Let (θ_M^*, u_M^*) be the solution to the above effective problem.

$$\|\theta_\varepsilon^* - \mathcal{A}^\varepsilon(\theta_M^*)\|_{L^2(\mathbb{R}^n)} + \|u_\varepsilon^* - \mathcal{A}^\varepsilon(u_M^*)\|_{L^2(\mathbb{R}^n)} \leq C \varepsilon^{2M}$$

Example: $U_{ad}^\varepsilon = \{\|\theta\|_{L^2(\mathbb{R}^n)} \leq L\} \Rightarrow U_{ad}^* = \{\|\theta\|_{L^2(\mathbb{R}^n)} \leq L, \text{supp}(\hat{\theta}) \subset K\}$

Outline of the proof

- 1) Expand the optimal control θ_ε^* and u_ε^* in Bloch waves. Show that θ_ε^* and u_ε^* are adaptations,

$$\theta_\varepsilon^* = \mathcal{A}^\varepsilon(\tilde{\theta}_\varepsilon^*) \text{ and } u_\varepsilon^* = \mathcal{A}^\varepsilon(\tilde{u}_\varepsilon^*).$$

- 2) Bloch analysis: Use analyticity of the first Bloch-eigenvalue λ_0^ε to derive the M -th order effective problem.
- 3) Error estimate

$$\|\tilde{\theta}_\varepsilon^* - \theta_M^*\|_{L^2(\mathbb{R}^n)} + \|\tilde{u}_\varepsilon^* - u_M^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$$

by means of a variational inequality in Bloch/Fourier-space.

Important fact: Both minimization problems are constraint by the same admissible set.

Expansion of the solution

$$\begin{aligned}\theta_\varepsilon^*(x) &= \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{\theta}_{\varepsilon,m}^*(\eta) \omega_m^\varepsilon(x; \eta) d\eta \\ u_\varepsilon^*(x) &= \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{u}_{\varepsilon,m}^*(\eta) \omega_m^\varepsilon(x; \eta) d\eta\end{aligned}$$

PDE-constraint

$$\begin{aligned}-\nabla \cdot (A^\varepsilon(x) \nabla u_\varepsilon^*) + u_\varepsilon^* &= \theta_\varepsilon^* \\ \Rightarrow \hat{u}_{\varepsilon,m}^*(\eta) &= \frac{\hat{\theta}_{\varepsilon,m}^*(\eta)}{1 + \lambda_m^\varepsilon(\eta)}\end{aligned}$$

Expansion of the solution

$$\begin{aligned} \theta_\varepsilon^*(x) &= \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{\theta}_{\varepsilon,m}^*(\eta) \omega_m^\varepsilon(x; \eta) d\eta \\ u_\varepsilon^*(x) &= \int_{Z/\varepsilon} \sum_{m \in \mathbb{N}_0} \hat{u}_{\varepsilon,m}^*(\eta) \omega_m^\varepsilon(x; \eta) d\eta \end{aligned}$$

PDE-constraint

$$\begin{aligned} -\nabla \cdot (A^\varepsilon(x) \nabla u_\varepsilon^*) + u_\varepsilon^* &= \theta_\varepsilon^* \\ \Rightarrow \hat{u}_{\varepsilon,m}^*(\eta) &= \frac{\hat{\theta}_{\varepsilon,m}^*(\eta)}{1 + \lambda_m^\varepsilon(\eta)} \end{aligned}$$

Step 1: Reduction of the expansion

$$\hat{\theta}_{\varepsilon,m}^*(\eta) = 0 \text{ for } m \geq 1 \text{ or } \eta \notin K$$

Proof:

- ▶ $J^\varepsilon(\theta_\varepsilon, u_\varepsilon) = \frac{\mu_1}{2} E^\varepsilon(u_\varepsilon - u_{d,1}^\varepsilon) + \frac{\mu_2}{2} \|u_\varepsilon - u_{d,2}^\varepsilon\|_{L^2(\mathbb{R}^n)}^2 + \frac{\kappa}{2} \|\theta^\varepsilon\|_{L^2(\mathbb{R}^n)}^2$
- ▶ Bloch coefficients of $u_{d,1}^\varepsilon, u_{d,2}^\varepsilon$ vanish for $m \geq 1$ or $\eta \notin K$

$$\Rightarrow J^\varepsilon(\underbrace{(\theta_\varepsilon)_K}_{\in U_{ad}^\varepsilon}, (u_\varepsilon)_K) \leq J^\varepsilon(\theta_\varepsilon, u_\varepsilon)$$

Bloch analysis

$$\begin{aligned}\theta_\varepsilon^*(x) &= \int_K \hat{\theta}_{\varepsilon,0}^*(\eta) \omega_0^\varepsilon(x; \eta) d\eta = \mathcal{A}^\varepsilon(\tilde{\theta}_\varepsilon^*) \\ u_\varepsilon^*(x) &= \int_K \frac{\hat{\theta}_{\varepsilon,0}^*(\eta)}{1+\lambda_0^\varepsilon(\eta)} \omega_0^\varepsilon(x; \eta) d\eta = \mathcal{A}^\varepsilon(\tilde{u}_\varepsilon^*)\end{aligned}$$

with

$$\begin{aligned}\tilde{\theta}_\varepsilon^*(x) &= \int_K \hat{\theta}_{\varepsilon,0}^*(\eta) e^{2\pi i \eta \cdot x} d\eta \\ \tilde{u}_\varepsilon^*(x) &= \int_K \frac{\hat{\theta}_{\varepsilon,0}^*(\eta)}{1+\lambda_0^\varepsilon(\eta)} e^{2\pi i \eta \cdot x} d\eta\end{aligned}$$

Taylor expansion of λ_0^ε

$$\lambda_0^\varepsilon(\eta) = \underbrace{(2\pi)^2 \mathbb{A}_2^* \cdot \eta^{\otimes 2} + \varepsilon^2 (2\pi)^4 \mathbb{A}_4^* \cdot \eta^{\otimes 4} + \dots + \varepsilon^{2M-2} (2\pi)^{2M} \mathbb{A}_{2M}^* \cdot \eta^{\otimes 2M}}_{=: P_M(\eta)} + O(\varepsilon^{2M})$$

Approximation: $\tilde{u}_{M,\text{app}}^*(x) := \int_K \frac{\hat{\theta}_{\varepsilon,0}^*(\eta)}{1+P_M(\eta)} e^{2\pi i \eta \cdot x} d\eta$

Error estimate

$$\|\tilde{u}_\varepsilon^* - \tilde{u}_{M,\text{app}}^*\|_{L^2(\mathbb{R}^n)} \leq C \varepsilon^{2M}$$

Observation:

$$\left(\sum_{k=1}^M \varepsilon^{2k-2} (-1)^k \mathbb{A}_{2k}^* \nabla^{2k} \tilde{u}_{M,\text{app}}^* \right) + \tilde{u}_{M,\text{app}}^* = \tilde{\theta}_\varepsilon^*$$

Problem: $\tilde{\theta}_\varepsilon^*$ is related to the original optimal control, not the effective one!

Reformulation in Bloch/Fourier space

1. Original energy:

$$J^\varepsilon(\theta_\varepsilon, u_\varepsilon) = \frac{\mu_1}{2} E^\varepsilon(u_\varepsilon - u_{d,1}^\varepsilon) + \frac{\mu_2}{2} \|u_\varepsilon - u_{d,2}^\varepsilon\|_{L^2(\mathbb{R}^n)}^2 + \frac{\kappa}{2} \|\theta^\varepsilon\|_{L^2(\mathbb{R}^n)}^2$$

For $\theta_\varepsilon(x) = (\theta_\varepsilon)_K(x) = \int_K \hat{\theta}_{\varepsilon,0}(\eta) \omega_0^\varepsilon(x; \eta) d\eta$ one finds

$$\begin{aligned} J^\varepsilon(\theta_\varepsilon, u_\varepsilon) &= \hat{J}^\varepsilon(\hat{\theta}_{\varepsilon,0}) \\ &= \frac{\mu_1}{2} \int_K \left| \frac{\hat{\theta}_{\varepsilon,0}(\eta)}{1 + \lambda_0^\varepsilon(\eta)} - \hat{u}_{d,1}(\eta) \right|^2 (1 + \lambda_0^\varepsilon(\eta)) d\eta \\ &\quad + \frac{\mu_2}{2} \int_K \left| \frac{\hat{\theta}_{\varepsilon,0}(\eta)}{1 + \lambda_0^\varepsilon(\eta)} - \hat{u}_{d,2}(\eta) \right|^2 d\eta + \frac{\kappa}{2} \int_K |\hat{\theta}_{\varepsilon,0}(\eta)|^2 d\eta \end{aligned}$$

2. Effective energy: For $\theta(x) = \int_K \hat{\theta}(\eta) e^{2\pi i \cdot x} d\eta$ one finds

$$\begin{aligned} J_M^*(\theta, u) &= \hat{J}_M^*(\hat{\theta}) \\ &= \frac{\mu_1}{2} \int_K \left| \frac{\hat{\theta}(\eta)}{1 + P_M(\eta)} - \hat{u}_{d,1}(\eta) \right|^2 (1 + P_M(\eta)) d\eta \\ &\quad + \frac{\mu_2}{2} \int_K \left| \frac{\hat{\theta}(\eta)}{1 + P_M(\eta)} - \hat{u}_{d,2}(\eta) \right|^2 d\eta + \frac{\kappa}{2} \int_K |\hat{\theta}(\eta)|^2 d\eta \end{aligned}$$

Admissible sets in Bloch/Fourier space

Lemma

Let $\hat{\theta} \in L^2(K; \mathbb{C})$. Consider

$$1) \theta_B(x) := \int_K \hat{\theta}(\eta) \omega_0^\varepsilon(x; \eta) d\eta$$

$$2) \theta_F(x) := \int_K \hat{\theta}(\eta) e^{2\pi i \eta \cdot x} d\eta$$

Then $\theta_B \in (U_{ad}^\varepsilon)_K \Leftrightarrow \theta_F \in U_{ad}^*$.

Effective admissible set:

$$U_{ad}^* = \left\{ \theta \in (\mathcal{A}^\varepsilon)^{-1}(U_{ad}^\varepsilon) \mid \text{supp}(\hat{\theta}) \subset K \right\}$$

Proof: Since by definition $\theta_B = \mathcal{A}^\varepsilon(\theta_F)$,

$$\theta_F \in U_{ad}^* \Leftrightarrow \theta_F = \int_K \hat{\theta}(\eta) e^{2\pi i \eta \cdot x} d\eta \in (\mathcal{A}^\varepsilon)^{-1}(U_{ad}^\varepsilon) \Leftrightarrow \theta_B \in (U_{ad}^\varepsilon)_K$$

Conclusion: The minimization problems for $\hat{J}^\varepsilon$ and $\hat{J}_M^*$ are constrained by the same set

$$\hat{U}_{ad} := \{ \hat{\theta} \in L^2(K; \mathbb{C}) \mid \theta_F \in U_{ad}^* \} = \{ \hat{\theta} \in L^2(K; \mathbb{C}) \mid \theta_B \in (U_{ad}^\varepsilon)_K \}$$

Estimate for the optimal controls

Theorem

The minimizers $\hat{\theta}_{\varepsilon,0}^*$ and $\hat{\theta}_M^*$ satisfy

$$\|\hat{\theta}_{\varepsilon,0}^* - \hat{\theta}_M^*\|_{L^2(K;\mathbb{C})} \leq C\varepsilon^{2M}$$

$$\hat{\theta}_{\varepsilon,0}^* = \operatorname{argmin}_{\hat{\theta} \in \hat{U}_{ad}} \hat{J}^\varepsilon(\hat{\theta})$$

$$\hat{\theta}_M^* = \operatorname{argmin}_{\hat{\theta} \in \hat{U}_{ad}} \hat{J}_M^*(\hat{\theta})$$

Proof: Apply the variational inequality for $\hat{J}^\varepsilon$ to $\hat{\theta}_M^*$ and for $\hat{J}_M^*$ to $\hat{\theta}_{\varepsilon,0}^*$:

$$\kappa \|\hat{\theta}_{\varepsilon,0}^* - \hat{\theta}_M^*\|_{L^2(K;\mathbb{C})} \leq \tilde{C}\varepsilon^{2M} \left(\|\hat{\theta}_{\varepsilon,0}^*\|_{L^2(K;\mathbb{C})} + \|\hat{u}_{d,2}\|_{L^2(K;\mathbb{C})} \right) \leq C\varepsilon^{2M},$$

where the factor $\tilde{C}\varepsilon^{2M}$ is due to $\sup_{\eta \in K} |\lambda_0^\varepsilon(\eta) - P_M(\eta)| \leq \tilde{C}\varepsilon^{2M}$.

Corollary

$$1) \quad \|\tilde{\theta}_\varepsilon^* - \theta_M^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$$

$$2) \quad \|\theta_\varepsilon^* - \mathcal{A}^\varepsilon(\theta_M^*)\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$$

- ▶ $\theta_\varepsilon^* = \mathcal{A}^\varepsilon(\tilde{\theta}_\varepsilon^*)$ with
$$\tilde{\theta}_\varepsilon^*(x) = \int_K \hat{\theta}_{\varepsilon,0}^*(\eta) e^{2\pi i \eta \cdot x} d\eta$$
- ▶ $\theta_M^* = \int_K \hat{\theta}_M^*(\eta) e^{2\pi i \eta \cdot x} d\eta$ effective optimal control

Estimate for the solutions

Recall:

- ▶ $\theta_\varepsilon^* = \mathcal{A}^\varepsilon(\tilde{\theta}_\varepsilon^*)$ and $u_\varepsilon^* = \mathcal{A}^\varepsilon(\tilde{u}_\varepsilon^*)$
- ▶ First approximation: $\|\tilde{u}_\varepsilon^* - \tilde{u}_{M,\text{app}}^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$, where

$$\left(\sum_{k=1}^M \varepsilon^{2k-2} (-1)^k \mathbb{A}_{2k}^* \nabla^{2k} \tilde{u}_{M,\text{app}}^* \right) + \tilde{u}_{M,\text{app}}^* = \tilde{\theta}_\varepsilon^*$$

Theorem

Let u_M^* be the solution to $\left(\sum_{k=1}^M \varepsilon^{2k-2} (-1)^k \mathbb{A}_{2k}^* \nabla^{2k} \tilde{u}_M^* \right) + \tilde{u}_M^* = \theta_M^*$.

- 1) $\|\tilde{u}_\varepsilon^* - u_M^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$
- 2) $\|u_\varepsilon^* - \mathcal{A}^\varepsilon(u_M^*)\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$

Proof: Use that $\|\tilde{\theta}_\varepsilon^* - \theta_M^*\|_{L^2(\mathbb{R}^n)} \leq C\varepsilon^{2M}$ and exploit a priori estimate for the effective equation.

Well-posedness of the effective problem

M -th order effective equation:

$$(-\mathbb{A}_2^* \nabla^2 u + \varepsilon^2 \mathbb{A}_4^* \nabla^4 u + \cdots + (-1)^M \varepsilon^{2M-2} \mathbb{A}_{2M}^* \nabla^{2M} u) + u = \theta$$

Solution in Fourier space:

$$u(x) = \int_{\mathbb{R}^n} \frac{\hat{\theta}(\eta)}{1 + P_M(\eta)} e^{2\pi i \eta \cdot x} d\eta$$

Problem: $\mathbb{A}_4^*$ is negative semidefinite (Conca, Orive, Vanninathan '06)!

$\Rightarrow$ The effective equation is not well-posed in general!

Lemma

Let $K \subset \mathbb{R}^n$ be compact and $M \in \mathbb{N}$ fixed.

Then for every $\varepsilon < \varepsilon_M$

$$P_M(\eta) \geq \frac{\lambda}{2} |\eta|^2 \quad \text{for all } \eta \in K,$$

where λ is the ellipticity constant of $(2\pi)^2 \mathbb{A}_2^*$.

- ▶ For ε below ε_M and $\text{supp}(\theta) \subset K$ the equation is well posed
- ▶ A priori estimate for $\|u\|_{H^1(\mathbb{R}^n)}$ uniformly in ε .

Substitution trick

Second order effective equation

$$-\mathbb{A}_2^* \nabla^2 u(x) + \varepsilon^2 \mathbb{A}_4^* \nabla^4 u + u = \theta$$

- ▶ $\mathbb{A}_2^*$ positive definite
- ▶ $\mathbb{A}_4^*$ negative semidefinite

Substitution trick

1-D: Replace $\varepsilon^2 \mathbb{A}_4^* \partial_x^4 u = \varepsilon^2 \frac{\mathbb{A}_4^*}{\mathbb{A}_2^*} \partial_x^2 (\mathbb{A}_2^* \partial_x^2 u)$ to lowest order in ε by

Formally: $\varepsilon^2 \mathbb{A}_4^* \partial_x^4 u = \varepsilon^2 \frac{\mathbb{A}_4^*}{\mathbb{A}_2^*} \partial_x^2 (\mathbb{A}_2^* \partial_x^2 u) = \varepsilon^2 \frac{\mathbb{A}_4^*}{\mathbb{A}_2^*} \partial_x^2 (u - \theta) + O(\varepsilon^4)$

Well-posed effective equation

$$\left(-\mathbb{A}_2^* + \varepsilon^2 \frac{\mathbb{A}_4^*}{\mathbb{A}_2^*}\right) \partial_x^2 u + u = \theta + \varepsilon^2 \frac{\mathbb{A}_4^*}{\mathbb{A}_2^*} \partial_x^2 \theta$$

- ▶ The effective energy has to be transformed, too.

n-D: Decomposition $\mathbb{A}_4^* \nabla^4 = -\mathbb{B}_2^* \nabla^2 (\mathbb{A}_2^* \nabla^2) + \mathbb{B}_4^* \nabla^4$ with $\mathbb{B}_2^*, \mathbb{B}_4^* \geq 0$.

$$\text{Effective equation: } (-\mathbb{A}_2^* + \varepsilon^2 \mathbb{B}_2^*) \nabla^2 u + \varepsilon^2 \mathbb{B}_4^* \nabla^4 u + u = (\text{Id} - \varepsilon^2 \mathbb{B}_2^* \nabla^2) \theta$$

Conclusion

- ▶ Derivation of a high order effective optimal control problem via Bloch analysis
- ▶ Bloch adaption of the effective solution required to capture the high order oscillations
- ▶ Effective admissible set given as inverse image under the Bloch adaption

[1] A.Lamacz-Keymlioglu, I. Yousept: High-order homogenization in optimal control by the Bloch wave method. Submitted (2020)

Thank you!