

Long time behavior of a two-phase optimal design for the heat equation

Arnaud Diego MÜNCH

Laboratoire de Mathématiques de Clermont-Ferrand, FRANCE
arnaud.munch@math.univ-bpclermont.fr

joint work with G. ALLAIRE (X-CMAP, Palaiseau) and F. PERIAGO (UPCT, Carthage)

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Motivation of the present work

Related to the work [*Relaxation of an optimal design problem for the heat equation*, J. Math. Pures Appl. (2008) AM-Pedregal-Periago] where the following optimal design problem is analyzed :

$$(P_T) \quad \text{Minimize in } \mathcal{X} \in \mathbf{CD} : \quad J_T(\mathcal{X}) = \frac{1}{T} \int_0^T \int_{\Omega} K(x) \nabla u(t, x) \cdot \nabla u(t, x) \, dx dt$$

$\Omega \subset \mathbb{R}^N$, where the state variable $u = u(t, x)$ is the solution of the system

$$\begin{cases} \beta(x) u'(t, x) - \operatorname{div}(K(x) \nabla u(t, x)) = f(t, x) & \text{in } (0, T) \times \Omega \\ u = 0 & \text{on } (0, T) \times \partial\Omega \\ u(0, x) = u_0(x) & \text{in } \Omega, \end{cases} \quad (1)$$

with

$$\begin{cases} \beta(x) = \mathcal{X}(x) \beta_1 + (1 - \mathcal{X}(x)) \beta_2 \\ K(x) = \mathcal{X}(x) k_1 l_N + (1 - \mathcal{X}(x)) k_2 l_N. \end{cases}$$

$k_i > 0$ - thermal conductivity , $\beta_i = \rho_i c_i$ ($\rho_i > 0$ mass density - $c_i > 0$ specific heat)

The design variable \mathcal{X} indicates the region occupied by the material (β_1, k_1) and is subjected to belong to the class of *classical designs* \mathbf{CD} defined as

$$\mathbf{CD} = \left\{ \mathcal{X} \in L^\infty(\Omega; \{0, 1\}) : \int_{\Omega} \mathcal{X}(x) dx = L|\Omega| \right\}, \quad (2)$$

⇒ Study the **asymptotic behavior as $T \rightarrow \infty$** of the solution (θ_T, K_T^*) of the relaxed problem (RP_T)

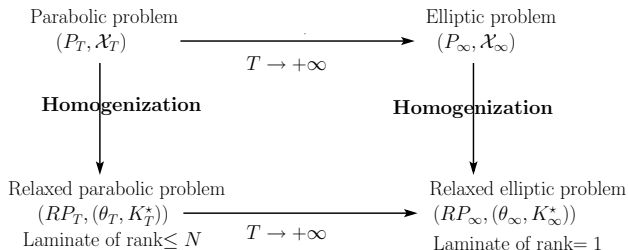


Figure: Commutation between Homogenization process and limit of the heat system as $T \rightarrow \infty$???

⇒ We assume that \mathcal{X} is **time independent** and use tools from Homogenization theory.

- 1 Overview of the relaxed formulation (RP_T) and (RP_∞)
- 2 H -convergence of optimal effective tensors K_T^* toward K_∞^*
- 3 Structure of the optimal effective tensor K_T^* in term of sequential laminates
- 4 (Formal) Analysis of the micro-structure of K_T^* for T arbitrarily large (and small)
- 5 Numerical experiments for $N = 2$
- 6 A word about the open case where \mathcal{X} is time-dependent.

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The relaxed formulation (RP_T) involves the space of *relaxed designs*

$$\mathbf{RD} = \left\{ (\theta, K^*) \in L^\infty(\Omega; [0, 1] \times \mathcal{M}_N^S(k_1, k_2)) : K^*(x) \in \mathbf{G}_{\theta(x)} \text{ a.e. } x \in \Omega, \|\theta\|_{L^1(\Omega)} = L|\Omega| \right\},$$

where $\mathcal{M}_N^S(k_1, k_2)$ is the space of *real symmetric squared matrices* M of order N satisfying, for all $\xi \in \mathbb{R}^N$, $k_1 |\xi|^2 \leq M\xi \cdot \xi$ and $k_2 |\xi|^2 \leq M^{-1}\xi \cdot \xi$.

For a given $\theta \in L^\infty(\Omega; [0, 1])$, the so-called \mathbf{G}_θ -closure is the set of all symmetric matrices with *eigenvalues* $\lambda_1, \dots, \lambda_N$ satisfying

$$\left\{ \begin{array}{l} \lambda_\theta^- \leq \lambda_j \leq \lambda_\theta^+, \quad 1 \leq j \leq N, \\ \sum_{j=1}^N \frac{1}{\lambda_j - k_1} \leq \frac{1}{\lambda_\theta^- - k_1} + \frac{N-1}{\lambda_\theta^+ - k_1}, \\ \sum_{j=1}^N \frac{1}{k_2 - \lambda_j} \leq \frac{1}{k_2 - \lambda_\theta^-} + \frac{N-1}{k_2 - \lambda_\theta^+}, \end{array} \right.$$

where $\lambda_\theta^- = \left(\frac{\theta}{k_1} + \frac{1-\theta}{k_2} \right)^{-1}$ is the harmonic mean and $\lambda_\theta^+ = \theta k_1 + (1-\theta) k_2$ the arithmetic mean of (k_1, k_2) .

The relaxed formulation (RP_T) (T fixed) - Overview

Theorem (PARABOLIC CASE - AM-Pedregal-Periago, JMPA 2008)

The following problem

$$(RP_T) \text{ Minimize in } (\theta, K^*) \in \mathbf{RD} : J_T^*(\theta, K^*) = \frac{1}{T} \int_0^T \int_{\Omega} K^*(x) \nabla u(t, x) \cdot \nabla u(t, x) dx dt$$

where u solves

$$\begin{cases} \beta^*(x) u'(t, x) - \operatorname{div}(K^*(x) \nabla u(t, x)) = f(t, x) & \text{in } (0, T) \times \Omega \\ u = 0 & \text{on } (0, T) \times \partial\Omega \\ u(0, x) = u_0(x) & \text{in } \Omega, \end{cases} \quad (3)$$

with $\beta^*(x) = \theta(x) \beta_1 + (1 - \theta(x)) \beta_2$ is a relaxation of (P_T) in the following sense:

- (i) there exists at least one minimizer for (RP_T) in the space \mathbf{RD} ,
- (ii) up to a subsequence, every minimizing sequence of classical designs \mathcal{X}_n converges, weak- $*$ in $L^\infty(\Omega; [0, 1])$, to a relaxed density θ , and its associated sequence of tensors

$$K_n = \mathcal{X}_n k_1 I_N + (1 - \mathcal{X}_n) k_2 I_N$$

H -converges to an effective tensor K^* such that (θ, K^*) is a minimizer for (RP_T) , and

- (iii) conversely, every relaxed minimizer $(\theta, K^*) \in \mathbf{RD}$ of (RP_T) is attained by a minimizing sequence \mathcal{X}_n of (P_T) in the sense that

$$\begin{cases} \mathcal{X}_n \rightarrow \theta & \text{weak } * \text{ in } L^\infty(\Omega), \\ K_n \xrightarrow{H} K^*. \end{cases}$$

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The limit (P_∞) of (P_T) as $T \rightarrow \infty$ and its relaxation (RP_∞) - Overview

Assuming that the heat source f depends only on the space variable, the unique solution of (1) converges as $t \rightarrow \infty$ to $\bar{u} \in H_0^1(\Omega)$, solution of the stationary equation

$$\begin{cases} -\operatorname{div} (K(x) \nabla \bar{u}(x)) = f(x) & \text{in } \Omega \\ u = 0 & \text{on } \partial\Omega. \end{cases} \quad (4)$$

Associated with this PDE we consider the design problem

$$(P_\infty) \quad \text{Minimize in } \mathcal{X} \in \mathbf{CD} : \quad J_\infty(\mathcal{X}) = \int_{\Omega} K(x) \nabla \bar{u}(x) \cdot \nabla \bar{u}(x) dx.$$

RELAXATION

Consider the following problem

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where $\bar{u} \in H_0^1(\Omega)$ solves

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(RP_∞) is a relaxation of (P_∞) in the sense of the previous theorem. Moreover, the optimal effective tensor for (RP_∞) is obtained in the form of a first-order laminate in any direction orthogonal to $\nabla \bar{u}$.

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Theorem (ELLIPTIC CASE)

Consider the following problem

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where $\bar{u} \in H_0^1(\Omega)$ solves

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We assume henceforth that $f \in L^2(\Omega)$ is time independent and that $u_0 \in L^2(\Omega)$.

Let $\{T_n\}_{n \in \mathbb{N}}$ be an increasing sequence of positive times converging to infinity. For each T_n , problem (RP_{T_n}) has (at least) a minimizer $(\theta_{T_n}, K_{T_n}^*) \in \mathbf{RD}$.

Since $(\theta_{T_n}, K_{T_n}^*)$ is bounded in $L^\infty(\Omega; [0, 1] \times \mathcal{M}_N^s(k_1, k_2))$, up to subsequences still labeled by n , we have

$$\begin{cases} \theta_{T_n} & \rightharpoonup \theta_{T_\infty} \text{ weak-}^* \text{ in } L^\infty(\Omega; [0, 1]) \\ K_{T_n}^* & \xrightarrow{H} K_{T_\infty}^* \end{cases} \quad \text{as } n \rightarrow \infty$$

Weak* Limit

If $(\theta_{T_n}, K_{T_n}^)$ is an optimal solution of (RP_{T_n}) , then any weak limit $(\theta_{T_\infty}, K_{T_\infty}^*)$ of a converging subsequence of $(\theta_{T_n}, K_{T_n}^*)$ is an optimal solution of (RP_∞) .*

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Theorem (Allaire-AM-Periago)

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Lemma

Let u_n be the solution of

$$\begin{cases} \beta_n^*(x) u_n'(t, x) - \operatorname{div} \left(K_{T_n}^*(x) \nabla u_n(t, x) \right) = f(x) & \text{in } (0, \infty) \times \Omega \\ u_n = 0 & \text{on } (0, \infty) \times \partial\Omega \\ u_n(0, x) = u_0(x) & \text{in } \Omega, \end{cases} \quad (5)$$

with $\beta_n^*(x) = \theta_{T_n}(x) \beta_1 + (1 - \theta_{T_n}(x)) \beta_2$. Then,

$$\lim_{n \rightarrow \infty} \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) \, dx dt = \int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) \, dx, \quad (6)$$

where $\bar{u}_\infty(x) \in H_0^1(\Omega)$ is the solution of

$$\begin{cases} -\operatorname{div} \left(K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \right) = f(x) & \text{in } \Omega \\ \bar{u}_\infty = 0 & \text{on } \partial\Omega. \end{cases} \quad (7)$$

Convergence of $I_1^n \rightarrow 0$ as $T_n \rightarrow \infty$

We decompose

$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) dx dt - \int_{\Omega} K_{T_n}^*(x) \nabla \bar{u}_n(x) \cdot \nabla \bar{u}_n(x) dx = I_1^n + I_2^n$$

where

$$I_1^n = \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^* \nabla u_n(t, x) \cdot \nabla u_n(t, x) dx dt - \int_{\Omega} K_{T_n}^* \nabla \bar{u}_n(x) \cdot \nabla \bar{u}_n(x) dx$$

$$I_2^n = \int_{\Omega} K_{T_n}^* \nabla \bar{u}_n(x) \cdot \nabla \bar{u}_n(x) dx - \int_{\Omega} K_{T_n}^*(x) \nabla \bar{u}_n(x) \cdot \nabla \bar{u}_n(x) dx.$$

where \bar{u}_n solves

$$\begin{cases} -\operatorname{div} (K_{T_n}^* \nabla \bar{u}_n) = f & \text{in } \Omega \\ \bar{u}_n = 0 & \text{on } \partial\Omega. \end{cases} \quad (8)$$

To show that $I_1^n \rightarrow 0$, we prove that there exist $C_1, C_2 > 0$, independent of n , such that

$$\|u_n(t) - \bar{u}_n\|_{L^2(\Omega)} \leq C_1 e^{-C_2 t}, \quad t > 0, \quad (9)$$

The function $v_n(t, x) = u_n(t, x) - \bar{u}_n(x)$ solves

$$\begin{cases} \beta_n^*(x) v_n'(t, x) - \operatorname{div} (K_{T_n}^*(x) \nabla v_n(t, x)) = 0 & \text{in } (0, \infty) \times \Omega \\ v_n = 0 & \text{on } (0, \infty) \times \partial\Omega \\ v_n(0, x) = u_0(x) - \bar{u}_n(x) & \text{in } \Omega. \end{cases}$$

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$$\begin{cases} \beta_n^* (x) v_n' (t, x) - \operatorname{div} (K_{T_n}^* (x) \nabla v_n (t, x)) = 0 & \text{in } (0, \infty) \times \Omega \\ v_n = 0 & \text{on } (0, \infty) \times \partial\Omega \\ v_n (0, x) = u_0 (x) - \bar{u}_n (x) & \text{in } \Omega. \end{cases}$$

Convergence of $I_n^n \rightarrow 0$ as $T_n \rightarrow \infty$

Using the Fourier method,

$$v_n(t, x) = \sum_{k=1}^{\infty} a_n^k e^{-\lambda_n^k t} \omega_n^k(x), \quad \omega_n^k \in H_0^1(\Omega), \quad \|\omega_n^k\|_{L^2_{\beta_n^*}(\Omega)}^2 = \int_{\Omega} \beta_n^* |\omega_n^k|^2 dx = 1$$

$$\begin{cases} -\operatorname{div}(K_{T_n}^* \nabla \omega_n^k) = \lambda_n^k \beta_n^* \omega_n^k & \text{in } \Omega \\ \omega_n^k = 0 & \text{on } \partial\Omega, \end{cases}$$

with $0 < \lambda_n^1 < \lambda_n^2 \leq \lambda_n^3 \leq \dots$, its associated eigenvalues, and

$$a_n^k = \int_{\Omega} \beta_n^*(x) (u_0(x) - \bar{u}_n(x)) \omega_n^k(x) dx, \quad k, n \in \mathbb{N}.$$

Using that $\beta_1 \leq \beta_n^*(x)$ a.e. $x \in \Omega$ and Parseval's identity, we have

$$\beta_1 \|v_n(t)\|_{L^2(\Omega)}^2 \leq \|v_n(t)\|_{L^2_{\beta_n^*}(\Omega)}^2 = \sum_{k=1}^{\infty} e^{-2\lambda_n^k t} |a_n^k|^2 \leq e^{-2\lambda_n^1 t} \|u_0 - \bar{u}_n\|_{L^2_{\beta_n^*}(\Omega)}^2.$$

Since $K_{T_n}^* \xrightarrow{H} K_{T_\infty}^*$ and $0 < \beta_1 \leq \beta_n^*(x) \leq \beta_2$ a.e. $x \in \Omega$, the term $\|u_0 - \bar{u}_n\|_{L^2_{\beta_n^*}(\Omega)}^2$ is uniformly bounded. Moreover, the uniform ellipticity of the sequence of tensors $K_{T_n}^*$ lead to

$$\lambda_n^1 = \min_{\varphi \neq 0, \varphi \in H_0^1} \frac{\int_{\Omega} K_{T_n}^* \nabla \varphi \cdot \nabla \varphi}{\|\varphi\|_{L^2_{\beta_n^*}(\Omega)}^2} \geq \frac{k_1}{\beta_2} \min_{\varphi \neq 0, \varphi \in H_0^1} \frac{\int_{\Omega} \nabla \varphi \cdot \nabla \varphi}{\|\varphi\|_{L^2(\Omega)}^2} = \frac{k_1}{\beta_2} \lambda_1,$$

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Using the weak form of (8), and multiplying the heat equation in system (5) by $u_n(t, x)$ and integrating by parts,

$$I_1^n = \frac{1}{2} \frac{1}{T_n} \int_{\Omega} \beta_n^* (u_0^2(x) - u_n^2(T_n, x)) dx + \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} f(x) (u_n(t, x) - \bar{u}_n(x)) dx dt.$$

By (9) and the boundedness of $\|\bar{u}_n\|_{L^2(\Omega)}$, the first term in the right-hand side of this expression converges to zero as $T_n \rightarrow \infty$. Using once again (9) and the Cauchy-Schwartz inequality,

$$\begin{aligned} \left| \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} f(x) (u_n(t, x) - \bar{u}_n(x)) dx dt \right| &\leq \|f\|_{L^2(\Omega)} \frac{1}{T_n} \int_0^{T_n} \|u_n(t) - \bar{u}_n\|_{L^2(\Omega)} dt \\ &\leq \|f\|_{L^2(\Omega)} \frac{1}{T_n} \int_0^{T_n} C_1 e^{-C_2 t} dt \\ &\rightarrow 0 \quad \text{as } T_n \rightarrow \infty. \end{aligned}$$

Asymptotics of (θ_T, K_T^*) for $T \rightarrow \infty$: proof of the Theorem

Assume that $(\theta_{T_\infty}, K_{T_\infty}^*)$ is not a solution of (RP_∞) . Then, there exists another $(\widehat{\theta}, \widehat{K}^*) \in \mathbf{RD}$ and $\varepsilon > 0$ such that

$$\int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) dx = \int_{\Omega} \widehat{K}^*(x) \nabla \widehat{u}(x) \cdot \nabla \widehat{u}(x) dx + \varepsilon,$$

where $\widehat{u}(x)$ is the solution of the elliptic equation with conductivity \widehat{K}^* . By (6), there exists $n_0 \in \mathbb{N}$ such that for all $n \geq n_0$

$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) dx > \int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) dx - \frac{\varepsilon}{3}.$$

Now let $u(t, x)$ solve

$$\begin{cases} \widehat{\beta}^*(x) u'(t, x) - \operatorname{div} (\widehat{K}^*(x) \nabla u(t, x)) = f(x) & \text{in } (0, T) \times \Omega \\ u = 0 & \text{on } (0, T) \times \partial\Omega \\ u(0, x) = u_0(x) & \text{in } \Omega, \end{cases}$$

with $\widehat{\beta}^*(x) = \widehat{\theta}(x) \beta_1 + (1 - \widehat{\theta}(x)) \beta_2$. Then, multiplying this equation by $u(t, x)$ and integrating by parts, we get the convergence

$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} \widehat{K}^*(x) \nabla u(t, x) \cdot \nabla u(t, x) dx dt \rightarrow \int_{\Omega} \widehat{K}^*(x) \nabla \widehat{u}(x) \cdot \nabla \widehat{u}(x) dx \quad \text{as } n \rightarrow \infty.$$

Asymptotics of (θ_T, K_T^*) for $T \rightarrow \infty$: proof of the Theorem

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$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) dx > \int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) dx - \frac{\varepsilon}{3}.$$

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$$\begin{cases} \hat{\beta}^*(x) u'(t, x) - \operatorname{div} \left(\hat{K}^*(x) \nabla u(t, x) \right) = f(x) & \text{in } (0, T) \times \Omega \\ u = 0 & \text{on } (0, T) \times \partial\Omega \\ u(0, x) = u_0(x) & \text{in } \Omega, \end{cases}$$

with $\hat{\beta}^*(x) = \hat{\theta}(x) \beta_1 + (1 - \hat{\theta}(x)) \beta_2$. Then, multiplying this equation by $u(t, x)$ and integrating by parts, we get the convergence

$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} \hat{K}^*(x) \nabla u(t, x) \cdot \nabla u(t, x) dx dt \rightarrow \int_{\Omega} \hat{K}^*(x) \nabla \hat{u}(x) \cdot \nabla \hat{u}(x) dx \quad \text{as } n \rightarrow \infty.$$

Therefore, there exists $n_1 \in \mathbb{N}$ such that for all $n \geq n_1$

$$\frac{1}{T_n} \int_0^{T_n} \int_{\Omega} \widehat{K}^*(x) \nabla u(t, x) \cdot \nabla u(t, x) \, dx dt < \int_{\Omega} \widehat{K}^*(x) \nabla \widehat{u}(x) \cdot \nabla \widehat{u}(x) \, dx + \frac{\varepsilon}{3}.$$

Hence, for $n \geq \max(n_0, n_1)$ we have

$$\begin{aligned} \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} \widehat{K}^*(x) \nabla u(t, x) \cdot \nabla u(t, x) \, dx dt &< \int_{\Omega} \widehat{K}^*(x) \nabla \widehat{u}(x) \cdot \nabla \widehat{u}(x) \, dx + \frac{\varepsilon}{3} \\ &= \int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) \, dx - \varepsilon + \frac{\varepsilon}{3} \\ &< \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) \, dx dt - \frac{\varepsilon}{3} \end{aligned}$$

which contradicts the fact that $(\theta_{T_n}, K_{T_n}^*)$ is an optimal solution of (RP_{T_n}) . ■

Therefore, there exists $n_1 \in \mathbb{N}$ such that for all $n \geq n_1$

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Hence, for $n \geq \max(n_0, n_1)$ we have

$$\begin{aligned} \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} \widehat{K}^*(x) \nabla u(t, x) \cdot \nabla u(t, x) \, dx dt &< \int_{\Omega} \widehat{K}^*(x) \nabla \widehat{u}(x) \cdot \nabla \widehat{u}(x) \, dx + \frac{\varepsilon}{3} \\ &= \int_{\Omega} K_{T_\infty}^*(x) \nabla \bar{u}_\infty(x) \cdot \nabla \bar{u}_\infty(x) \, dx - \varepsilon + \frac{\varepsilon}{3} \\ &< \frac{1}{T_n} \int_0^{T_n} \int_{\Omega} K_{T_n}^*(x) \nabla u_n(t, x) \cdot \nabla u_n(t, x) \, dx dt - \frac{\varepsilon}{3} \end{aligned}$$

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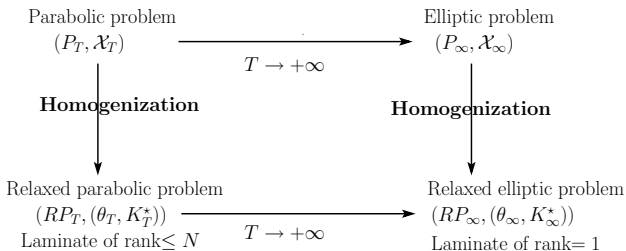


Figure: Commutation between Homogenization process and limit of the heat system as $T \rightarrow \infty$???

What about the structure of the optimal effective tensor K_T^* and its behavior w.r.t. T ?

$$\bar{J}_T^*(\theta, K^*) = \frac{1}{T} \int_0^T \int_{\Omega} K^* \nabla u \cdot \nabla u \, dx dt + I \int_{\Omega} \theta(x) \, dx. \quad (10)$$

Theorem

The objective function $\bar{J}_T^*(\theta, K^*)$ is Gâteaux differentiable on the space of admissible relaxed designs **RD** and

$$\delta \bar{J}_T^*(\theta, K^*) = \int_{\Omega} \left[I - 2(\beta_2 - \beta_1) \frac{1}{T} \int_0^T u' p dt \right] \delta \theta \, dx + \frac{1}{T} \int_0^T \int_{\Omega} \delta K^* \nabla u \cdot (2\nabla p + \nabla u) \, dx dt \quad (11)$$

where $\delta \theta$ and δK^* are admissible increments in **RD** and p the solution of the adjoint equation

$$\begin{cases} -\beta^* p' - \operatorname{div}(K^* \nabla p) = \operatorname{div}(K^* \nabla u) & \text{in } (0, T) \times \Omega \\ p = 0 & \text{on } (0, T) \times \partial \Omega \\ p(T) = 0 & \text{in } \Omega. \end{cases} \quad (12)$$

Consequently, if (θ, K^*) is a minimizer of the function \bar{J}_T^* , it must satisfy $\delta \bar{J}_T^*(\theta, K^*) \geq 0$ for any admissible increments $\delta \theta, \delta K^*$.

Definition

Let $(\theta, K^*) \in \mathbf{RD}$ satisfy the optimality condition $\delta \bar{J}_T^*(\theta, K^*) \geq 0$. For any fixed $T > 0$, we introduce the symmetric matrix of order N

$$M_T = -\frac{1}{T} \int_0^T \nabla u \odot (2\nabla p + \nabla u) dt \quad (13)$$

where \odot denotes the symmetrized tensor product of two vectors, with entries

$$(M_T)_{ij} = -\frac{1}{2T} \int_0^T [(\nabla u)_i (2\nabla p + \nabla u)_j + (\nabla u)_j (2\nabla p + \nabla u)_i] dt, \quad 1 \leq i, j \leq N$$

where u and p are its associated state and adjoint state, respectively.

Remark

If $(f, u_0) \in (L^2(\Omega))^2$, then $u \in L^2(0, T; H_0^1(\Omega)) \cap C^0(0, T; L^2(\Omega))$ and then $p \in L^2(0, T; H_0^1(\Omega))$. This implies that $M_T \in L^1(\Omega)$.

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Characterization of the effective optimal tensor K_T^* in term of sequential laminates

Theorem (ORDER OF LAMINATION)

Let (θ_T, K_T^*) be a minimizer of \bar{J}_T^* and let u and p be its associated state and adjoint state, respectively.

1 (θ_T, K_T^*) satisfies the following characterization

$$K_T^* : M_T = \max_{K^0 \in G_{\theta_T}} K^0 : M_T \quad \text{a.e. } x \in \Omega$$

where $M_T \in L^1(\Omega; \mathbb{R}^{N \times N})$ is given by (13) and $:$ the full contraction of two matrices.

2 K_T^* is a tensor corresponding to a sequential laminate of rank at most N with lamination directions given by the eigenvectors of M_T .

3 The function

$$\theta_T \longmapsto f(\theta_T, M_T) \equiv \max_{K^0 \in G_{\theta_T}} K^0 : M_T$$

is $C^1([0, 1])$ and the optimal density θ_T satisfies

$$\begin{cases} \theta_T(x) = 0 & \text{if and only if} & Q_T(x) > 0 \\ \theta_T(x) = 1 & \text{if and only if} & Q_T(x) < 0 \\ 0 \leq \theta_T(x) \leq 1 & \text{if} & Q_T(x) = 0 \end{cases} \quad (14)$$

and $Q_T(x) = 0$ if $0 < \theta_T(x) < 1$, where Q_T is given by

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$$Q_T(x) = \frac{1}{2} (\beta_+ - \beta_-) \int_{\beta_-}^{\beta_+} u' \, p \, dt + \frac{\partial f(\theta_T, M_T)}{\partial \theta_T} \quad (15)$$

The case where $x \in \Omega$ is such that $M_T(x) \neq 0$

We fix θ and consider the path $K_T^*(s) = sK^0 + (1-s)K_T^*$ for any $K^0 \in G_\theta$, $0 \leq s \leq 1$ and for K_T^* an optimal tensor for (RP_T) . Consequently, $\delta\theta = 0$ and $\delta K_T^* = K^0 - K_T^*$. The optimality condition $\delta \bar{J}_T^*(\theta, K_T^*) \geq 0$ implies that

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Since M_T is well-defined and it belongs to $L^1(\Omega)$, we may therefore apply the Localization principle to conclude that (16) is equivalent a.e. $x \in \Omega$ to the following characterization of the optimal tensor K_T^*

$$K_T^* : M_T = \max_{K^0 \in G_\theta} K^0 : M_T, \quad \text{a.e. } x \in \Omega. \quad (17)$$

It is known that the optimal tensor K_T^* of (17) must be simultaneously diagonalizable with M_T . Consequently, if $(e_j)_{1 \leq j \leq N}$ is a basis of eigenvectors of M_T with associated eigenvalues $(\mu_j)_{1 \leq j \leq N}$, then (17) transforms into

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We fix θ and consider the path $K_T^*(s) = sK^0 + (1-s)K_T^*$ for any $K^0 \in G_\theta$, $0 \leq s \leq 1$ and for K_T^* an optimal tensor for (RP_T) . Consequently, $\delta\theta = 0$ and $\delta K_T^* = K^0 - K_T^*$. The optimality condition $\delta \bar{J}_T^*(\theta, K_T^*) \geq 0$ implies that

$$\int_{\Omega} K_T^* : M_T dx \geq \int_{\Omega} K^0 : M_T dx \quad \forall K^0 \in G_\theta. \quad (16)$$

Since M_T is well-defined and it belongs to $L^1(\Omega)$, we may therefore apply the Localization principle to conclude that (16) is equivalent a.e. $x \in \Omega$ to the following characterization of the optimal tensor K_T^*

$$K_T^* : M_T = \max_{K^0 \in G_\theta} K^0 : M_T, \quad \text{a.e. } x \in \Omega. \quad (17)$$

It is known that the optimal tensor K_T^* of (17) must be simultaneously diagonalizable with M_T . Consequently, if $(e_j)_{1 \leq j \leq N}$ is a basis of eigenvectors of M_T with associated eigenvalues $(\mu_j)_{1 \leq j \leq N}$, then (17) transforms into

$$K_T^* : M_T = \max_{(\lambda_j) \in G_\theta} \sum_{j=1}^N \lambda_j \mu_j, \quad (\lambda_j)_{1 \leq j \leq N} \in \sigma(K^0), K^0 \in G_\theta$$

The case where $x \in \Omega$ is such that $M_T(x) \neq 0$

$$K_T^* : M_T = \max_{(\lambda_j) \in G_\theta} \sum_{j=1}^N \lambda_j \mu_j, \quad (\lambda_j)_{1 \leq j \leq N} \in \sigma(K^0), K^0 \in G_\theta \quad (18)$$

Assume that $x \in \Omega$ is such that $M_T(x) \neq 0$. Since the cost function in (18) is linear and the set G_θ convex, the solution belongs to the boundary of G_θ . This implies that K_T^* corresponds to a sequential laminate of rank at most N with lamination directions given by the eigenvectors of M_T .

Assume that $x \in \Omega$ is such that $M_T(x) = 0$. one can not conclude directly from the relation (18) which degenerates. However, in that case and in dimension $N = 2$, the optimal tensor $K_T^* \in G_\theta$ may be replaced by a tensor which belongs to the boundary of G_θ without changing the value of the objective function. Indeed, assume that K_T^* belongs to the interior of G_θ

$$\frac{1}{\lambda_1 - k_1} + \frac{1}{\lambda_2 - k_1} < g^-(\theta) \equiv \frac{1}{\lambda^-(\theta) - k_1} + \frac{1}{\lambda^+(\theta) - k_1}$$

and that

$$\frac{1}{k_2 - \lambda_1} + \frac{1}{k_2 - \lambda_2} < g^+(\theta) \equiv \frac{1}{k_2 - \lambda^-(\theta)} + \frac{1}{k_2 - \lambda^+(\theta)}.$$

Since the continuous function g^- is strictly increasing and satisfies $g^-(0) = 2/(k_2 - k_1) \leq (\lambda_1 - k_1)^{-1} + (\lambda_2 - k_1)^{-1}$, there exists $\theta^- \in (0, \theta)$ such that

$$\frac{1}{\lambda_1 - k_1} + \frac{1}{\lambda_2 - k_1} = g^-(\theta^-).$$

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The case where $x \in \Omega$ is such that $M_T(x) \neq 0$

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and that

$$\frac{1}{k_2 - \lambda_1} + \frac{1}{k_2 - \lambda_2} < g^+(\theta) \equiv \frac{1}{k_2 - \lambda^-(\theta)} + \frac{1}{k_2 - \lambda^+(\theta)}.$$

Since the continuous function g^- is strictly increasing and satisfies $g^-(0) = 2/(k_2 - k_1) \leq (\lambda_1 - k_1)^{-1} + (\lambda_2 - k_1)^{-1}$, there exists $\theta^- \in (0, \theta)$ such that

$$\frac{1}{\lambda_1 - k_1} + \frac{1}{\lambda_2 - k_1} = g^-(\theta^-).$$

Similarly, since the continuous function g^+ is strictly decreasing and satisfies $g^+(1) = 2/(k_2 - k_1) \leq (k_2 - \lambda_1)^{-1} + (k_2 - \lambda_2)^{-1}$, there exists $\theta^+ \in (\theta, 1)$ such that

$$\frac{1}{k_2 - \lambda_1} + \frac{1}{k_2 - \lambda_2} = g^+(\theta^+).$$

Consequently, at the point x where $M_T(x) = 0$, we may consider the composite with materials k_1 and k_2 in proportions θ^- and $(1 - \theta^-)$, respectively, or the composite with materials k_1 and k_2 in proportions $1 - \theta^+$ and θ^+ . Notice that this choice allows us to ensure that the volume constraint $\|\theta\|_{L^1(\Omega)} = L|\Omega|$ holds. In both cases, the eigenvalues λ_1, λ_2 of K_T^* remain unchanged and so the value of the cost \bar{J}_T^* .

Lemma

For any T , we note by $\mu_1^T, \mu_2^T, \mu_1^T \leq \mu_2^T$ the eigenvalues of the matrix M_T of order $N = 2$ defined by (13). The solution of the linear problem

$$\max_{K^0 \in G_{\theta_T}} K^0 : M_T = \max_{(v_1^T, v_2^T) \in G_{\theta_T}} v_1^T \mu_1^T + v_2^T \mu_2^T \quad \text{is given by}$$

$$\left\{ \begin{array}{l} (v_1^T, v_2^T) = (k_2, k_2) + \frac{\sqrt{\mu_1^T} + \sqrt{\mu_2^T}}{(\lambda_{\theta_T}^+ - k_2)^{-1} + (\lambda_{\theta_T}^- - k_2)^{-1}} \left(\frac{1}{\sqrt{\mu_1^T}}, \frac{1}{\sqrt{\mu_2^T}} \right) \\ \quad \text{if } \mu_1^T \geq 0 \quad \text{and} \quad \sqrt{\mu_1^T}(k_2 - \lambda_{\theta_T}^-) > \sqrt{\mu_2^T}(k_2 - \lambda_{\theta_T}^+) \text{ [Second order laminate]} \\ (v_1^T, v_2^T) = (k_1, k_1) + \frac{\sqrt{-\mu_1^T} + \sqrt{-\mu_2^T}}{(\lambda_{\theta_T}^+ - k_1)^{-1} + (\lambda_{\theta_T}^- - k_1)^{-1}} \left(\frac{1}{\sqrt{-\mu_1^T}}, \frac{1}{\sqrt{-\mu_2^T}} \right) \\ \quad \text{if } \mu_2^T \leq 0 \quad \text{and} \quad \sqrt{-\mu_1^T}(\lambda_{\theta_T}^- - k_1) < \sqrt{-\mu_2^T}(\lambda_{\theta_T}^+ - k_1) \text{ [Second order laminate]} \\ (v_1^T, v_2^T) = (\lambda_{\theta_T}^-, \lambda_{\theta_T}^+) \quad \text{else. [First order laminate]} \end{array} \right.$$

(19)

Structure of the matrix M_T as $T \rightarrow \infty$ for $N = 2$ (Formal analysis)

For any T fixed, we consider the normalized eigenfunctions $(w_m^T)_{m>0}$ and corresponding eigenvalues $(\lambda_m^T)_{m>0}$ of

$$\begin{cases} -\operatorname{div}(K_T^* \nabla w_m^T) = \lambda_m^T \beta_T^* w_m^T & \text{in } \Omega, \\ w_m^T = 0 & \text{on } \partial\Omega \end{cases}$$

where K_T^* is the optimal tensor for (RP_T) . Since $K_T^* \in \partial G_{\theta_T}$, v_1^T, v_2^T are uniformly bounded with respect to T as well as $\{\lambda_m^T\}_m$ and $(w_m^T)_{m>0}$ in $H_0^1(\Omega)$.

Assume that the source f and the initial datum u_0 are expanded as follows:

$$\begin{aligned} f(x) &= \sum_{m>0} f_m^T w_m^T(x), & u_0(x) &= \sum_{m>0} a_m^T w_m^T(x), & \{a_m^T\}_{m>0}, \{f_m^T\}_{m>0} &\in \ell^2(\mathbb{N}), \\ u(t, x) &= \sum_{m>0} a_m^T(t) w_m^T(x), & \rho(t, x) &= \sum_{m>0} b_m^T(t) w_m^T(x) \end{aligned} \quad (20)$$

We rewrite the symmetric matrix M_T as follows:

$$-M_T(x) = \begin{pmatrix} \sum_{m,n>0} c_{mn}^T (w_m^T)_{x_1} (w_n^T)_{x_1} & \frac{1}{2} \sum_{m,n>0} c_{mn}^T \left((w_m^T)_{x_1} (w_n^T)_{x_2} + (w_m^T)_{x_2} (w_n^T)_{x_1} \right) \\ \text{sym.} & \sum_{m,n>0} c_{mn}^T (w_m^T)_{x_2} (w_n^T)_{x_2} \end{pmatrix}.$$

$$\text{with } c_{mn}^T = \frac{1}{T} \int_0^T a_m^T(t) (a_n^T(t) + 2b_n^T(t)) dt, \quad m, n > 0$$

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$$\text{with } c_{mn}^T = \frac{1}{T} \int_0^T a_m^T(t) (a_n^T(t) + 2b_n^T(t)) dt, \quad m, n > 0$$

Structure of the matrix $M_T - T$ large

$$c_{mn}^T = -\frac{f_m^T}{\lambda_m^T} \frac{f_n^T}{\lambda_n^T} + \frac{1}{T} \frac{f_n^T}{(\lambda_m^T)^2 (\lambda_n^T)^2} (\lambda_n^T f_m^T - \lambda_m^T a_m^T \lambda_n^T) + O(e^{-\lambda_m T}, e^{-\lambda_n T})$$

where the coefficients $c_{mn}^{i,T}$ are bounded with respect to T . Notice that only the coefficients of the heat source f are involved in the first order terms $c_{mn}^{0,T}$. We put

$$M_T(x) = M_T^0(x) + \frac{M_T^1(x)}{T} + M^2(T, x), \quad x \in \Omega.$$

Now, we observe that the symmetric matrix $M_T^0(x)$, which is given by

$$M_T^0(x) = - \begin{pmatrix} \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_1} \right)^2 & \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_1} \right) \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_2} \right) \\ \text{SYM} & \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_2} \right)^2 \end{pmatrix}$$

is singular: $\det(M_T^0(x)) = 0$ so that $\arg(\max_{K^0 \in G_{\theta_T}} K^0 : M_T^0) = (\lambda_{\theta_T}^-, \lambda_{\theta_T}^+)$

(Valadier)

For any $T \geq \sup_{x \in \Omega} T^+(x)$, the solution of the problem $\max_{K^0 \in G_{\theta_T}} K^0 : M_T$ is $(\lambda_{\theta_T}^-, \lambda_{\theta_T}^+)$ so that the optimal tensor K_T^+ is a first order laminate.



Structure of the matrix $M_T - T$ large

$$c_{mn}^T = -\frac{f_m^T}{\lambda_m^T} \frac{f_n^T}{\lambda_n^T} + \frac{1}{T} \frac{f_n^T}{(\lambda_m^T)^2 (\lambda_n^T)^2} (\lambda_n^T f_m^T - \lambda_m^T a_m^T \lambda_n^T) + O(e^{-\lambda_m T}, e^{-\lambda_n T})$$

where the coefficients $c_{mn}^{i,T}$ are bounded with respect to T . Notice that only the coefficients of the heat source f are involved in the first order terms $c_{mn}^{0,T}$. We put

$$M_T(x) = M_T^0(x) + \frac{M_T^1(x)}{T} + M^2(T, x), \quad x \in \Omega.$$

Now, we observe that the symmetric matrix $M_T^0(x)$, which is given by

$$M_T^0(x) = - \begin{pmatrix} \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_1} \right)^2 & \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_1} \right) \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_2} \right) \\ \text{SYM} & \left(\sum_{m>0} \frac{f_m^T}{\lambda_m^T} (w_m^T)_{x_2} \right)^2 \end{pmatrix}$$

is singular: $\det(M_T^0(x)) = 0$ so that $\arg(\max_{K^0 \in G_{\theta T}} K^0 : M_T^0) = (\lambda_{\theta T}^-, \lambda_{\theta T}^+)$

For any $T \geq \sup_{x \in \Omega} T^+(x)$, the solution of the problem $\max_{K^0 \in G_{\theta T}} K^0 : M_T$ is $(\lambda_{\theta T}^-, \lambda_{\theta T}^+)$ so that the optimal tensor K_T^* is a first order laminate.

Structure of the matrix $M_T - T$ large

$$c_{mn}^T = -\frac{f_m^T}{\lambda_m^T} \frac{f_n^T}{\lambda_n^T} + \frac{1}{T} \frac{f_n^T}{(\lambda_m^T)^2 (\lambda_n^T)^2} (\lambda_n^T f_m^T - \lambda_m^T a_m^T \lambda_n^T) + O(e^{-\lambda_m T}, e^{-\lambda_n T})$$

where the coefficients $c_{mn}^{i,T}$ are bounded with respect to T . Notice that only the coefficients of the heat source f are involved in the first order terms $c_{mn}^{0,T}$. We put

$$M_T(x) = M_T^0(x) + \frac{M_T^1(x)}{T} + M^2(T, x), \quad x \in \Omega.$$

Now, we observe that the symmetric matrix $M_T^0(x)$, which is given by

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is singular: $\det(M_T^0(x)) = 0$ so that $\arg(\max_{K^0 \in G_{\theta_T}} K^0 : M_T^0) = (\lambda_{\theta_T}^-, \lambda_{\theta_T}^+)$

Proposition (T large)

For any $T \geq \sup_{x \in \Omega} T^+(x)$, the solution of the problem $\max_{K^0 \in G_{\theta_T}} K^0 : M_T$ is $(\lambda_{\theta_T}^-, \lambda_{\theta_T}^+)$ so that the optimal tensor K_T^* is a first order laminate.

The analysis for T arbitrarily small is similar. Precisely, we obtain

$$c_{mn}^T = a_m^T a_n^T - \frac{1}{2} a_m^T a_n^T (\lambda_m^T + 3\lambda_n^T) T + \frac{1}{2} (a_m^T f_n^T + a_n^T f_m^T) T + O(T^2), \quad m, n > 0 \quad (21)$$

so that we decompose the matrix M_T as $M_T(x) = M_T^0(x) + TM_T^1(x) + M_T^2(T, x)$ in Ω , where the symmetric matrix M_T^0 depends only on the coefficients a_m^T of the initial condition u_0 , assumed different from zero. By arguing as before, we obtain that $M_T^0(x)$ is singular so that for T small enough, says $T \leq T^-(x)$, the determinant of $M_T(x)$ is negatif.

Proposition (T small)

For any $T \leq \inf_{x \in \Omega} T^-(x)$, the solution of the problem $\max_{K^0 \in G_{\theta_T}} K^0 : M_T$ is $(\lambda_{\theta_T}^-, \lambda_{\theta_T}^+)$ so that the micro-structure is recovered by first order laminates.

If $u_0 = 0$, $M_T^0 = M_T^1 = 0$ and $M_T(x) = T^2 M_T^2 + \dots$ with $\det(M_T^2(x)) \leq 0$ for all $x \in \Omega$
 $T > 0$.

Algorithm of minimization of $\overline{J}_T^*(\theta_T, K_T^*)$

⇒ Gradient method for the variable θ_T and optimality condition for the variable K_T^* :

Given $T > 0$, $L \in (0, 1)$, $\Omega \subset \mathbb{R}^2$, $u_0, f \in L^2(\Omega)$ and $0 < \varepsilon \ll 1$,

• Initialization: $\theta_T^0 = L$, $(v_1^0, v_2^0, \dots, v_N^0) \in \partial G_\theta$, $P^0 = Id_N$, Λ_T^0 the diagonal matrix such that $(\Lambda_T^0)_{ii} = v_i^0$ and finally put $K_T^{*,0} = P^0 \Lambda_T^0 (P^0)^t$.

• For $n \geq 1$, iteration until convergence

$$(|J_T^*(\theta_T^n, K_T^{*,n-1}) - J_T^*(\theta_T^n, K_T^{*,n})| \leq \varepsilon |J_T^*(\theta_T^0, K_T^{*,0})|) :$$

1. Compute the exact θ_T^n and the adjoint state ψ^n .

2. Compute the descent direction $\partial \theta(\theta_T^n, \psi^n)$ given by

$$\partial \theta(\theta_T^n, \psi^n) = \int_{\Omega} \lambda_T^n \psi^n \nabla \cdot (\nabla \theta_T^n - \nabla \psi^n) dx - \int_{\Omega} \lambda_T^n \psi^n \nabla \cdot \nabla \theta_T^n dx$$

3. Then take the search direction $\theta_T^{n+1} = \theta_T^n - \alpha \partial \theta(\theta_T^n, \psi^n)$.

4. Compute the exact $K_T^{*,n}$ and the adjoint state ψ^n for θ_T^{n+1} .

5. Compute the exact θ_T^{n+1} and the adjoint state ψ^n for $K_T^{*,n}$.

6. Compute the matrix $\Lambda_T^{n+1} = \text{diag}(v_1^{n+1}, \dots, v_N^{n+1})$ by diagonalizing

$$K_T^{*,n} - \alpha \partial K_T(K_T^{*,n}, \theta_T^{n+1}, \psi^n)$$

with $v_i^{n+1} = \max(0, \lambda_i)$ and finally put $P^{n+1} = \Lambda_T^{n+1} (P^{n+1})^t$.

Algorithm of minimization of $\bar{J}_T^*(\theta_T, K_T^*)$

⇒ Gradient method for the variable θ_T and optimality condition for the variable K_T^* :

Given $T > 0$, $L \in (0, 1)$, $\Omega \subset \mathbb{R}^2$, $u_0, f \in L^2(\Omega)$ and $0 < \varepsilon \ll 1$,

- Initialization: $\theta_T^0 = L$, $(v_1^0, v_2^0, \dots, v_N^0) \in \partial G_\theta$, $P^0 = Id_N$, Λ_T^0 the diagonal matrix such that $(\Lambda_T^0)_{ii} = v_i^0$ and finally put $K_T^{*,0} = P^0 \Lambda_T^0 (P^0)^t$.

- For $n \geq 1$, iteration until convergence

$$(|\bar{J}_T^*(\theta_T^n, K^{*,n-1}) - \bar{J}_T^*(\theta_T^n, K^{*,n})| \leq \varepsilon |\bar{J}_T^*(\theta_T^0, K^{*,0})|) :$$

1. Compute the state u^n and the adjoint state p^n .
2. Compute the descent direction $\delta\theta(u^n, p^n)$ given by

$$\delta\theta = \frac{1}{T} \int_0^T K_{T,\theta}^n \nabla u \cdot (\nabla u + 2\nabla p) dt - 2(\beta_2 - \beta_1) \frac{1}{T} \int_0^T u' p dt + l \quad \text{in } \Omega,$$

and then update the density $\theta_T^n := \theta_T^{n-1} + \eta \delta\theta(u^n, p^n)$, where $\eta \in L^\infty(\Omega, \mathbb{R}_+^*)$ denotes a function, which depends on the multiplier l and chosen so that θ^n satisfies the volume constraint.

3. Compute the matrix $M_T^n = M_T^n(u^n, p^n)$ in Ω , its eigenvalues $\mu_1^n, \mu_2^n, \dots, \mu_N^n$ and corresponding eigenvectors $e_1^n, e_2^n, \dots, e_N^n$, and set $P^n = (e_1^n, e_2^n, \dots, e_N^n)$.
4. Solve the linear problem $\max_{K^0 \in G_{\theta_T^n}} K^0 : M_T^n$ leading to $(v_1^n, v_2^n, \dots, v_N^n) \in \partial G_{\theta_T^n}$. Consider the matrix $\Lambda_T^n = (v_1^n, v_2^n, \dots, v_N^n)$ and then put $K_T^{*,n} = P^n \Lambda_T^n (P^n)^t$.

Algorithm of minimization of $\bar{J}_T^*(\theta_T, K_T^*)$

⇒ Gradient method for the variable θ_T and optimality condition for the variable K_T^* :

Given $T > 0$, $L \in (0, 1)$, $\Omega \subset \mathbb{R}^2$, $u_0, f \in L^2(\Omega)$ and $0 < \varepsilon \ll 1$,

- Initialization: $\theta_T^0 = L$, $(v_1^0, v_2^0, \dots, v_N^0) \in \partial G_\theta$, $P^0 = Id_N$, Λ_T^0 the diagonal matrix such that $(\Lambda_T^0)_{ii} = v_i^0$ and finally put $K_T^{*,0} = P^0 \Lambda_T^0 (P^0)^t$.

- For $n \geq 1$, iteration until convergence

$$(|\bar{J}_T^*(\theta_T^n, K^{*,n-1}) - \bar{J}_T^*(\theta_T^n, K^{*,n})| \leq \varepsilon |\bar{J}_T^*(\theta_T^0, K^{*,0})|) :$$

1. Compute the state u^n and the adjoint state p^n .
2. Compute the descent direction $\delta\theta(u^n, p^n)$ given by

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Example 1: Uniform heat source $f = 1$ and $u_0 = 0$

$$N = 2, \quad \Omega = (0, 1)^2, \quad (\beta_1, k_1) = (1, 0.07), \quad (\beta_2, k_2) = (1, 0.14), \quad L = 0.5, \quad f = 1, \quad u_0 = 0$$

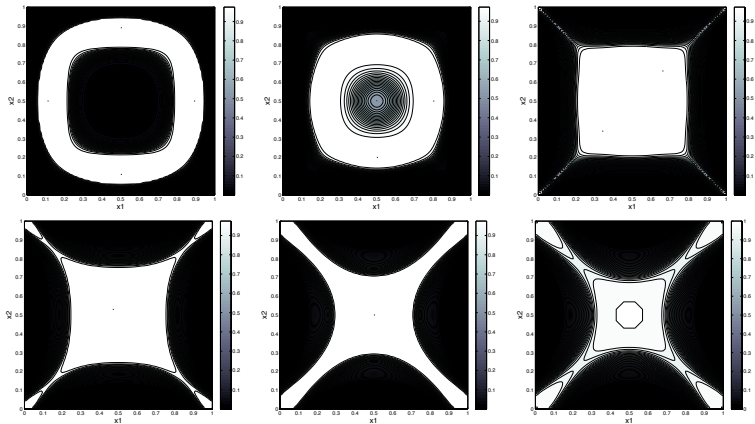


Figure: Isovalues of θ_T in Ω for $T = 0.5$ (Top left), $T = 1$, $T = 1.5$, $T = 2$, $T = 4$ and the limit elliptic case " $T = \infty$ " (Bottom right). The white zones correspond to the weaker conductor phase (β_1, k_1) .

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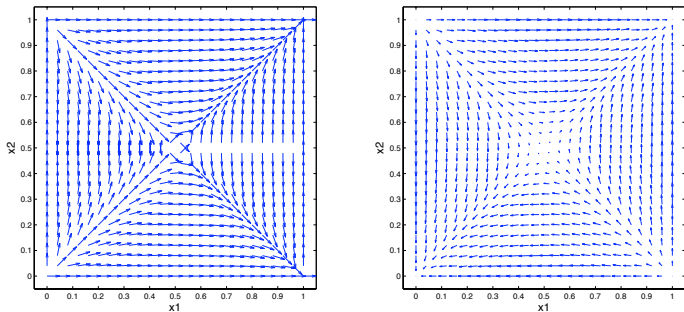


Figure: Direction of lamination for $T = 2$ (**Left**) and $T = \infty$ (**Right**).

\implies We observe first order laminates for all $x \in \Omega$ and all $T > 0$.

Exemple 2: Non uniform heat source and $u_0 = 0$

$$f(x) = \mathcal{X}_{(0.05,0.15) \times (0.1,0.9)}(x) - \mathcal{X}_{(0.85,0.95) \times (0.1,0.9)}(x)$$

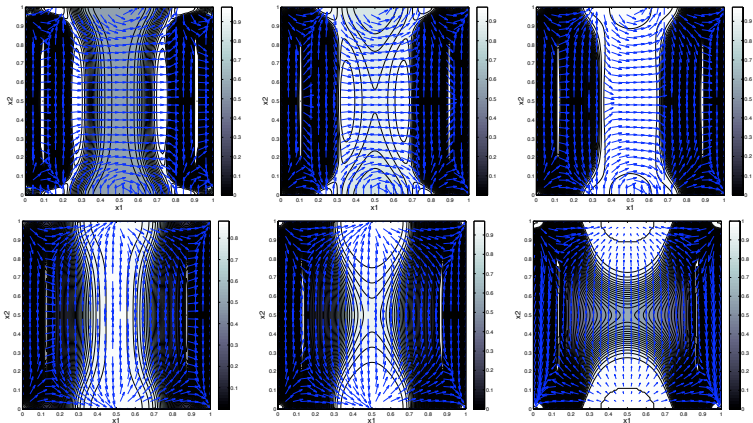


Figure: Isovalues of θ_T and direction of lamination in Ω for $T = 0.25$ (Top left), $T = 0.5$, $T = 1$, $T = 2$, $T = 4$ and the limit elliptic case " $T = \infty$ " (Bottom right).

Example 3: Interplay between f and u_0

$$u_0(x) = \frac{1}{4} \chi_{(0.2,0.8) \times (0.1,0.2)}(x) - \frac{1}{4} \chi_{(0.2,0.8) \times (0.8,0.9)}(x)$$

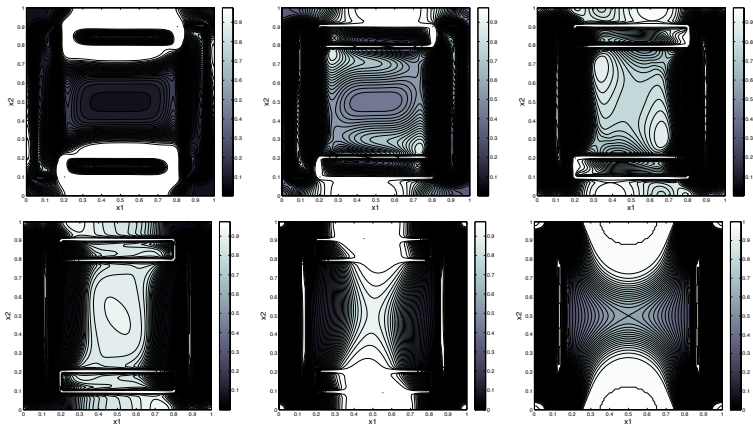


Figure: Isovals of θ_T in Ω for $T = 0.125, 0.25, 0.5, T = 1, T = 4$ and " $T = \infty$ ".

→ We observe second order laminates for $T \in [0, 125]$

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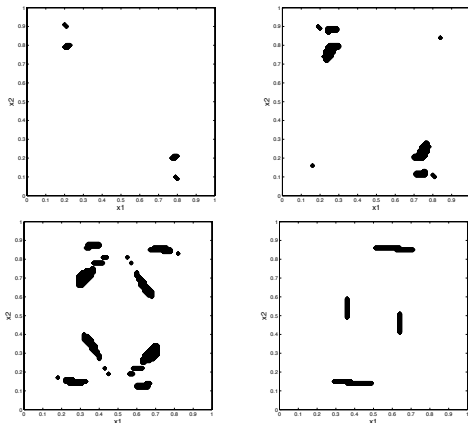


Figure: Second order laminate zone in Ω for $T = 0.125, 0.25, 0.5$ and $T = 1$.

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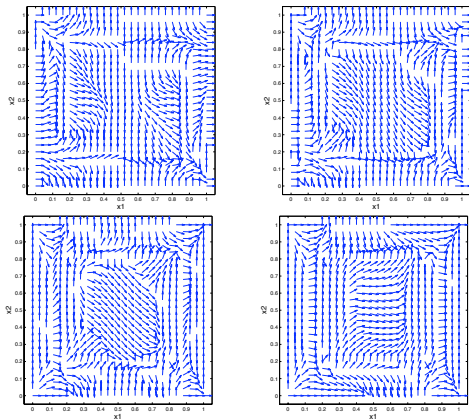


Figure: First eigenvector of the matrix M_T for $T = 0.125, 0.25, 0.5$ and $T = 1$.

- The minimizers K_T^* H-converge toward the minimizers of K_∞^*
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The time dependent case $\mathcal{X} = \mathcal{X}(t, x)$ - Variational approach and Young measure

Theorem (AM-Pedregal-Periago, JMPA 2008)

Assume that the solution of the heat system has the regularity $u \in L^2(0, T; H^2(\Omega))$ and depends continuously on the initial datum in the corresponding norms. Then the variational problem

$$(RP_t) \text{ Minimize in } (\theta, \bar{G}, u) : \bar{J}_t(\theta, \bar{G}, u) = \frac{1}{2} \int_0^T \int_{\Omega} \left[k_1 \frac{|\bar{G} - k_2 \nabla u|^2}{\theta (k_1 - k_2)^2} + k_2 \frac{|\bar{G} - k_1 \nabla u|^2}{(1 - \theta) (k_2 - k_1)^2} \right] dx dt$$

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is a relaxation of (VP_t) in the sense that

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More details on www.math.univ-bpclermont.fr/~munch/

THANK YOU FOR YOUR ATTENTION