

Explicit solutions of multiple state optimal design problems



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Joint work with **Marko Vrdoljak**



[INT. WORKSHOP ON PDES: ANALYSIS AND MODELLING]

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Outline

1 Energy minimization and relaxation

Posing the problem

Relaxation

2 Convex minimization problem

Simpler problem

Spherically symmetric case

3 Examples

One state

Multiple states



Multiple state optimal design problem

$\Omega \subseteq \mathbf{R}^d$ open and bounded, $f_1, \dots, f_m \in L^2(\Omega)$ given; stationary diffusion equations with homogenous Dirichlet b. c.:

$$\begin{cases} -\operatorname{div}(\mathbf{A}\nabla u_i) = f_i & , \quad i = 1, \dots, m \\ u_i \in H_0^1(\Omega) \end{cases} \quad (1)$$

where \mathbf{A} is a mixture of two isotropic materials with conductivities $0 < \alpha < \beta$: $\mathbf{A} = \chi\alpha\mathbf{I} + (1 - \chi)\beta\mathbf{I}$, where $\chi \in L^\infty(\Omega; \{0, 1\})$, $\int_\Omega \chi \, d\mathbf{x} = q_\alpha$, for given $0 < q_\alpha < |\Omega|$.

For given Ω , α , β , q_α , f_i , and some given weights $\mu_i > 0$, we want to find such material \mathbf{A} which minimizes the weighted sum of energies (total amounts of heat/electrical energy dissipated in Ω):

$$I(\chi) := \sum_{i=1}^m \mu_i \int_\Omega f_i u_i \, d\mathbf{x} \rightarrow \min, \quad \chi \in L^\infty(\Omega; \{0, 1\})$$



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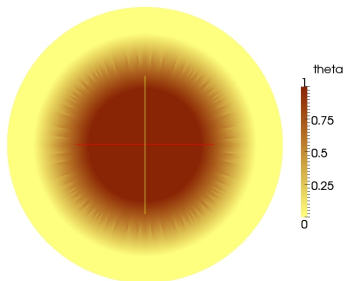
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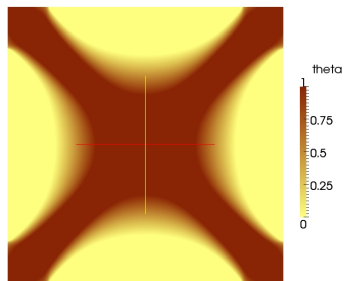


single state, $f \equiv 1$, Ω circle / square

Murat & Tartar



Lurie & Cherkvaev



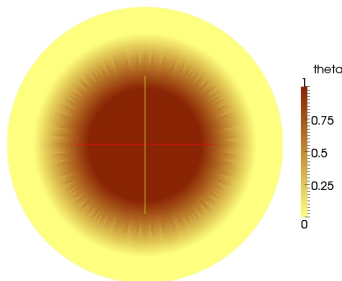
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 composite material - relaxation

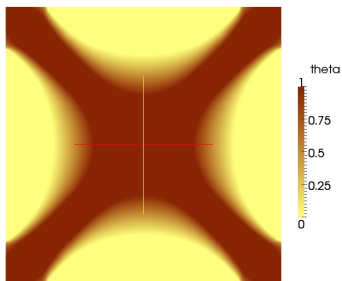


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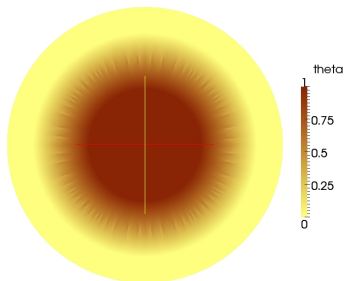
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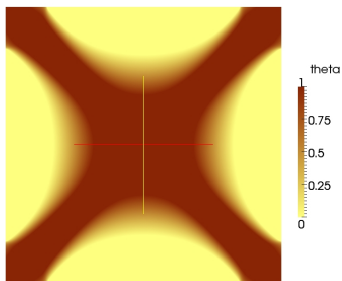


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Definition

If a sequence of characteristic functions $\chi_\varepsilon \in L^\infty(\Omega; \{0, 1\})$ and conductivities $\mathbf{A}^\varepsilon(x) = \chi_\varepsilon(x)\alpha\mathbf{I} + (1 - \chi_\varepsilon(x))\beta\mathbf{I}$ satisfy $\chi_\varepsilon \rightharpoonup \theta$ weakly $*$ and \mathbf{A}^ε H -converges to \mathbf{A}^* , then it is said that \mathbf{A}^* is homogenised tensor of two-phase composite material with proportions θ of first material and microstructure defined by the sequence (χ_ε) .

Example – simple laminates: if χ_ε depend only on x_1 , then

$$\mathbf{A}^* = \text{diag}(\lambda_\theta^-, \lambda_\theta^+, \lambda_\theta^+, \dots, \lambda_\theta^+),$$

where

$$\lambda_\theta^+ = \theta\alpha + (1 - \theta)\beta, \quad \frac{1}{\lambda_\theta^-} = \frac{\theta}{\alpha} + \frac{1 - \theta}{\beta}.$$

Set of all composites:

$$\mathcal{A} := \{(\theta, \mathbf{A}) \in L^\infty(\Omega; [0, 1] \times \mathbb{M}_d(\mathbf{R})) : \int_\Omega \theta \, dx = q_\alpha, \mathbf{A} \in \mathcal{K}(\theta) \text{ a.e.}\}$$



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Effective conductivities – set $\mathcal{K}(\theta)$

G-closure problem: for given θ find all possible homogenised (effective) tensors \mathbf{A}^*

2D:

$\mathcal{K}(\theta)$ is given in terms of eigenvalues (Murat & Tartar; Lurie & Cherkhaev):

$$\lambda_{\theta}^{-} \leq \lambda_j \leq \lambda_{\theta}^{+} \quad j = 1, \dots, d$$

$$\sum_{j=1}^d \frac{1}{\lambda_j - \alpha} \leq \frac{1}{\lambda_{\theta}^{-} - \alpha} + \frac{d-1}{\lambda_{\theta}^{+} - \alpha} \quad 3D:$$

$$\sum_{j=1}^d \frac{1}{\beta - \lambda_j} \leq \frac{1}{\beta - \lambda_{\theta}^{-}} + \frac{d-1}{\beta - \lambda_{\theta}^{+}},$$

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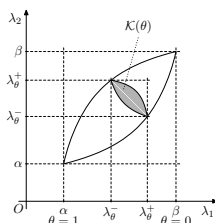
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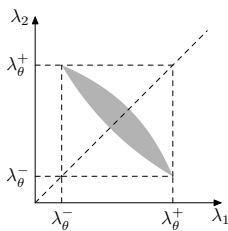
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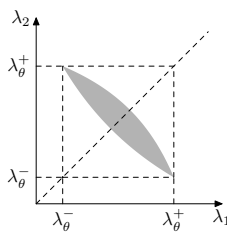
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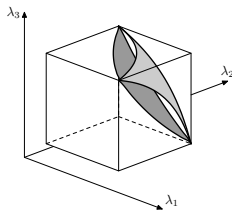
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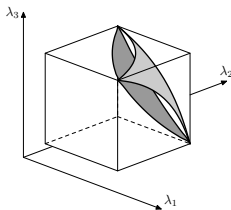
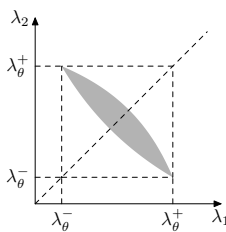
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How do we find a solution?

Goal: find explicit solution for some simple domains (circle)

Motivation: test examples for robust numerical algorithms

A. Single state equation: [Murat & Tartar]

This problem can be rewritten as a simpler convex minimization problem.

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Motivation: test examples for robust numerical algorithms

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$$\min_{\mathcal{T}} I \iff \min_{\mathcal{A}} J \text{ if } m < d$$

Theorem

If $m < d$ then $\min_{\mathcal{A}} J = \min_{\mathcal{T}} I$ and:

- *There is unique $u^* \in H_0^1(\Omega; \mathbf{R}^m)$ which is the state for every solution of $\min_{\mathcal{A}} J$ and $\min_{\mathcal{T}} I$.*
- *If (θ^*, \mathbf{A}^*) is an optimal design for the problem $\min_{\mathcal{A}} J$, then θ^* is optimal design for $\min_{\mathcal{T}} I$.*
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Spherical symmetry: $\min_{\mathcal{A}} J \iff \min_{\mathcal{T}} I$

Theorem

Let $\Omega \subseteq \mathbf{R}^d$ be spherically symmetric, and let the right-hand sides $f_i = f_i(r)$, $r \in \omega$, $i = 1, \dots, m$ be radial functions. Then $\min_{\mathcal{A}} J = \min_{\mathcal{T}} I$ and there is unique (radial) u^* which is the state for any solution of $\min_{\mathcal{A}} J$ and $\min_{\mathcal{T}} I$. Moreover,

- a) For any minimizer θ of functional I over \mathcal{T} , let us define a radial function $\theta^* : \Omega \rightarrow \mathbf{R}$ as the average value over spheres of θ : for $r \in \omega$ we take

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Spherical symmetry... cont.

Theorem

- b) For any radial minimizer θ^* of I over \mathcal{T} , let us define $\mathbf{A}^* \in \mathcal{K}(\theta^*)$ as a simple laminate with the lamination direction orthogonal to the radial vector \mathbf{e}_r , almost everywhere on Ω . To be specific, we define

$$\mathbf{A}^*(\mathbf{x}) := \text{diag}(\lambda_{\theta^*}^+(|\mathbf{x}|), \lambda_{\theta^*}^-(|\mathbf{x}|), \lambda_{\theta^*}^+(|\mathbf{x}|), \dots, \lambda_{\theta^*}^+(|\mathbf{x}|)) .$$

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Then (θ^*, \mathbf{A}^*) is an optimal design for $\min_{\mathcal{A}} J$.

- c) If $(\theta^*, \mathbf{A}^*) \in \mathcal{A}$ is a solution of the relaxed problem $\min_{\mathcal{A}} J$ then θ^* is optimal for $\min_{\mathcal{T}} I$, and $\mathbf{A}^* \nabla u_i^* = \lambda_{\theta^*}^+ \nabla u_i^*$, almost everywhere, $i = 1, \dots, m$.



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Uniqueness on a ball

Lemma

Let Ω be ball $B(\mathbf{0}, R)$, and let the right-hand sides f_i be radial functions, such that mappings $r \mapsto r^{\frac{d-1}{2}} f_i(r)$ belong to $L^2(\langle 0, R \rangle)$, $i = 1, \dots, m$. Then there are unique radial fluxes

$$\sigma_i^*(r) = -\frac{1}{r^{d-1}} \int_0^r \rho^{d-1} f_i(\rho) d\rho \mathbf{e}_r$$

corresponding to each minimizer of $\min_{\mathcal{T}} I$, and this minimizer is radial and unique on the set where at least one σ_i^* does not vanish. If the Lagrange multiplier c is positive, this holds true on the whole $B(\mathbf{0}, R)$.



Optimality conditions for $\min_{\mathcal{T}} I$

Lemma

$\theta^* \in \mathcal{T}$ is a solution $\min_{\mathcal{T}} I$ if and only if there exists a Lagrange multiplier $c \geq 0$ such that

$$\begin{aligned}\theta^* \in \langle 0, 1 \rangle &\Rightarrow \sum_{i=1}^m \mu_i |\nabla u_i^*|^2 = c, \\ \theta^* = 0 &\Rightarrow \sum_{i=1}^m \mu_i |\nabla u_i^*|^2 \geq c, \\ \theta^* = 1 &\Rightarrow \sum_{i=1}^m \mu_i |\nabla u_i^*|^2 \leq c,\end{aligned}$$

or equivalently

$$\begin{aligned}\sum_{i=1}^m \mu_i |\nabla u_i^*|^2 > c &\Rightarrow \theta^* = 0, \\ \sum_{i=1}^m \mu_i |\nabla u_i^*|^2 < c &\Rightarrow \theta^* = 1.\end{aligned}$$



Ball $\Omega = B(0, 2) \subseteq \mathbb{R}^2$ with nonconstant right-hand side

In all examples $\alpha = 1, \beta = 2$, one state equation, $f(r) = 1 - r$

State equation in polar coordinates $-\frac{1}{r} \left(r \lambda_{\theta(r)}^+ u' \right)' = 1 - r.$

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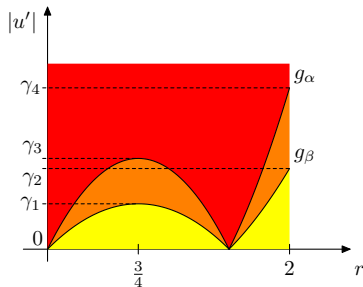
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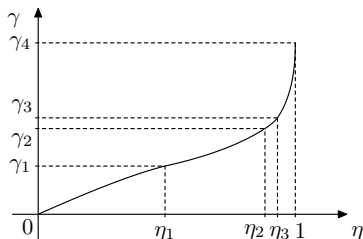
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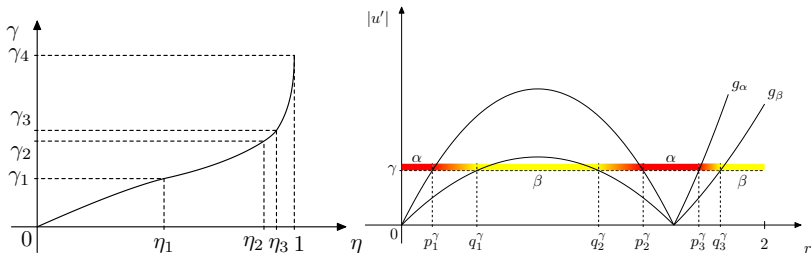
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$$u_i'(r) = \frac{\psi_i(r)}{\theta(r)\alpha + (1 - \theta(r))\beta}, \quad i = 1, 2,$$

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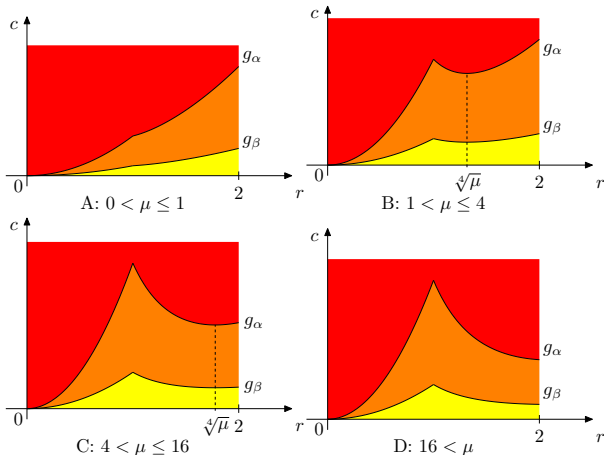
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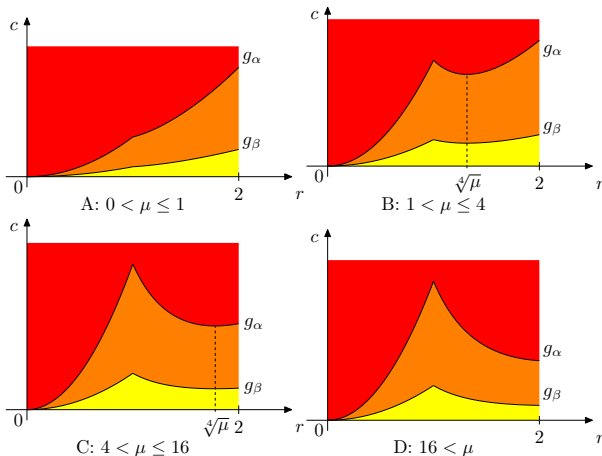
Geometric interpretation of optimality conditions



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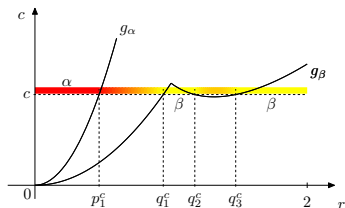
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Optimal θ^* for case B

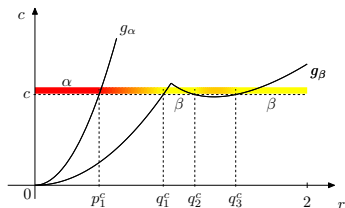


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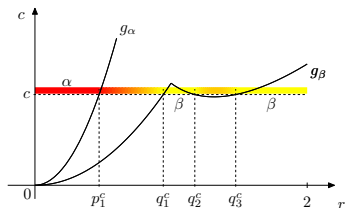


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