

## Explicit solutions of multiple state optimal design problems

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Joint work with Marko Vrdoljak





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Posing the problem Relaxation



#### Multiple state optimal design problem

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

$$\begin{cases} -\operatorname{div}\left(\mathbf{A}\nabla u_{i}\right) = f_{i}\\ u_{i} \in \mathrm{H}_{0}^{1}(\Omega) \end{cases}, \quad i = 1, \dots, m \tag{1}$$

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  $\Omega, \alpha, \beta, q_{\alpha}, f_i$ , 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  $\Omega$ ):

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

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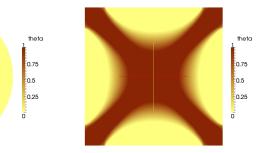
Energy minimization and relaxation problem Convex minimization Examples

Posing the problem



#### single state, $f \equiv 1, \Omega$ circle / square

#### Murat & Tartar



Lurie & Cherkaev



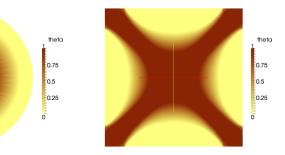
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 $\chi \in \mathrm{L}^\infty(\Omega; \{0, 1\}) \quad \cdots \quad \theta \in \mathrm{L}^\infty(\Omega; [0, 1])$ 

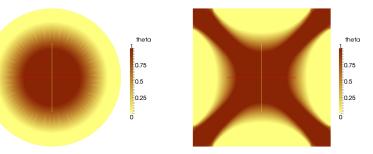
 $\mathbf{A} = \chi \alpha \mathbf{I} + (1 - \chi) \beta \mathbf{I}$   $\mathbf{A} \in \mathcal{K}(\theta)$  a.e. on  $\Omega$ classical material composite mateiral - relaxation

Posing the problem Relaxation



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$$\begin{split} \boldsymbol{\chi} &\in \mathrm{L}^\infty(\Omega; \{0,1\})\\ \mathbf{A} &= \boldsymbol{\chi} \boldsymbol{\alpha} \mathbf{I} + (1-\boldsymbol{\chi}) \boldsymbol{\beta} \mathbf{I}\\ \text{classical material} \end{split}$$

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Posing the problem Relaxation



### Composite material 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  $\mathbf{A}^{\varepsilon}$  *H*-converges to  $\mathbf{A}^{*}$ , then it is said that  $\mathbf{A}^{*}$  is homogenised tensor of two-phase composite material with proportions  $\theta$  of first material and microstructure defined by the sequence ( $\chi_{\varepsilon}$ ).

Example – simple laminates: if  $\chi_{\varepsilon}$  depend only on  $x_1$ , then

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

where

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

Set of all composites:

$$\mathcal{A} := \{( heta, \mathbf{A}) \in \mathrm{L}^{\infty}(\Omega; [0, 1] imes \mathrm{M}_{d}(\mathbf{R})) : \int_{\Omega} heta \, d\mathbf{x} = q_{lpha} \,, \; \mathbf{A} \in \mathcal{K}(m{ heta}) \; \mathsf{a.e.} \,\}$$

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Posing the problem Relaxation



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

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

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

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

$$\sum_{j=1}^{d} \frac{1}{\lambda_{j} - \alpha} \leq \frac{1}{\lambda_{\theta}^{-} - \alpha} + \frac{d - 1}{\lambda_{\theta}^{+} - \alpha} \stackrel{\text{3D}}{}$$

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 $\min_{\mathcal{A}} J$  is a proper relaxation of  $\min_{\mathrm{L}^\infty(\Omega;\{0,1\})} I$ 

Posing the Relaxation

2D:

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

problem

G-closure problem: for given  $\theta$  find all possible homogenised (effective) tensors  $A^*$ 

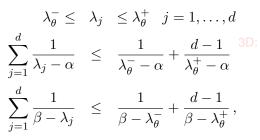
Energy minimization and relaxation

minimization

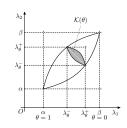
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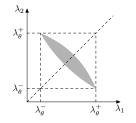
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Energy minimization and relaxation problem

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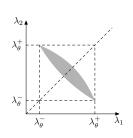
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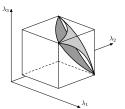
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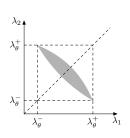
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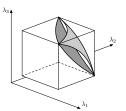
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Simpler problem Spherically symmetric case



#### 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.

$$\begin{split} I(\theta) &= \int_{\Omega} f u \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \\ \theta \in \mathcal{T} \text{, and } u \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^{+} \nabla u\right) = f \\ u \in \mathcal{H}_{0}^{1}(\Omega) \end{array} \right. \end{split}$$

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B. Multiple state equations: Simpler relaxation fails; in spherically symmetric case or when m < d, it can be done!

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 $\min_{\mathcal{A}} J \quad \Longleftrightarrow \quad \min_{\mathcal{T}} I$ 

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Krešimir Burazin

 $\min_{A} J$ 

Simpler problem Spherically symmetric case



#### How do we find a solution?

Goal: find explicit solution for some simple domains (circle) Motivation: test examples for robust numerical algorithms

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 $\min_{A} J$ 

Simpler problem Spherically symmetric case



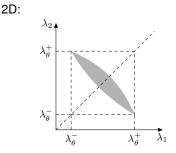
 $\min_{\mathcal{B}} J$ 

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

Further relaxation:

$$\mathcal{B} \qquad \dots \quad \int_{\Omega} \theta \, d\mathbf{x} = q_{\alpha}$$
$$\lambda_{\theta}^{-} \leq \lambda_{\min}(\mathbf{A}) \,, \, \lambda_{\max}(\mathbf{A}) \leq \lambda_{\theta}^{+}$$

 $\mathcal B$  is convex and compact and J is continuous on  $\mathcal B$ , so there is a solution of  $\min_{\mathcal B} J$ .



Simpler problem Spherically symmetric case



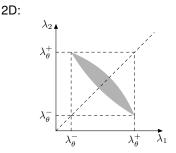
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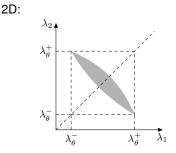


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Simpler problem Spherically symmetric case



#### $\min_{\mathcal{B}} J \Longleftrightarrow \min_{\mathcal{T}} I \Longleftrightarrow \min_{\mathcal{A}} J \text{ if } m < d$

- There is unique  $u^* \in H^1_0(\Omega; \mathbb{R}^m)$  which is the state for every solution of  $\min_{\mathcal{B}} J$  and  $\min_{\mathcal{T}} I$ .
- If (θ\*, A\*) is an optimal design for the problem min<sub>B</sub> J, then θ\* is optimal design for min<sub>T</sub> I.
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- If aditionally m < d, then above is valid for min<sub>A</sub> J instead min<sub>B</sub> J and optimal design can be realized as a simple laminate.

Simpler problem Spherically symmetric case



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Simpler problem Spherically symmetric case



#### Spherical symmetry: $\min_{\mathcal{A}} J \iff \min_{\mathcal{B}} J \iff \min_{\mathcal{T}} I$

#### Theorem

Let  $\Omega \subseteq \mathbf{R}^d$  be spherically symmetric, and let the right-hand sides  $f_i = f_i(r), r \in \omega, i = 1, ..., m$  be radial functions. Then  $\min_{\mathcal{A}} J = \min_{\mathcal{B}} J = \min_{\mathcal{T}} I$ , and there exists a minimizer  $(\theta^*, \mathbf{A}^*)$  of the optimal design problem  $\min_{\mathcal{A}} J$  which is a radial function. More precisely,

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

$$\theta^*(r) := \int_{\partial B(\mathbf{0},r)} \theta \, dS \,,$$

where S denotes the surface measure on a sphere. Then  $\theta^*$  is also minimizer for I over  $\mathcal{T}$ .

Simpler problem Spherically symmetric case



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Simpler problem Spherically symmetric case



#### Spherical symmetry...cont.

#### Theorem

b) For any radial minimizer  $\theta^*$  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  $\Omega$ . To be specific, we define

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in spherical basis (e<sub>r</sub>(x), e<sub>φ1</sub>(x), e<sub>φ2</sub>(x), ..., e<sub>φd-1</sub>(x)). Then (θ\*, A\*) is an optimal design for min<sub>B</sub> J. Moreover, (θ\*, A\*) ∈ A, and thus it is also a solution for min<sub>A</sub> J.
c) If (θ, A) ∈ A is a solution of the relaxed problem min<sub>A</sub> J with corresponding state function ũ, then θ is optimal for min<sub>T</sub> I, an (θ, A) is also a minimizer of J on B. Consequently, we have ũ = u\* and A = = λ (θ) = almost everywhere

Simpler problem Spherically symmetric case



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Simpler problem Spherically symmetric case



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

 $\theta^{i}$ 

#### 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

$$\begin{array}{ll} {}^{*} \in \langle 0,1\rangle & \Rightarrow & \displaystyle \sum_{i=1}^{m} \mu_{i} |\nabla u_{i}^{*}|^{2} = c \,, \\ \\ \theta^{*} = 0 & \Rightarrow & \displaystyle \sum_{i=1}^{m} \mu_{i} |\nabla u_{i}^{*}|^{2} \geq c \,, \\ \\ \theta^{*} = 1 & \Rightarrow & \displaystyle \sum_{i=1}^{m} \mu_{i} |\nabla u_{i}^{*}|^{2} \leq c \,, \end{array}$$

or equivalently

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

One state Multiple states



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

optimality conditions:  $\gamma := \sqrt{c} > 0$ 

 $\gamma$  is uniquely determined by

$$\int_{\Omega} \theta^* \, d\mathbf{x} = \eta := \frac{q_{\alpha}}{|\Omega|} \in [0, 1],$$

One state Multiple states

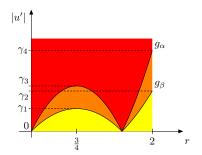


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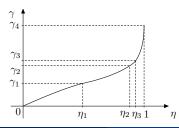


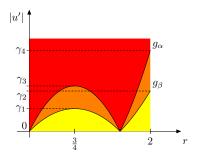
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One state Multiple states

|u'|

 $\gamma_4$ 

 $\gamma_3$ 

 $\gamma_2$  $\gamma_1$ .



 $g_{\alpha}$ 

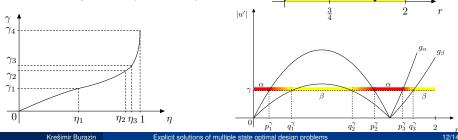
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One state Multiple states



# Two state equations on a ball $\Omega=B(\mathbf{0},2)$

• 
$$f_1 = \chi_{B(\mathbf{0},1)}, \ f_2 \equiv 1,$$

• 
$$\mu \int_{\Omega} f_1 u_1 \, d\mathbf{x} + \int_{\Omega} f_2 u_2 \, d\mathbf{x} \to \min$$

One state Multiple states



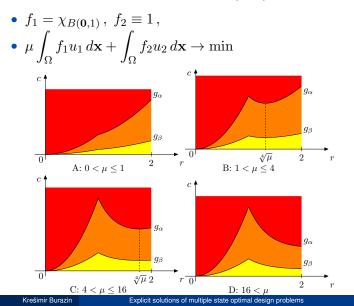
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# Two state equations on a ball $\Omega = B(\mathbf{0}, 2)$



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One state Multiple states



# Optimal $\theta^*$ for case B

As before, Lagrange multiplier can be numerically calculated from corresponding algebraic equation  $\int_{\Omega} \theta^* d\mathbf{x} = \eta$ .

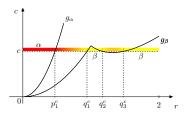
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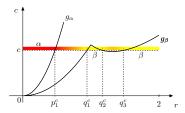
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