



Explicit solutions in optimal design problems for stationary diffusion equation

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Joint work with Marko Vrdoljak



Outline



Compliance optimization, composite materials and relaxation

Multiple states - spherically symmetric case

Examples



Optimal design problem (single state)

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For given Ω , α , β , q_α and f we want to find such material \mathbf{A} which minimizes the compliance functional (total amount of heat/electrical energy dissipated in Ω):

$$J(\chi) = \int_{\Omega} f(\mathbf{x})u(\mathbf{x})d\mathbf{x} = \int_{\Omega} \mathbf{A}(\mathbf{x})\nabla u(\mathbf{x}) \cdot \nabla u(\mathbf{x}) \, d\mathbf{x} \longrightarrow \min,$$

where u is the solution of the state equation (1).



Relaxation by homogenisation



$$\begin{array}{ll} \chi \in L^\infty(\Omega; \{0, 1\}) & \dots \quad \theta \in L^\infty(\Omega; [0, 1]) \\ \mathbf{A} = \chi\alpha\mathbf{I} + (1 - \chi)\beta\mathbf{I} & \mathbf{A} \in \mathcal{K}(\theta) \quad \text{a.e. on } \Omega \\ \text{classical material} & \text{composite material - relaxation} \end{array}$$





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Definition

A sequence of matrix functions \mathbf{A}^ε is said to *H-converge* to \mathbf{A}^* if for every f the sequence of solutions of

$$\begin{cases} -\operatorname{div}(\mathbf{A}^\varepsilon \nabla u_\varepsilon) = f \\ u_\varepsilon \in H_0^1(\Omega) \end{cases}$$

satisfies $u_\varepsilon \rightharpoonup u$ in $H_0^1(\Omega)$, $\mathbf{A}^\varepsilon \nabla u_\varepsilon \rightharpoonup \mathbf{A}^* \nabla u$ in $L^2(\Omega; \mathbf{R}^d)$, where u is the solution of the homogenised equation

$$\begin{cases} -\operatorname{div}(\mathbf{A}^* \nabla u) = f \\ u \in H_0^1(\Omega). \end{cases}$$





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}^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) .





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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

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



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Set of all composites:

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



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



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



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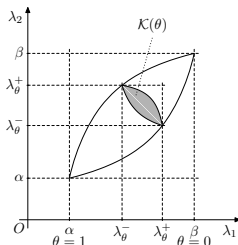
$\mathcal{K}(\theta)$ is given in terms of eigenvalues (Murat & Tartar; Lurie & Cherkhaev):

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

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

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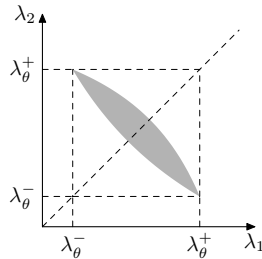
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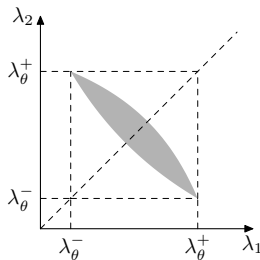
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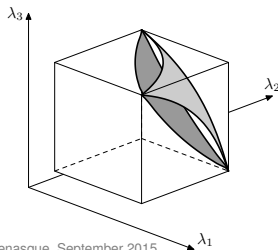
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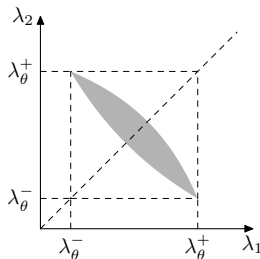
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$\min_{\mathcal{A}} J$ is a proper relaxation of

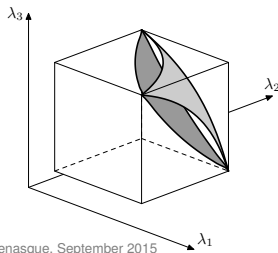
$$\min_{L^{\infty}(\Omega; \{0,1\})} J$$

Krešimir Burazin

2D:



3D:



Benasque, September 2015





Multiple state optimal design problem

State equations

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

State function $u = (u_1, \dots, u_m)$





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$$\begin{cases} I(\chi) = \sum_{i=1}^m \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \rightarrow \min \\ \mathbf{u} = (u_1, \dots, u_m) \text{ state function for } \mathbf{A} = \chi\alpha\mathbf{I} + (1 - \chi)\beta\mathbf{I} \\ \chi \in L^\infty(\Omega; \{0, 1\}), \int_{\Omega} \chi \, d\mathbf{x} = q_\alpha, \end{cases}$$

for some given weights $\mu_i > 0$.





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for some given weights $\mu_i > 0$. Proper relaxation:

$$J(\theta, \mathbf{A}) = \sum_{i=1}^m \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \rightarrow \min \quad \text{on}$$

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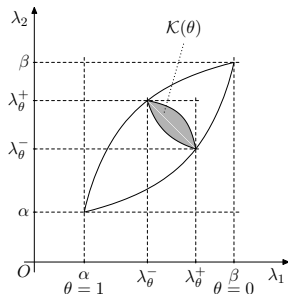
In spherically symmetric case the simpler relaxation can be done!



Relaxed designs



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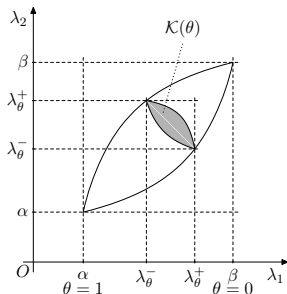


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Further relaxation:

$$\mathcal{B} \quad \dots \quad \int_{\Omega} \theta \, d\mathbf{x} = q_\alpha$$

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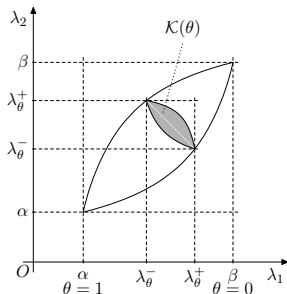
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\mathcal{B} is convex and compact and J is continuous on \mathcal{B} , so there is a solution of $\min_{\mathcal{B}} J$.



Equivalence of $\min_{\mathcal{B}} J$ and $\min_{\mathcal{T}} I$



Theorem

- ▶ *There is unique $u^* \in H_0^1(\Omega; \mathbf{R}^m)$ which is the state for every solution of $\min_{\mathcal{B}} J$ and $\min_{\mathcal{T}} I$.*





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- ▶ *Conversely, if θ^* is a solution of optimal design problem $\min_{\mathcal{T}} I$, then any $(\theta^*, \mathbf{A}^*) \in \mathcal{B}$ satisfying $\mathbf{A}^* \nabla u_i^* = \lambda_+(\theta^*) \nabla u_i^*$ almost everywhere on Ω (e.g. $\mathbf{A}^* = \lambda_+(\theta^*) \mathbf{I}$) is an optimal design for the problem $\min_{\mathcal{B}} J$.*





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- ▶ *If $m < d$, then there exists minimizer (θ^*, \mathbf{A}^*) for J on \mathcal{B} , such that $(\theta^*, \mathbf{A}^*) \in \mathcal{A}$, and thus it is also minimizer for J on \mathcal{A} .*



Simpler relaxation in case of spherical symmetry



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 a radial function. Then there exists a minimizer (θ^, \mathbf{A}^*) of the optimal design problem $\min_{\mathcal{A}} J$ which is a radial function.*



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- 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

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

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



Simpler relaxation in case of spherical symmetry... cont.



Theorem

- b) For any radial minimizer θ^* of I over \mathcal{T} , let us define \mathbf{A}^* as a simple laminate with layers orthogonal to a radial direction \mathbf{e}_r and local proportion of the first material θ^* . To be specific, we can define $\mathbf{A}^* : \Omega \rightarrow \mathbb{M}_d(\mathbf{R})$ in the following way:



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▶ If $\mathbf{x} = r\mathbf{e}_1 = (r, 0, 0, \dots, 0)$, then

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



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▶ For all other $\mathbf{x} \in \Omega$, we take the unique rotation $\mathbf{R}(\mathbf{x}) \in SO(d)$ such that $\mathbf{x} = |\mathbf{x}|\mathbf{R}(\mathbf{x})\mathbf{e}_1$, and define

$$\mathbf{A}^*(\mathbf{x}) := \mathbf{R}(\mathbf{x})\mathbf{A}^*(\mathbf{R}^T(\mathbf{x})\mathbf{x})\mathbf{R}^T(\mathbf{x}) .$$



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



Simpler relaxation in case of spherical symmetry... cont.



Theorem

b) For any radial minimizer θ^* of I over \mathcal{T} , let us define \mathbf{A}^* as a simple laminate with layers orthogonal to a radial direction \mathbf{e}_r and local proportion of the first material θ^* . To be specific, we can define $\mathbf{A}^* : \Omega \rightarrow \mathbb{M}_d(\mathbf{R})$ in the following way:

▶ If $\mathbf{x} = r\mathbf{e}_1 = (r, 0, 0, \dots, 0)$, then

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

▶ For all other $\mathbf{x} \in \Omega$, we take the unique rotation $\mathbf{R}(\mathbf{x}) \in SO(d)$ such that $\mathbf{x} = |\mathbf{x}|\mathbf{R}(\mathbf{x})\mathbf{e}_1$, and define

$$\mathbf{A}^*(\mathbf{x}) := \mathbf{R}(\mathbf{x})\mathbf{A}^*(\mathbf{R}^T(\mathbf{x})\mathbf{x})\mathbf{R}^T(\mathbf{x}) .$$

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

Moreover, $(\theta^*, \mathbf{A}^*) \in \mathcal{A}$, and thus it is also a solution for $\min_{\mathcal{A}} J$.





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 with nonconstant right-hand side



In all examples $\alpha = 1, \beta = 2$.

$\Omega = B(\mathbf{0}, 2) \subseteq \mathbf{R}^2$, one state equation, $f(r) = 1 - r$





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

Integration gives $|u'(r)| = \frac{\psi(r)}{\alpha\theta(r) + \beta(1-\theta(r))}$, where $\psi(r) = \frac{|2r^2 - 3r|}{6}$.





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Conditions of optimality: there exists a constant $\gamma := \sqrt{c} > 0$ such that for optimal θ^* we have:

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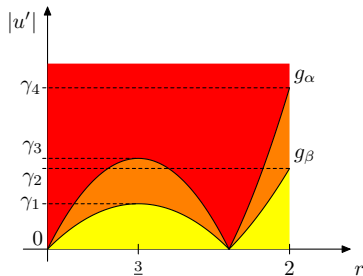
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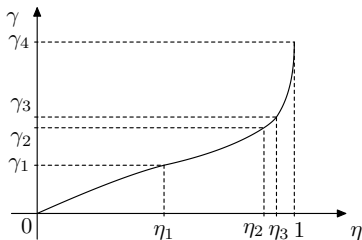
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Ball with nonconstant right-hand side

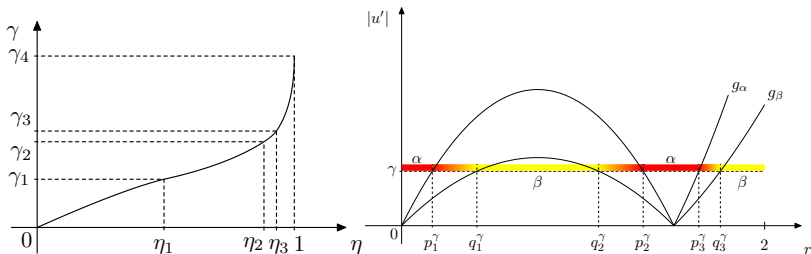
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Solving state equation

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

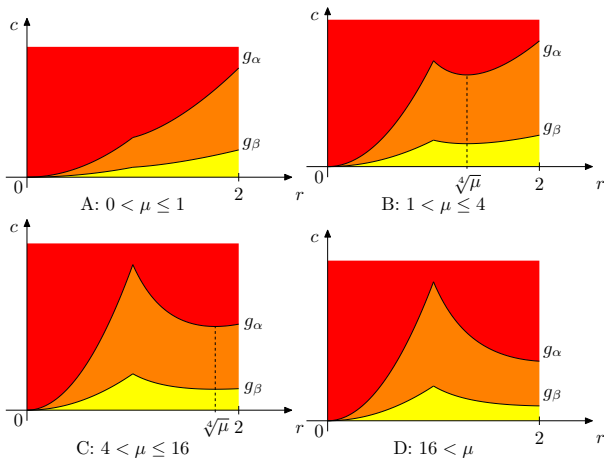
with

$$\psi_1(r) = \begin{cases} -\frac{r}{2}, & 0 \leq r < 1, \\ -\frac{1}{2r}, & 1 \leq r \leq 2, \end{cases} \quad \text{and} \quad \psi_2(r) = -\frac{r}{2}.$$

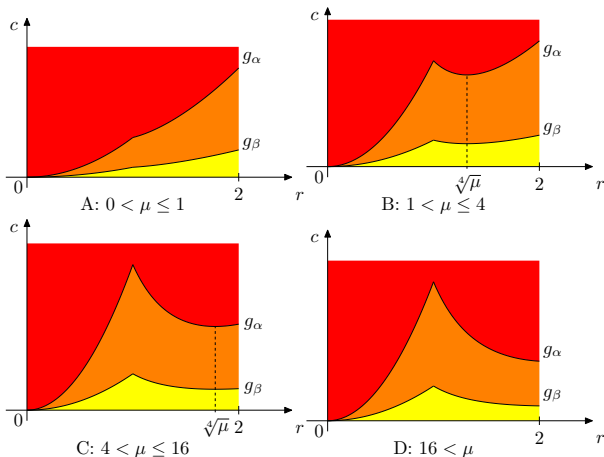
Similarly as in the first example: $\psi := \mu\psi_1^2 + \psi_2^2$, $\mathbf{g}_\alpha := \frac{\psi}{\alpha^2}$, $\mathbf{g}_\beta := \frac{\psi}{\beta^2}$.



Geometric interpretation of optimality conditions

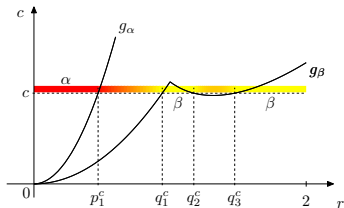


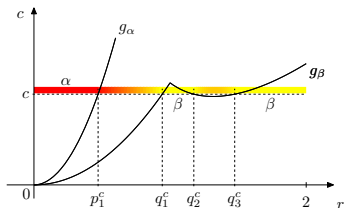
Geometric interpretation of optimality conditions



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



Optimal θ^* for case B

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In orange region:

$$\theta^*(r) = \frac{1}{\beta - \alpha} \left(\beta - \sqrt{\frac{\psi(r)}{c}} \right)$$

