



# 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  $\Omega$ ,  $\alpha$ ,  $\beta$ ,  $q_\alpha$  and  $f$  we want to find such material  $\mathbf{A}$  which minimizes the compliance functional (total amount of heat/electrical energy dissipated in  $\Omega$ ):

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

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  $\mathbf{A}^\varepsilon$  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  $\mathbf{A}^\varepsilon$   $H$ -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  $(\chi_\varepsilon)$ .





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Example – **simple laminates**: if  $\chi_\varepsilon$  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 $\mathcal{K}(\theta)$



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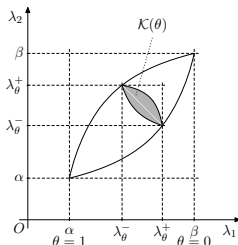
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$$\lambda_{\theta}^{-} \leq \lambda_j \leq \lambda_{\theta}^{+} \quad j = 1, \dots, d$$

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

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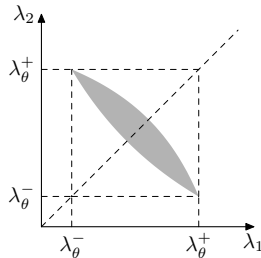
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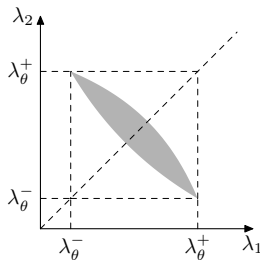
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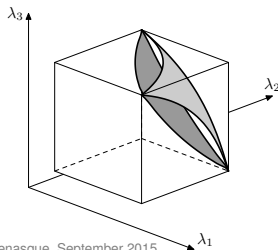
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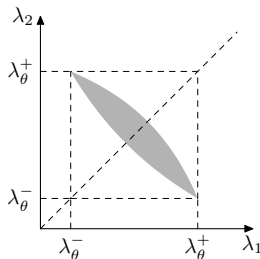
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$\min_{\mathcal{A}} J$  is a proper relaxation of

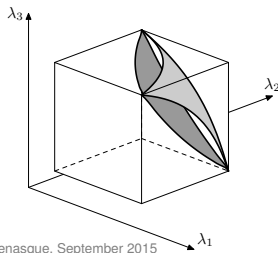
$$\min_{L^{\infty}(\Omega; \{0,1\})} J$$

Krešimir Burazin

2D:



3D:



Benasque, September 2015





## Multiple state optimal design problem

State equations

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

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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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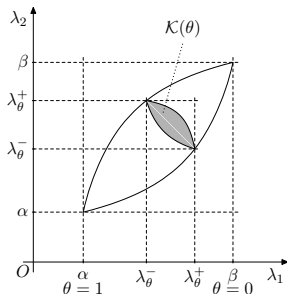
In spherically symmetric case the simpler relaxation can be done!



# Relaxed designs



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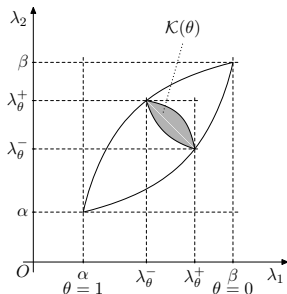


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Further relaxation:

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

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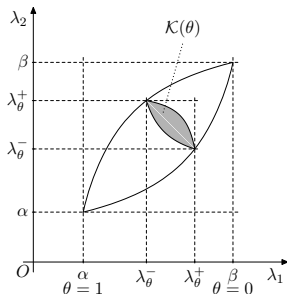
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$\mathcal{B}$  is convex and compact and  $J$  is continuous on  $\mathcal{B}$ , so there is a solution of  $\min_{\mathcal{B}} J$ .



# 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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- ▶ *If  $m < d$ , then there exists minimizer  $(\theta^*, \mathbf{A}^*)$  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  $(\theta^*, \mathbf{A}^*)$  of the optimal design problem  $\min_{\mathcal{A}} J$  which is a radial function.*



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- a) For any minimizer  $\theta$  of functional  $I$  over  $\mathcal{T}$ , let us define a radial function  $\theta^* : \Omega \rightarrow \mathbf{R}$  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 relaxation in case of spherical symmetry... cont.



## Theorem

- b) *For any radial minimizer  $\theta^*$  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  $\theta^*$ . 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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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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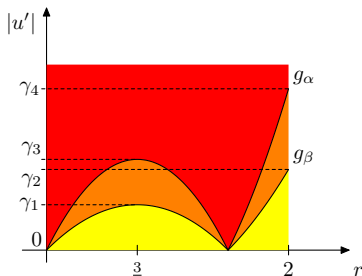
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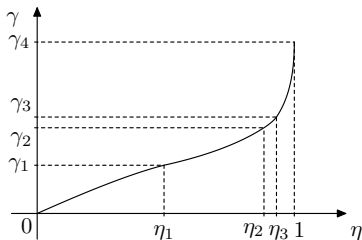
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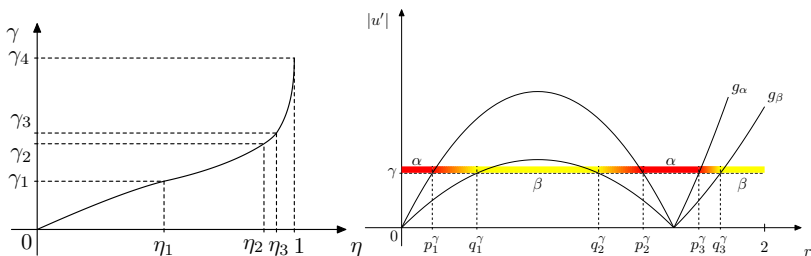
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Solving state equation

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

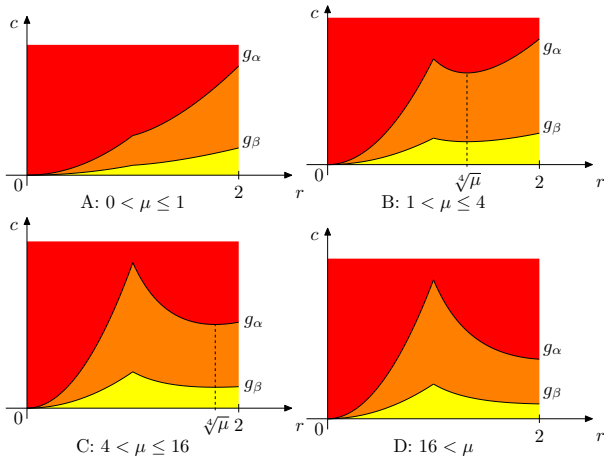
with

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

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

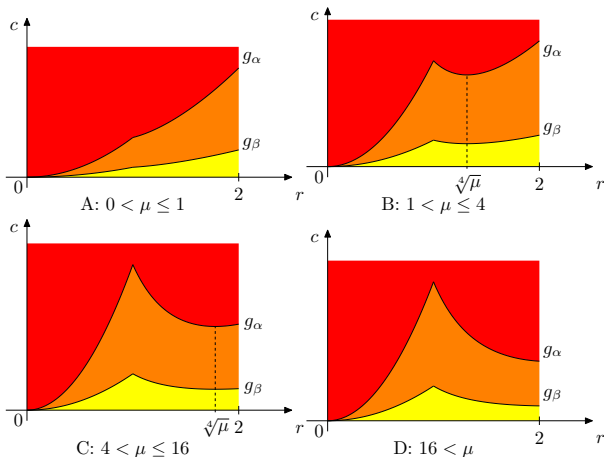


# Geometric interpretation of optimality conditions



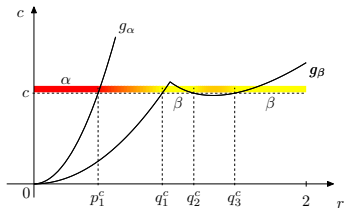


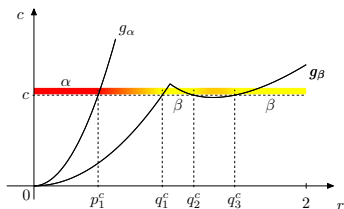
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As before, Lagrange multiplier can be numerically calculated from corresponding algebraic equation  $\int_{\Omega} \theta^* dx = \eta$ .



Optimal  $\theta^*$  for case B

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In orange region:

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

