## Exact solutions in optimal design problems for stationary diffusion equation

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Joint work with Marko Vrdoljak, University of Zagreb

## Multiple state optimal design problem

$\Omega \subseteq \mathbf{R}^{d}$ open and bounded, $f_{1}, \ldots, f_{m} \in \mathrm{~L}^{2}(\Omega)$ given; stationary diffusion equations with homogenous Dirichlet b. c.:

$$
\left\{\begin{array}{l}
-\operatorname{div}\left(\mathbf{A} \nabla u_{i}\right)=f_{i}  \tag{1}\\
u_{i} \in \mathrm{H}_{0}^{1}(\Omega)
\end{array} \quad, \quad i=1, \ldots, m\right.
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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 $\chi \in \mathrm{L}^{\infty}(\Omega ;\{0,1\})$, $\int_{\Omega} \chi d \mathbf{x}=q_{\alpha}$, for given $0<q_{\alpha}<|\Omega|$.

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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 compliances (total amounts of heat/electrical energy dissipated in $\Omega$ ):

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$$
I(\chi):=\sum_{i=1}^{m} \mu_{i} \int_{\Omega} f_{i} u_{i} d \mathbf{x} \rightarrow \min , \quad \chi \in \mathrm{~L}^{\infty}(\Omega ;\{0,1\})
$$

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

Murat \& Tartar
Lurie \& Cherkaev


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\begin{array}{rll}
\chi \in \mathrm{L}^{\infty}(\Omega ;\{0,1\}) & \cdots & \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) \text { a.e. on } \Omega \\
\text { classical material } & & \text { composite mateiral - relaxation }
\end{array}
$$

## Composite material

## Definition

If a sequence of characteristic functions $\chi_{\varepsilon} \in \mathrm{L}^{\infty}(\Omega ;\{0,1\})$ and conductivities

$$
\mathbf{A}^{\varepsilon}(x)=\chi_{\varepsilon}(x) \alpha \mathbf{I}+\left(1-\chi_{\varepsilon}(x)\right) \beta \mathbf{I}
$$

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 $\left(\chi_{\varepsilon}\right)$.

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Example - simple laminates: if $\chi_{\varepsilon}$ depend only on $x_{1}$, then

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\mathbf{A}^{*}=\operatorname{diag}\left(\lambda_{\theta}^{-}, \lambda_{\theta}^{+}, \lambda_{\theta}^{+}, \ldots, \lambda_{\theta}^{+}\right),
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\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:

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\mathcal{A}:=\left\{(\theta, \mathbf{A}) \in \mathrm{L}^{\infty}\left(\Omega ;[0,1] \times \mathrm{M}_{d}(\mathbf{R})\right): \int_{\Omega} \theta d \mathbf{x}=q_{\alpha}, \mathbf{A} \in \mathcal{K}(\theta) \text { a.e. }\right\}
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\lambda_{\theta}^{-} \leq \lambda_{j} \leq \lambda_{\theta}^{+} \quad j=1, \ldots, 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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\begin{aligned}
& I(\theta)=\int_{\Omega} f u d \mathbf{x} \longrightarrow \min \\
& \mathcal{T}=\left\{\theta \in \mathrm{L}^{\infty}(\Omega ;[0,1]): \int_{\Omega} \theta=q_{\alpha}\right\} \\
& \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 \mathrm{H}_{0}^{1}(\Omega)
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\end{aligned}
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B. Multiple state equations: Simpler relaxation fails, but in spherically symmetric case it can be done!
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## Minimization problem $\min _{\mathcal{B}} J$

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

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\mathcal{B} & \ldots \quad \int_{\Omega} \theta d \mathbf{x}=q_{\alpha} \\
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$\mathcal{B}$ is convex and compact and $J$ is continuous on $\mathcal{B}$, so there is a solution of $\min _{\mathcal{B}} \mathrm{J}$.


University of Osijek - Department of Mathematics
$\min _{\mathcal{B}} J \Longleftrightarrow \min _{\mathcal{T}} I$

Theorem

- There is unique $\mathrm{u}^{*} \in \mathrm{H}_{0}^{1}\left(\Omega ; \mathbf{R}^{m}\right)$ which is the state for every solution of $\min _{\mathcal{B}} J$ and $\min _{\mathcal{T}} l$.

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- If $\left(\theta^{*}, \mathbf{A}^{*}\right)$ is an optimal design for the problem $\min _{\mathcal{B}} J$, then $\theta^{*}$ is optimal design for $\min _{\mathcal{T}}$ I.


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- If $\left(\theta^{*}, \mathbf{A}^{*}\right)$ is an optimal design for the problem $\min _{\mathcal{B}} \mathrm{J}$, then $\theta^{*}$ is optimal design for $\min _{\mathcal{T}} \mathrm{l}$.
- Conversely, if $\theta^{*}$ is a solution of optimal design problem $\min _{\mathcal{T}}$ I, then any $\left(\theta^{*}, \mathbf{A}^{*}\right) \in \mathcal{B}$ satisfying $\mathbf{A}^{*} \nabla u_{i}^{*}=\lambda_{\theta^{*}}^{+} \nabla u_{i}^{*}$ almost everywhere on $\Omega$ (e.g. $\mathbf{A}^{*}=\lambda_{\theta^{*}}^{+} \mathbf{I}$ ) is an optimal design for the problem $\min _{\mathcal{B}} \mathrm{J}$.


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- Conversely, if $\theta^{*}$ is a solution of optimal design problem $\min _{\mathcal{T}}$ I, then any $\left(\theta^{*}, \mathbf{A}^{*}\right) \in \mathcal{B}$ satisfying $\mathbf{A}^{*} \nabla u_{i}^{*}=\lambda_{\theta^{*}}^{+} \nabla u_{i}^{*}$ almost everywhere on $\Omega$ (e.g. $\mathbf{A}^{*}=\lambda_{\theta^{*}}^{+}$) is an optimal design for the problem $\min _{\mathcal{B}} \mathrm{J}$.
- If $m<d$, then there exists minimizer $\left(\theta^{*}, \mathbf{A}^{*}\right)$ for $J$ on $\mathcal{B}$, such that $\left(\theta^{*}, \mathbf{A}^{*}\right) \in \mathcal{A}$, and thus it is also minimizer for $J$ on $\mathcal{A}$.


## Spherical symmetry: $\min _{\mathcal{A}} J \Longleftarrow \min _{\mathcal{B}} J \Longleftrightarrow \min _{\mathcal{T}} I$

Theorem
Let $\Omega \subseteq \mathbf{R}^{d}$ be spherically symmetric, and let the right-hand sides $f_{i}=f_{i}(r), r \in \omega, i=1, \ldots, m$ be radial functions. Then there exists a minimizer $\left(\theta^{*}, \mathbf{A}^{*}\right)$ 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 $\theta^{*}: \Omega \longrightarrow \mathbf{R}$ as the average value over spheres of $\theta$ : for $r \in \omega$ we take

$$
\theta^{*}(r):=f_{\partial B(0, r)} \theta d S,
$$

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

## Spherical symmetry...cont.

## Theorem

b) For any radial minimizer $\theta^{*}$ of I over $\mathcal{T}$, let us define $\mathbf{A}^{*}$ as a simple laminate layered with respect to a radial direction $\mathbf{e}_{r}$, as below, and local proportion of the first material $\theta^{*}$. To be specific, we can define $\mathbf{A}^{*}: \Omega \longrightarrow \mathrm{M}_{d}(\mathbf{R})$ in the following way:

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- If $\mathbf{x}=r \mathrm{e}_{1}=(r, 0,0, \ldots, 0)$, then

$$
\mathbf{A}^{*}(\mathbf{x}):=\operatorname{diag}\left(\lambda_{\theta^{*}}^{+}(r), \lambda_{\theta^{*}}^{-}(r), \lambda_{\theta^{*}}^{+}(r), \ldots, \lambda_{\theta^{*}}^{+}(r)\right) .
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- For all other $\mathbf{x} \in \Omega$, we take the unique rotation $\mathbf{R}(\mathbf{x}) \in S O(d)$ such that $\mathbf{x}=|\mathbf{x}| \mathbf{R}(\mathbf{x}) \mathrm{e}_{1}$, and define

$$
\mathbf{A}^{*}(\mathbf{x}):=\mathbf{R}(\mathbf{x}) \mathbf{A}^{*}\left(\mathbf{R}^{\tau}(\mathbf{x}) \mathbf{x}\right) \mathbf{R}^{\tau}(\mathbf{x}) .
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Then $\left(\theta^{*}, \mathbf{A}^{*}\right)$ is a radial optimal design for $\min _{\mathcal{B}} J$.

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- For all other $\mathbf{x} \in \Omega$, we take the unique rotation $\mathbf{R}(\mathbf{x}) \in S O(d)$ such that $\mathbf{x}=|\mathbf{x}| \mathbf{R}(\mathbf{x}) \mathrm{e}_{1}$, and define

$$
\mathbf{A}^{*}(\mathbf{x}):=\mathbf{R}(\mathbf{x}) \mathbf{A}^{*}\left(\mathbf{R}^{\tau}(\mathbf{x}) \mathbf{x}\right) \mathbf{R}^{\tau}(\mathbf{x}) .
$$

Then $\left(\theta^{*}, \mathbf{A}^{*}\right)$ is a radial optimal design for $\min _{\mathcal{B}} J$. Moreover, $\left(\theta^{*}, \mathbf{A}^{*}\right) \in \mathcal{A}$, and thus it is also a solution for $\min _{\mathcal{A}} \mathrm{J}$.

## 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
or equivalently

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

$$
\begin{aligned}
& \sum_{\substack{i=1 \\
m}}^{m} \mu_{i}\left|\nabla u_{i}^{*}\right|^{2}>c \Rightarrow \theta^{*}=0 \\
& \sum_{i=1}^{m} \mu_{i}\left|\nabla u_{i}^{*}\right|^{2}<c \Rightarrow \theta^{*}=1
\end{aligned}
$$

Ball $\Omega=B(\mathbf{0}, 2) \subseteq \mathbf{R}^{2}$ with nonconstant right-hand side
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C: $4<\mu \leq 16$


## Optimal $\theta^{*}$ for case B

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

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Thank you for your attention!


