

Explicit solutions of multiple state optimal design problems

Krešimir Burazin

J. J. STROSSMAYER UNIVERSITY OF OSIJEK DEPARTMENT OF MATHEMATICS Trg Ljudevita Gaja 6 31000 Osijek, Hrvatska http://www.mathos.unios.hr

kburazin@mathos.hr



Joint work with Marko Vrdoljak





[6TH CROATIAN MATHEMATICAL CONGRESS, ZAGREB]

June 2016

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 Ω):

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

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 **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 **A** which minimizes the weighted sum of energies (total amounts of heat/electrical energy dissipated in Ω):

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

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 **A** is a mixture of two isotropic materials with conductivities $0 < \alpha < \beta$: **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 **A** which minimizes the weighted sum of energies (total amounts of heat/electrical energy dissipated in Ω):

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

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 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):

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

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 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):

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

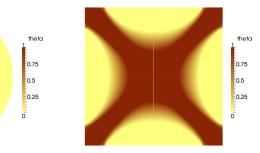
Energy minimization and relaxation problem Convex minimization Examples

Posing the problem



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

Murat & Tartar



Lurie & Cherkaev



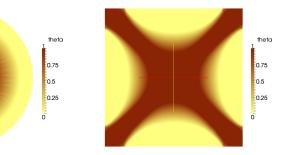
Energy minimization and relaxation problem Convex minimization Examples

Posing the problem



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

Murat & Tartar



Lurie & Cherkaev

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

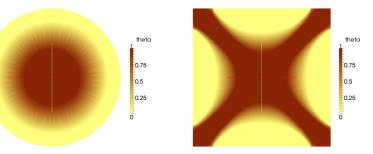
 $\mathbf{A} = \chi \alpha \mathbf{I} + (1 - \chi) \beta \mathbf{I}$ $\mathbf{A} \in \mathcal{K}(\theta)$ a.e. on Ω classical material composite mateiral - relaxation

Posing the problem Relaxation



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

Murat & Tartar



Lurie & Cherkaev

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

 $\begin{array}{ll} \cdots & \theta \in \mathrm{L}^{\infty}(\Omega; [0,1]) \\ & \mathbf{A} \in \mathcal{K}(\theta) \quad \text{a.e. on } \Omega \\ & \text{composite mateiral - relaxation} \end{array}$

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}^{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

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

where

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

Set of all composites:

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

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}^{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

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

where

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

Set of all composites:

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

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}^{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

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

where

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

Set of all composites:

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

Posing the problem Relaxation



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

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

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

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

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

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

 $\min_{\mathcal{A}} J$ 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 θ find all possible homogenised (effective) tensors A^*

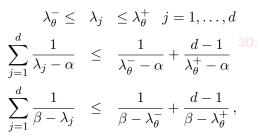
Energy minimization and relaxation

minimization

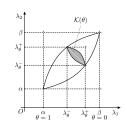
Convex

Examples

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



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



Posing the problem Relaxation

2D:

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

problem

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

Energy minimization and relaxation

minimization

Convex

Examples

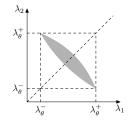
 $\mathcal{K}(\theta)$ is given in terms of eigenvalues (Murat & Tartar; Lurie & Cherkaev):

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

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

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

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



Energy minimization and relaxation problem

Posing the problem Relaxation

2D:



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

G-closure problem: for given θ find all possible homogenised (effective) tensors A*

minimization

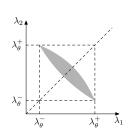
Convex

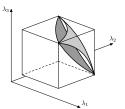
Examples

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

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

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





Energy minimization and relaxation problem

Posing the problem Relaxation

2D:



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

G-closure problem: for given θ find all possible homogenised (effective) tensors A*

minimization

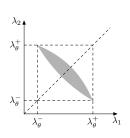
Convex

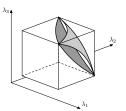
Examples

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

$$\begin{aligned} \lambda_{\theta}^{-} &\leq \lambda_{j} &\leq \lambda_{\theta}^{+} \quad j = 1, \dots, d \\ \sum_{j=1}^{d} \frac{1}{\lambda_{j} - \alpha} &\leq \frac{1}{\lambda_{\theta}^{-} - \alpha} + \frac{d - 1}{\lambda_{\theta}^{+} - \alpha} \overset{\text{3D:}}{\sum_{j=1}^{d} \frac{1}{\beta - \lambda_{j}}} &\leq \frac{1}{\beta - \lambda_{\theta}^{-}} + \frac{d - 1}{\beta - \lambda_{\theta}^{+}}, \\ \min A U \text{ is a proper relaxation of} \end{aligned}$$

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





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.

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \right\} \\ \theta \in \mathcal{T} \text{ , and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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.

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \right\} \\ \theta \in \mathcal{T} \text{ , and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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.

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_\alpha \right\} \\ \theta \in \mathcal{T} \text{, and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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.

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \right\} \\ \theta \in \mathcal{T} \text{, and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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.

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

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \right\} \\ \theta \in \mathcal{T} \text{ , and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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

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

 $\min_{\mathcal{T}} I$

A. Single state equation: [Murat & Tartar] This problem can be rewritten as a simpler convex minimization problem.

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

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_{\alpha} \right\} \\ \theta \in \mathcal{T} \text{, and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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

 $\min_{\mathcal{T}} I$

A. Single state equation: [Murat & Tartar] This problem can be rewritten as a simpler convex minimization problem.

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

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_\alpha \right\} \\ \theta \in \mathcal{T} \text{, and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

 $\min_{A} J$

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

 $\min_{\mathcal{T}} I$

A. Single state equation: [Murat & Tartar] This problem can be rewritten as a simpler convex minimization problem.

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

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

$$\begin{split} I(\theta) &= \sum_{i=1}^{m} \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \longrightarrow \min \\ \mathcal{T} &= \left\{ \theta \in \mathcal{L}^{\infty}(\Omega; [0, 1]) : \int_{\Omega} \theta = q_\alpha \right\} \\ \theta \in \mathcal{T} \text{, and } u_i \text{ determined uniquely by} \\ \left\{ \begin{array}{l} -\operatorname{div} \left(\lambda_{\theta}^+ \nabla u_i\right) = f_i \\ u_i \in \mathcal{H}_0^1(\Omega) \end{array} \right. i = 1, \dots, m \end{split}$$

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

 $\min_{A} J$

Simpler problem Spherically symmetric case



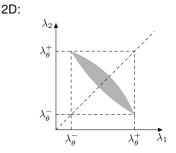
 $\min_{\mathcal{B}} J$

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

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



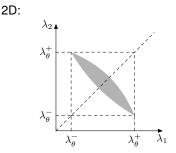
 $\min_{\mathcal{B}} J$

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

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 $\min_{\mathcal B} J$.



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

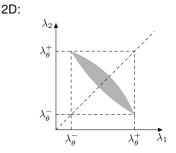


$$\mathcal{A} := \{(\theta, \mathbf{A}) \in \mathcal{L}^{\infty}(\Omega; [0, 1] \times \mathcal{M}_{d}(\mathbf{R})) : \int_{\Omega} \theta \, d\mathbf{x} = q_{\alpha} \,, \; \mathbf{A} \in \mathcal{K}(\theta) \text{ a.e. } \}$$

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



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

- 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_B J, then θ* is optimal design for min_T I.
- Conversely, if θ* is a solution of optimal design problem min_T I, then any (θ*, A*) ∈ B satisfying A*∇u^{*}_i = λ⁺_{θ*}∇u^{*}_i almost everywhere on Ω (e.g. A* = λ⁺_{θ*}I) is an optimal design for the problem min_B J.
- If aditionally m < d, then above is valid for min_A J instead min_B J and optimal design can be realized as a simple laminate.

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_B J, then θ* is optimal design for min_T I.
- Conversely, if θ^{*} is a solution of optimal design problem min_T I, then any (θ^{*}, A^{*}) ∈ B satisfying A^{*}∇u^{*}_i = λ⁺_{θ^{*}}∇u^{*}_i almost everywhere on Ω (e.g. A^{*} = λ⁺_{θ^{*}}I) is an optimal design for the problem min_B J.
- If aditionally m < d, then above is valid for min_A J instead min_B J and optimal design can be realized as a simple laminate.

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_B J, then θ* is optimal design for min_T I.
- Conversely, if θ^{*} is a solution of optimal design problem min_T I, then any (θ^{*}, A^{*}) ∈ B satisfying A^{*}∇u^{*}_i = λ⁺_{θ^{*}}∇u^{*}_i almost everywhere on Ω (e.g. A^{*} = λ⁺_{θ^{*}}I) is an optimal design for the problem min_B J.
- If aditionally m < d, then above is valid for min_A J instead min_B J and optimal design can be realized as a simple laminate.

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_B J, then θ* is optimal design for min_T I.
- Conversely, if θ^{*} is a solution of optimal design problem min_T I, then any (θ^{*}, A^{*}) ∈ B satisfying A^{*}∇u^{*}_i = λ⁺_{θ^{*}}∇u^{*}_i almost everywhere on Ω (e.g. A^{*} = λ⁺_{θ^{*}}I) is an optimal design for the problem min_B J.
- If aditionally m < d, then above is valid for min_A J instead min_B J and optimal design can be realized as a simple laminate.

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

a) For any minimizer θ of functional I over \mathcal{T} , let us define a radial function $\theta^* : \Omega \longrightarrow \mathbf{R}$ 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 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 (θ^*, \mathbf{A}^*) of the optimal design problem $\min_{\mathcal{A}} J$ which is a radial function. More precisely,

a) For any minimizer θ of functional I over \mathcal{T} , let us define a radial function $\theta^* : \Omega \longrightarrow \mathbf{R}$ 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 problem Spherically symmetric case



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

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

in spherical basis (e_r(x), e_{φ1}(x), e_{φ2}(x), ..., e_{φd-1}(x)). Then (θ*, A*) is an optimal design for min_B J. Moreover, (θ*, A*) ∈ A, and thus it is also a solution for min_A J.
c) If (θ, A) ∈ A is a solution of the relaxed problem min_A J with corresponding state function ũ, then θ is optimal for min_T 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



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

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

in spherical basis (e_r(x), e_{φ1}(x), e_{φ2}(x), ..., e_{φd-1}(x)). Then (θ*, A*) is an optimal design for min_B J. Moreover, (θ*, A*) ∈ A, and thus it is also a solution for min_A J.
c) If (θ, A) ∈ A is a solution of the relaxed problem min_A J with corresponding state function ũ, then θ is optimal for min_T I, and (θ, A) is also a minimizer of J on B. Consequently, we have ũ = u*, and Ae_r = λ₊(θ)e_r, almost everywhere.

Simpler problem Spherically symmetric case



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

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

in spherical basis $(\mathbf{e}_r(\mathbf{x}), \mathbf{e}_{\phi_1}(\mathbf{x}), \mathbf{e}_{\phi_2}(\mathbf{x}), \dots, \mathbf{e}_{\phi_{d-1}}(\mathbf{x}))$. Then (θ^*, \mathbf{A}^*) is an optimal design for $\min_{\mathcal{B}} J$. Moreover, $(\theta^*, \mathbf{A}^*) \in \mathcal{A}$, and thus it is also a solution for $\min_{\mathcal{A}} J$.

c) If (θ, A) ∈ A is a solution of the relaxed problem min_A J with corresponding state function ũ, then θ is optimal for min_T I, and (θ, A) is also a minimizer of J on B. Consequently, we have ũ = u*, and Ae_r = λ₊(θ)e_r, almost everywhere.

Simpler problem Spherically symmetric case



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

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

in spherical basis (e_r(x), e_{φ1}(x), e_{φ2}(x), ..., e_{φd-1}(x)). Then (θ*, A*) is an optimal design for min_B J. Moreover, (θ*, A*) ∈ A, and thus it is also a solution for min_A J.
c) If (θ, A) ∈ A is a solution of the relaxed problem min_A J with corresponding state function ũ, then θ is optimal for min_T I, and (θ, A) is also a minimizer of J on B. Consequently, we have ũ = u*, and Ae_r = λ₊(θ)e_r, almost everywhere.

Simpler problem Spherically symmetric case



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

 θ^{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: $\gamma := \sqrt{c} > 0$

 γ is uniquely determined by

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

One state Multiple states

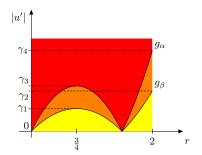


Ball $\Omega = B(0,2) \subseteq \mathbb{R}^2$ with nonconstant right-hand side In all examples $\alpha = 1, \beta = 2$, one state equation f(r) = 1 - r

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

 γ is uniquely determined by

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



One state Multiple states

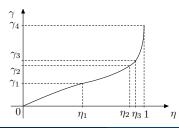


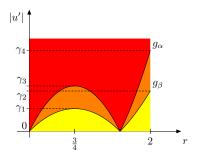
Ball $\Omega = B(0, 2) \subseteq \mathbf{R}^2$ with nonconstant right-hand side In all examples $\alpha = 1, \beta = 2$, one state equation f(r) = 1 - r

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

 γ is uniquely determined by

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





One state Multiple states

|u'|

 γ_4

 γ_3

 γ_2 γ_1 .



 g_{α}

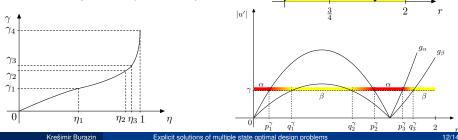
 g_{β}

Ball $\Omega = B(0, 2) \subseteq \mathbf{R}^2$ with nonconstant right-hand side In all examples $\alpha = 1, \beta = 2$, one state equation f(r) = 1 - r

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

 γ is uniquely determined by

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



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



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

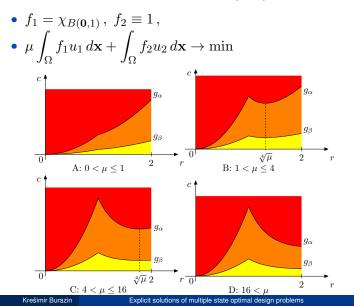
•
$$f_1 = \chi_{B(\mathbf{0},1)}, \ f_2 \equiv 1,$$

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

One state Multiple states



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



13/14

One state Multiple states



Optimal θ^* for case B

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

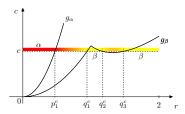
Thank you for your attention!

One state Multiple states



Optimal θ^* for case B

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



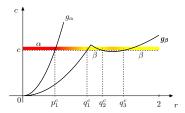
Thank you for your attention!

One state Multiple states



Optimal θ^* for case B

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



Thank you for your attention!