

Numerical approximation of classical optimal design on annuli

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The necessary and sufficient condition of optimality

Statement of the problem

Let $\Omega \subset \mathbb{R}^d$ be open and bounded set. It consists of two phases each with different isotropic conductivity: α, β ($0 < \alpha < \beta$).
 q_α is the prescribed volume of the first phase ($0 < q_\alpha < |\Omega|$).
 $\chi \in L^\infty(\Omega, \{0, 1\})$ a measurable characteristic function.

Conductivity can be expressed as

$$\mathbf{A}(\chi) := \chi\alpha \mathbf{I} + (1 - \chi)\beta \mathbf{I},$$

where

$$\int_{\Omega} \chi(\mathbf{x}) \, d\mathbf{x} = q_\alpha.$$

State functions $u_i \in H_0^1(\Omega)$, $i = 1, 2, \dots, m$ are solutions of the following boundary value problems:

$$(1) \quad \begin{cases} -\operatorname{div}(\mathbf{A} \nabla u_i) = f_i & \text{in } \Omega \\ u_i = 0 & \text{on } \partial\Omega, \end{cases} \quad i = 1, 2, \dots, m.$$

Energy functional:

$$I(\chi) := \sum_{i=1}^m \mu_i \int_{\Omega} f_i(\mathbf{x}) u_i(\mathbf{x}) \, d\mathbf{x}.$$

where $\mu_i > 0$, $i = 1, 2, \dots, m$.

Optimal design problem:

$$(2) \quad \begin{cases} I(\chi) = \sum_{i=1}^m \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \rightarrow \max \\ \text{s.t. } \chi \in L^\infty(\Omega, \{0, 1\}), \int_{\Omega} \chi \, d\mathbf{x} = q_\alpha, \\ \mathbf{u} \text{ solves (1) with } \mathbf{A} = \chi\alpha \mathbf{I} + (1 - \chi)\beta \mathbf{I}. \end{cases}$$

If solution χ exists for (2) we call it *classical solution*.

Important: For general optimal design problems the classical solutions usually do not exist.

Relaxed design: Effective conductivity

For characteristic functions relaxation consists of:

$$(3) \quad \chi \in L^\infty(\Omega, \{0, 1\}) \rightsquigarrow \theta \in L^\infty(\Omega, [0, 1]),$$

with $\int_{\Omega} \theta \, d\mathbf{x} := q_\alpha$.

Set of effective conductivities $\mathcal{K}(\theta)$:

$\mathbf{A} \in \mathcal{K}(\theta)$ iff there exists sequence of characteristic functions

$$\begin{cases} \chi_n \xrightarrow{L^\infty} \theta \\ \mathbf{A}^n = \chi_n \alpha \mathbf{I} + (1 - \chi_n) \beta \mathbf{I} \xrightarrow{H} \mathbf{A}. \end{cases}$$

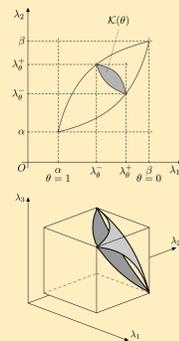
Visual representation of a set $\mathcal{K}(\theta)$

$\mathcal{K}(\theta)$ is given in terms of eigenvalues

$$\begin{aligned} \lambda_{\theta}^- &\leq \lambda_j \leq \lambda_{\theta}^+ \quad j = 1, \dots, d \\ \sum_{j=1}^d \frac{1}{\lambda_j - \alpha} &\leq \frac{1}{\lambda_{\theta}^- - \alpha} + \frac{d-1}{\lambda_{\theta}^+ - \alpha} \\ \sum_{j=1}^d \frac{1}{\beta - \lambda_j} &\leq \frac{1}{\beta - \lambda_{\theta}^-} + \frac{d-1}{\beta - \lambda_{\theta}^+}, \end{aligned}$$

where

$$\begin{aligned} \lambda_{\theta}^+ &= \theta\alpha + (1 - \theta)\beta \\ \frac{1}{\lambda_{\theta}^-} &= \frac{\theta}{\alpha} + \frac{1 - \theta}{\beta}. \end{aligned}$$



Relaxed problem A:

$$\mathcal{A} = \left\{ (\theta, \mathbf{A}) \in L^\infty(\Omega, [0, 1]) \times \operatorname{Sym}_d \mid \begin{array}{l} \int_{\Omega} \theta \, d\mathbf{x} = q_\alpha, \\ \mathbf{A}(\mathbf{x}) \in \mathcal{K}(\theta(\mathbf{x})), \text{ a.e. } \mathbf{x} \end{array} \right\}$$

Relaxed problem can be written as:

$$(A) \quad \max_{(\theta, \mathbf{A}) \in \mathcal{A}} J(\theta, \mathbf{A}) = \max_{(\theta, \mathbf{A}) \in \mathcal{A}} \sum_{i=1}^m \mu_i \int_{\Omega} f_i(\mathbf{x}) u_i(\mathbf{x}) \, d\mathbf{x}$$

Generalized (convex) problem B

Unfortunately, \mathcal{A} is not a convex set. To achieve convexity, an enlarged (artificial) set is introduced:

$$\mathcal{B} = \left\{ (\theta, \mathbf{A}) \in L^\infty(\Omega, [0, 1]) \times \operatorname{Sym}_d \mid \begin{array}{l} \int_{\Omega} \theta \, d\mathbf{x} = q_\alpha, \\ \lambda_{\theta(\mathbf{x})}^- \mathbf{I} \leq \mathbf{A}(\mathbf{x}) \leq \lambda_{\theta(\mathbf{x})}^+ \mathbf{I}, \text{ a.e. } \mathbf{x} \end{array} \right\}$$

and with it

$$(B) \quad \max_{(\theta, \mathbf{A}) \in \mathcal{B}} J(\theta, \mathbf{A}) = \max_{(\theta, \mathbf{A}) \in \mathcal{B}} \sum_{i=1}^m \mu_i \int_{\Omega} f_i(\mathbf{x}) u_i(\mathbf{x}) \, d\mathbf{x}$$

Using fluxes one can rewrite problem (B) as max-min problem and prove:

Theorem

Optimization problem (B) is equivalent to following optimization problem:

$$(I) \quad \begin{cases} I(\theta) = \sum_{i=1}^m \mu_i \int_{\Omega} f_i u_i \, d\mathbf{x} \rightarrow \max \\ \text{s.t. } \theta \in L^\infty(\Omega, [0, 1]), \int_{\Omega} \theta \, d\mathbf{x} = q_\alpha, \text{ where } \mathbf{u} \text{ satisfies} \\ -\operatorname{div}(\lambda_{\theta}^- \nabla u_i) = f_i, \quad u_i \in H_0^1(\Omega), \quad i = 1, \dots, m, \end{cases}$$

The necessary and sufficient condition of optimality

Define

$$\psi := \sum_{i=1}^m \mu_i |\sigma_i^*|^2.$$

Lemma

The necessary and sufficient condition of optimality for solution θ^* of optimal design problem (I) simplifies to the existence of a Lagrange multiplier $c \geq 0$ such that

$$(4) \quad \begin{cases} \psi = \sum_{i=1}^m \mu_i |\sigma_i^*|^2 > c \Rightarrow \theta^* = 1, \\ \psi = \sum_{i=1}^m \mu_i |\sigma_i^*|^2 < c \Rightarrow \theta^* = 0. \end{cases}$$

Analytical example on annulus for single state problem

For spherically symmetric problem such that:

$\Omega = R(\Omega)$ for any rotation R

f_i are radial functions

it can be proved that there exists radial solution θ_R^* of (I).

In particular, it can be shown that

$$\max_{(\theta, \mathbf{A}) \in \mathcal{A}} J(\theta, \mathbf{A}) = I(\theta_R^*).$$

Single state equation:

$$(5) \quad \begin{cases} -\operatorname{div}(\lambda_{\theta}^- \nabla u) = 1 & \text{in } \Omega \\ u = 0 & \text{on } \partial\Omega \end{cases}$$

where $\lambda_{\theta}^- = \left(\frac{\theta(x)}{\alpha} + \frac{1 - \theta(x)}{\beta} \right)^{-1}$.

Optimization problem:

For $\theta \in \mathcal{T} := \{ \theta \in L^\infty(\Omega, [0, 1]) : \int_{\Omega} \theta \, d\mathbf{x} = q_\alpha \}$

$$I(\theta) = \int_{\Omega} u \, d\mathbf{x} \rightarrow \max$$

$\Omega = \overline{K}(0, r_2) \setminus K(0, r_1)$

One can rewrite (5) in polar coordinates:

$$-\frac{1}{r^{d-1}} (r^{d-1} \lambda_{\theta}^- u'(r))' = 1 \quad \text{in } \langle r_1, r_2 \rangle, \quad u(r_1) = u(r_2) = 0.$$

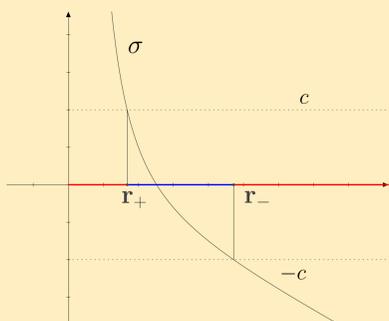
Observe that σ satisfies

$$\sigma = -\frac{r}{d} + \frac{\gamma}{r^{d-1}}, \quad \gamma > 0$$

$\sigma(r) : \langle 0, \infty \rangle \rightarrow \mathbb{R}$ is a strictly decreasing function.

The necessary and sufficient condition of optimality for θ^* states

$$\begin{cases} |\sigma^*| > c \Rightarrow \theta^* = 1, \\ |\sigma^*| < c \Rightarrow \theta^* = 0. \end{cases}$$



There are only three possible candidates for optimal design:

- 1) $\theta^*(r) = \begin{cases} 1, & r \in [r_1, r_+] \\ 0, & r \in [r_+, r_-] \\ 1, & r \in [r_-, r_2] \end{cases}$ **alpha-beta-alpha**
- 2) $\theta^*(r) = \begin{cases} 1, & r \in [r_1, r_+] \\ 0, & r \in [r_+, r_2] \end{cases}$ **alpha-beta**
- 3) $\theta^*(r) = \begin{cases} 0, & r \in [r_1, r_-] \\ 1, & r \in [r_-, r_2] \end{cases}$ **beta-alpha**

Direct calculations

Necessary and sufficient condition of optimality can also be expressed as a non-linear system (unknowns γ, c, r_+, r_-):

$$(6) \quad \begin{cases} S_d \int_{r_1}^{r_2} \theta(\rho) \rho^{d-1} \, d\rho = q_\alpha \\ u(r_2) = 0 \iff \gamma \int_{r_1}^{r_2} \left(\frac{1}{a(\rho) \rho^{d-1}} \right) \, d\rho = \int_{r_1}^{r_2} \frac{\rho}{a(\rho)} \, d\rho \\ \sigma(r_+) = c, \quad \sigma(r_-) = -c, \quad \text{where } c > 0 \end{cases}$$

where

$$\sigma(r) = \frac{\gamma}{r^{d-1}} - \frac{r}{d}, \quad \& \quad a(r) = \left(\frac{\theta(r)}{\alpha} + \frac{1 - \theta(r)}{\beta} \right)^{-1}.$$

Results for $d = 2, 3$

3) case **beta-alpha**

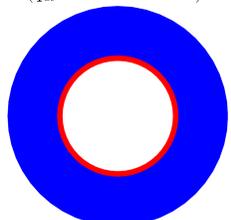


Non-linear system (6) does not admit a solution (proved for $d = 2$ and $d = 3$).

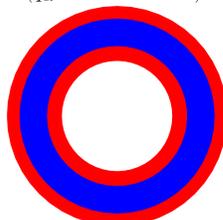
Therefore, cases 1) and 2) should be considered as only possible solutions. One can easily prove if q_α is very small, case **alpha-beta** is always solution (for arbitrary chosen parameters α, β, r_1, r_2).

Furthermore, one can numerically obtain critical value for which optimal design changes from case **alpha-beta** to **alpha-beta-alpha**.

alpha-beta
($q_\alpha < \text{critical value}$)



alpha-beta-alpha
($q_\alpha > \text{critical value}$)



Remark:

• Problem can be easily generalized to multi-state problem for example $m = 2$;

$$f_1(r) = 1, \quad f_2(r) = \frac{b}{r(b-r)^2}, \quad \text{where } b > r_2$$

• Existence of such solutions is important for any numerical method like shape derivative method.

Gradient method using shape derivative

Perturbation of the set Ω is given with

$$\Omega_t = (\operatorname{Id} + t\psi)\Omega$$

where $\psi \in W^{k, \infty}(\mathbb{R}^d, \mathbb{R}^d)$

If t is small (i.e. $\|t\psi\|_{W^{k, \infty}} \ll 1$) mapping $\operatorname{Id} + t\psi$ is homeomorphism. This allows us to define shape derivative:

Definition (Shape derivative)

Let $J = J(\Omega)$ be a shape functional. J is said to be shape differentiable at Ω in direction ψ if

$$J'(\Omega, \psi) := \lim_{t \searrow 0} \frac{J(\Omega_t) - J(\Omega)}{t}$$

exists and the mapping $\psi \mapsto J'(\Omega, \psi)$ is linear and continuous.

$J'(\Omega, \psi)$ is called the **shape derivative**.

For our optimal design problem: shape derivative is given with:

$$\begin{aligned} J'(\Omega, \psi) &= \int_{\Omega} \mathbf{A}(-\operatorname{div}(\psi) + \nabla\psi + \nabla\psi^T) \nabla u_0 \cdot \nabla u_0 \, d\mathbf{x} \\ &+ \int_{\Omega} 2(\operatorname{div}(\psi)f + \nabla f \cdot \psi) u_0 \, d\mathbf{x} \end{aligned}$$

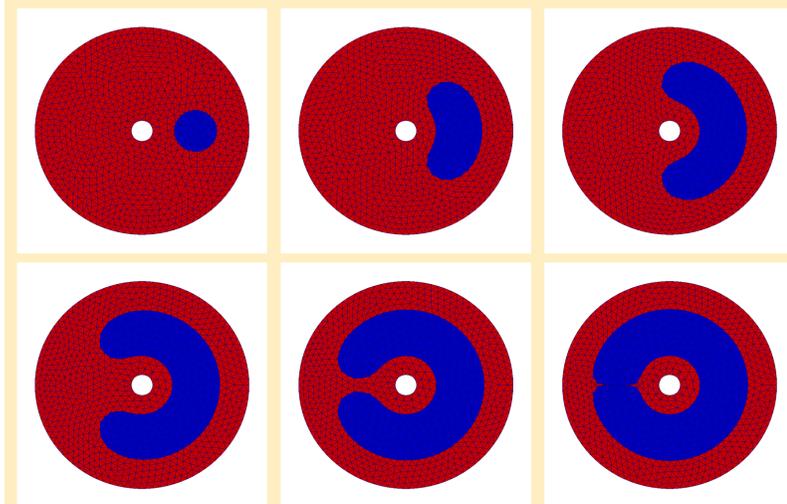
where u_0 is solution of BVP (1) on domain Ω with \mathbf{A} .

Vector field $\psi \in H_0^1(\Omega)$ is constructed from:

$$\int_{\Omega} \nabla\psi : \nabla\varphi + \int_{\Omega} \psi \cdot \varphi = J'(\Omega, \varphi), \quad \forall \varphi \in H_0^1(\Omega)$$

The shape is evolved by gradually moving the boundary between phases.

Numerical results:



References

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