

Shape optimization of the ground state for two-phase conductors

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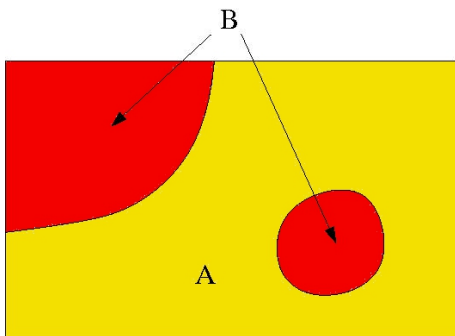
joint work with Carlos Conca and Rajesh Mahadevan

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Problem Statement

Find the optimal distribution of two conducting materials A and B of given volume and conductivities α and β in a fixed domain Ω in order to minimize the ground state eigenvalue.



Problem setting

Eigenvalue problem

- $\Omega \subset \mathbb{R}^d$, $0 < \alpha < \beta$, $0 < m < |\Omega|$
- $B \subset \Omega$ measurable, $A = \Omega \setminus B$
-

$$-\operatorname{div}(\sigma(B)\nabla u) = \lambda(B)u \text{ in } \Omega$$

$$u = 0 \text{ on } \partial\Omega.$$

- $\sigma(B) = \alpha\chi_A + \beta\chi_B$, (χ_A and χ_B are indicator functions)
- $\lambda(B)$ is the first eigenvalue or **ground state**.

Shape Optimization Problem

$$\begin{array}{ll} \text{minimize} & \lambda(B) \\ \text{subject to} & B \in \mathcal{B} := \{B \subset \Omega, B \text{ measurable}, |B| = m\} \end{array}$$

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Known results

Existence

- Open question for general geometries of Ω .
- Existence of relaxed solutions:
 - > Steven Cox and Robert Lipton (1996). “Extremal eigenvalue problems for two-phase conductors.” In: *Arch. Rational Mech. Anal.* 136.2, pp. 101–117
 - Existence of a radially symmetric solution when Ω is a ball.
 - > A. Alvino, G. Trombetti, and P.-L. Lions (1989). “On optimization problems with prescribed rearrangements.” In: *Nonlinear Anal.* 13.2, pp. 185–220
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Characterization of minimizers

Can we find some explicit solutions?

- The problem is solved explicitly in 1D.
- > M. G. Krein (1955). “On certain problems on the maximum and minimum of characteristic values and on the Lyapunov zones of stability.” In: *Amer. Math. Soc. Transl. (2)* 1, pp. 163–187

Conjecture (Conca et al., Dambrine)

- When $\Omega \subset \mathbb{R}^d$ is a ball, the minimizer is also a ball:

$$B^* = B(0, r^*) = \operatorname{argmin}_{B \in \mathcal{B}} \lambda(B)$$

Solution in a particular case

We exhibit **global minimizers** in low contrast regime.

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Asymptotic Expansion

- **Low contrast regime:** $\beta = \alpha + \varepsilon$ with $\varepsilon > 0$ small.
- Conductivity $\sigma^\varepsilon = \alpha + \varepsilon \chi_B$

Theorem (Rellich)

The first eigenvalue λ^ε of

$$\begin{aligned} -\operatorname{div}(\sigma^\varepsilon \nabla u^\varepsilon) &= \lambda^\varepsilon u^\varepsilon \text{ in } \Omega, \\ u^\varepsilon &= 0 \text{ on } \partial\Omega, \end{aligned}$$

is an analytic function of ε in a neighbourhood of $\varepsilon = 0$ and the positive eigenfunction u^ε satisfying

$$\int_{\Omega} (u^\varepsilon)^2 = 1$$

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Asymptotic Expansion

We plug the following ansätze:

$$u^\varepsilon = v_0 + \varepsilon v_1 + \dots,$$

$$\lambda^\varepsilon = \lambda_0 + \varepsilon \lambda_1 + \dots,$$

in $-\operatorname{div}(\sigma^\varepsilon \nabla u^\varepsilon) = \lambda^\varepsilon$ and $u^\varepsilon = 0$. Gather terms of similar order in ε :

$$-\operatorname{div}(\alpha \nabla v_0) = \lambda_0 v_0 \text{ in } \Omega, \quad (2.1)$$

$$v_0 = 0 \text{ on } \partial\Omega. \quad (2.2)$$

$$-\operatorname{div}(\alpha \nabla v_1) - \lambda_0 v_1 = \operatorname{div}(\chi_B \nabla v_0) + \lambda_1 v_0 \text{ in } \Omega, \quad (2.3)$$

$$v_1 = 0 \text{ on } \partial\Omega. \quad (2.4)$$

(2.3)-(2.4) has a solution if and only if (Fredholm alternative)

$$\int_{\Omega} \operatorname{div}(\chi_B \nabla v_0) v_0 + \lambda_1 \int_{\Omega} v_0^2 = 0.$$

Asymptotic Expansion

Using $\int_{\Omega} v_0^2 = 1$ we obtain

$$\lambda_1 = - \int_{\Omega} \operatorname{div}(\chi_B \nabla v_0) v_0 \implies \lambda_1 = \lambda_1(B) = \int_B |\nabla v_0|^2.$$

Theorem

If $B_{\varepsilon}^* \in \mathcal{B}$ is a minimizer of $\lambda^{\varepsilon}(\cdot)$ then:

$$\left| \lambda_1(B_{\varepsilon}^*) - \inf_{B \in \mathcal{B}} \lambda_1(B) \right| \leq C \varepsilon^{\frac{1}{2}}.$$

Optimality conditions

Theorem

- *There exists $c^* \geq 0$ such that whenever B is a measurable subset of Ω satisfying*

$$\{x : |\nabla v_0(x)| < c^*\} \subset B \subset \{x : |\nabla v_0(x)| \leq c^*\}$$

and $|B| = m$, then B is a solution for the problem of minimizing $\lambda_1(B)$ over $B \in \mathcal{B}$.

- *If $\{x : |\nabla v_0(x)| = c^*\}$ is of measure zero, then the unique solution (up to a set of measure zero) is the set*

$$B^* = \{x : |\nabla v_0(x)| < c^*\}.$$

This is the case if Ω is a disk.

The Disk Case

- $\Omega = \mathbb{B}(0, 1)$ in 2D or 3D
- The solution of $-\operatorname{div}(\alpha \nabla v_0) = \lambda_0 v_0$ in Ω and $v_0 = 0$ on $\partial\Omega$ is radial: $v_0(x) = w(|x|)$

$$r^2 w_0''(r) + (d-1)rw_0'(r) + r^2 \frac{\lambda_0}{\alpha} w_0(r) = 0,$$

$$w_0'(0) = 0, \quad w_0(1) = 0.$$

- In 2D, $w_0(r) = J_0(\eta_d r)$ where J_0 is the Bessel function of the first kind and η_d is its first zero.

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- In 2D, $w_0(r) = J_0(\eta_d r)$ where J_0 is the Bessel function of the first kind and η_d is its first zero.

The Disk Case

$|\nabla v_0|^2(x) = (w_1(r))^2 := (-w_0'(r))^2$ and the solution is:

$$\lambda_1(B) = \int_B |\nabla v_0|^2 \implies B^* = \{x : w_1(r) < c^*\}$$

where c^* is such that $|B^*| = m$.

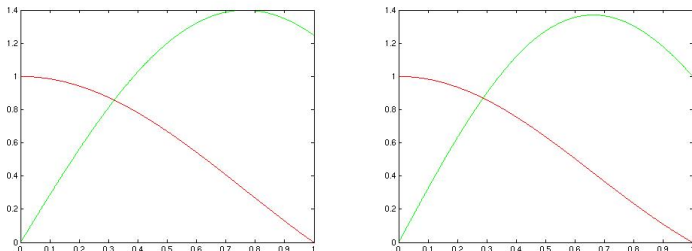


Figure : $w_0(r)$ (red), and $w_1(r) := -w_0'(r)$ (green) in dimensions $d = 2$ (left) and $d = 3$ (right), w_1 increasing on $[0, r_d^1]$ and decreasing on $[r_d^1, 1]$, and r_d^0 is such that $w_1(r_d^0) = w_1(1)$.

The Disk Case

Theorem

The solution $B^* = \min_{B \in \mathcal{B}} \lambda_1(B)$ is of two possible types.

There exists $\bar{m} = \omega_d (r_d^0)^d$ such that

- Type I: If $m \leq \bar{m}$ then $B^* = B(0, (m/\omega_d)^{1/d})$ or,
- Type II: If $m > \bar{m}$ then there exists ξ^0 and ξ^1 with $(m/\omega_d)^{1/d} < \xi^0 < \xi^1 < 1$ such that

$$B^* = B(0, \xi^0) \cup \left(B(0, 1) \setminus \overline{B(0, \xi^1)} \right).$$

Theorem

When $\Omega = B(0, 1)$, for $\beta = \alpha + \varepsilon$ sufficiently close to α and $m > \bar{m}$, $B = \mathbb{B}(0, r^*)$ does not minimize $\lambda^\varepsilon(B)$ in \mathcal{B} .

Low contrast regime - other geometries

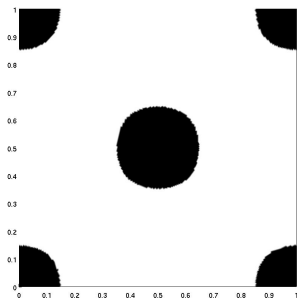


Figure : Optimal distribution of the material B (black) and A (white) when Ω is a square in low contrast regime. The set B contains the corners and the center. $m/|\Omega| \approx 14\%$.

Low contrast regime - other geometries

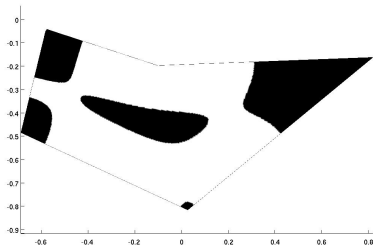


Figure : Optimal distribution of the material B (black) and A (white) when Ω is a polygon in low contrast regime. The set A contains the reentrant corner.
 $m/|\Omega| \approx 34\%$.

Low contrast regime - other geometries

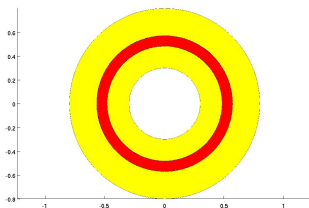


Figure : Optimal distribution of the material B (red) and A (yellow) when Ω is a ring in low contrast regime. The set B is also a ring. $m/|\Omega| \approx 17\%$.

C. Conca, A. Laurain, and R. Mahadevan (2012). “Minimization of the Ground State for Two Phase Conductors in Low Contrast Regime.”
In: *SIAM Journal on Applied Mathematics* 72.4, pp. 1238–1259

Global optimum in low contrast regime

- We want to prove that $B^* = \operatorname{argmin} \lambda_1(B)$ is also a minimizer of $\lambda^\varepsilon(B)$ for small ε .
- We have found minimizers of $\lambda_1(B)$ but not of $\lambda^\varepsilon(B)$, it was enough to disprove the conjecture.
- The minimizer $B_\varepsilon = \operatorname{argmin} \lambda^\varepsilon(B)$ does not necessarily converge as $\varepsilon \rightarrow 0$.
- If it does, B_ε does not necessarily converge to $B^* = \operatorname{argmin} \lambda_1(B)$.
- We need to prove first $B_\varepsilon \rightarrow B^*$ in an appropriate sense. The convergence of B_ε is linked to the convergence $\nabla u_\varepsilon \rightarrow \nabla u_0$. We need a convergence of ∇u_ε stronger than just L^2 .

L^∞ -convergence of the gradient

Theorem (arbitrary Ω)

For $\varepsilon > 0$ small, there exists c independent of ε and B such that

$$\|u_\varepsilon(B) - u_0\|_{H_0^1(\Omega)} \leq c\varepsilon^{\frac{1}{2}} \quad \forall B \in \mathcal{B}.$$

Theorem (case $\Omega = \mathbb{B}(0, 1)$)

Assume $\Omega = \mathbb{B}(0, 1)$ and B is radially symmetric. The functions u_ε and u_0 are in $W^{1,\infty}(\Omega)$ and there exists $\varepsilon_0 > 0$ such that for all $\varepsilon < \varepsilon_0$,

$$\|\nabla u_\varepsilon - \nabla u_0\|_{L^\infty(\Omega)} \leq c\sqrt{\varepsilon}.$$

Idea of the proof: the radial symmetry brings additional regularity, and use Hardy's inequality.

Quasi-optimal sets

Theorem

Let

$$r^* = (m/\omega_d)^{1/d}, \quad \omega_d = |\mathbb{B}(0, 1)|.$$

Let $B \subset \Omega$ be a radially symmetric measurable set and $m < \bar{m}$. For all $\delta > 0$, there exists $\varepsilon_0 = \varepsilon_0(\delta) > 0$ and B_δ^* radially symmetric and containing the origin such that for all $0 < \varepsilon \leq \varepsilon_0(\delta)$ we have

$$\lambda^\varepsilon \lambda^\varepsilon(B_\delta^*) \leq \lambda^\varepsilon(B) \quad |B_\delta^*| = m,$$

and

$$\mathbb{B}(0, r^* - \delta) \subset B_\delta^* \subset \mathbb{B}(0, r^* + \delta)$$

Idea of the proof: use $\|\nabla u_\varepsilon - \nabla u_0\|_{L^\infty(\Omega)} \leq c\sqrt{\varepsilon}$ and threshold.

Global optimum in low contrast regime

- **Goal:** prove the existence of $\varepsilon_0 > 0$ such that

$$\lambda^\varepsilon(B^*) \leq \lambda^\varepsilon(B),$$

for all $B \in \mathcal{B}$ and $\varepsilon \leq \varepsilon_0$, where $B^* = \mathbb{B}(0, r^*)$.

- **Fact:** for all $\varepsilon > 0$ there exists a $\delta(\varepsilon) > 0$ such that

$$\lambda^\varepsilon(B_{\delta(\varepsilon)}^*) \leq \lambda^\varepsilon(B)$$

holds with $\delta(\varepsilon) \rightarrow 0$ as $\varepsilon \rightarrow 0$ and $\delta(\varepsilon)$ strictly increasing.

- **We need:** the other inequality

$$\lambda^\varepsilon(B^*) \leq \lambda^\varepsilon(B_{\delta(\varepsilon)}^*).$$

- $B_{\delta(\varepsilon)}^*$ is “close” to B^* , otherwise no information.
- It is just enough to perform an asymptotic expansion of the eigenvalue with respect to $\delta(\varepsilon)$.

Global optimum in low contrast regime

- **We prove:** For all $0 < \varepsilon \leq \varepsilon_0$ and $0 < \delta \leq \delta_0$ we have

$$\lambda^\varepsilon(B^*) \leq \lambda^\varepsilon(B_\delta),$$

where B_δ is any radially symmetric set satisfying

$$\mathbb{B}(0, r^* - \delta) \subset B_\delta \subset \mathbb{B}(0, r^* + \delta).$$

- Choose $B_\delta = B_{\delta(\varepsilon)}^*$ for ε small enough

$$\lambda^\varepsilon(B^*) \leq \lambda^\varepsilon(B_\delta) = \lambda^\varepsilon(B_{\delta(\varepsilon)}^*) \leq \lambda^\varepsilon(B),$$

- **Idea of the proof:** find an expansion with $\rho(\delta) > 0$

$$\lambda^\varepsilon(B_\delta) = \lambda^\varepsilon(B^*) + \rho(\delta)\bar{\lambda}^\varepsilon + \mathcal{R}(\varepsilon, \delta) \text{ as } \rho(\delta) \rightarrow 0$$

and $\mathcal{R}(\varepsilon, \delta)/\rho(\delta) \rightarrow 0$ uniformly as $(\delta, \varepsilon) \rightarrow 0$. Prove then that $\bar{\lambda}^\varepsilon \geq 0$.

Global optimum in low contrast regime - type I

Theorem

If $m < \bar{m}$ there exists $\varepsilon_0 > 0$ such that for all $B \in \mathcal{B}$ we have

$$\lambda^\varepsilon(B^*) \leq \lambda^\varepsilon(B) \text{ for all } 0 < \varepsilon < \varepsilon_0$$

and the equality occurs only when $B = B^*$ almost everywhere in Ω .

Global optimum in low contrast regime - type II

Theorem

If $m > \bar{m}$ there exists $\varepsilon_0 > 0$ such that for all $B \in \mathcal{B}$ and for all $0 < \varepsilon < \varepsilon_0$ there exists $\xi_\varepsilon^0, \xi_\varepsilon^1$ such that

$$\lambda^\varepsilon(B_\varepsilon^*) \leq \lambda^\varepsilon(B)$$

where

$$B_\varepsilon^* = \mathbb{B}(0, \xi_\varepsilon^0) \cup \mathbb{B}(0, 1) \setminus \overline{\mathbb{B}(0, \xi_\varepsilon^1)}$$

and the equality occurs only when $B = B_\varepsilon^*$ almost everywhere in Ω . In addition we have

$$(\xi_\varepsilon^0, \xi_\varepsilon^1) \rightarrow (\xi^0, \xi^1) \text{ as } \varepsilon \rightarrow 0.$$

A. Laurain. "Global minimizer of the ground state for two phase conductors in low contrast regime." In: *ESAIM Control Optim. Calc. Var.* (To appear)

Descent Algorithm-general α, β

Variational formulation for λ

$$\lambda(B) = \min_{u \in H_0^1(\Omega)} \frac{\int_{\Omega} \sigma(B) |\nabla u|^2}{\int_{\Omega} u^2} = \min_{u \in H_0^1(\Omega), \|u\|_2=1} \int_{\Omega} \sigma(B) |\nabla u|^2.$$

Descent Algorithm

- Initial measurable set B_0 , $|B_0| = m$.
- $\mathcal{M}(B_0, c) := |\{x : |\nabla u_{B_0}(x)| \leq c\}|$.
- $c_0 := \inf\{c : \mathcal{M}(B_0, c) \geq m\}$.
- Under suitable conditions $\mathcal{M}(B_0, c_0) = m$.
- Update $B_1 = \{x : |\nabla u_{B_0}(x)| \leq c_0\}$.

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Descent Algorithm-general α, β

Variational formulation for λ

$$\lambda(B) = \min_{u \in H_0^1(\Omega)} \frac{\int_{\Omega} \sigma(B) |\nabla u|^2}{\int_{\Omega} u^2} = \min_{u \in H_0^1(\Omega), \|u\|_2=1} \int_{\Omega} \sigma(B) |\nabla u|^2.$$

Descent Algorithm

- Initial measurable set $B_0, |B_0| = m$.
- $\mathcal{M}(B_0, c) := |\{x : |\nabla u_{B_0}(x)| \leq c\}|$.
- $c_0 := \inf\{c : \mathcal{M}(B_0, c) \geq m\}$.
- Under suitable conditions $\mathcal{M}(B_0, c_0) = m$.
- Update $B_1 = \{x : |\nabla u_{B_0}(x)| \leq c_0\}$.

Descent Algorithm-general α, β

Theorem

$\lambda(B_1) \leq \lambda(B_0)$; equality holds if and only if $B_1 = B_0$ a.e. (under extra hypotheses). If B_0 is optimal, then $B_0 = \{x : |\nabla u_{B_0}(x)| \leq c_0\}$ a.e.

Corollary

- **The disk case.** $\Omega = \mathbb{B}(0, R)$, then B^* should include the origin.
- **The ring case.** The gradient of u vanishes on a circle whose center is the center of the ring. This circle is in the optimal set.
- **Domains with corners in two dimensions.** B^* contains a neighbourhood of the corners with angle smaller than π and $A^* = \Omega \setminus B^*$ contains a neighbourhood of the corners with angle greater than π .

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The Disk Case

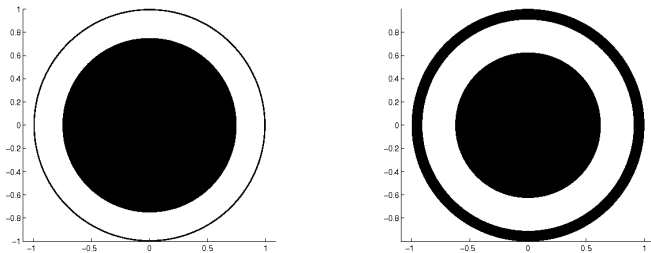


Figure : Initial domain $B_0 = \mathcal{B}(0, 0.75)$ (left). Optimal distribution (right).

The Disk Case

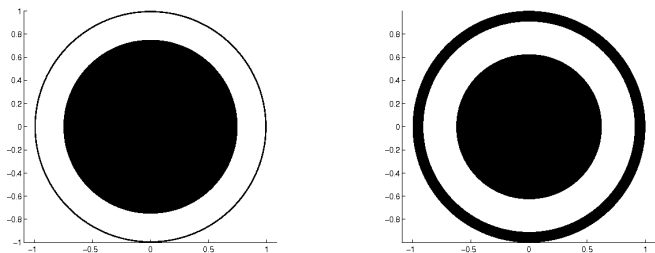


Figure : Initial domain $B_0 = \mathcal{B}(0, 0.75)$ (left). Optimal distribution (right).

Thank you for your attention !!