

Numerical Methods for a Shape Optimization Problem ¹

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Dedicated to S. Kesavan on his 60th birthday

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Problem Statement

Shape Optimization Problem.

- domain $\Omega \subset \mathbb{R}^n$, $0 < \alpha < \beta$, $0 < m < |\Omega|$
- $B \subset \Omega$ measurable; $A = \Omega \setminus B$; $|B| = m$.
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$$\inf \{ \lambda(B) : B \subset \Omega \text{ measurable, } |B| = m. \} .$$

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

- $\sigma = \alpha \chi_A + \beta \chi_B$; $\lambda(B)$ the first eigenvalue.

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Existence

- General Domains - Open
- Case of a Ball
- Alvino, Lions, Trombetti - Nonlinear Analysis 1989 [1]
- Conca, Mahadevan, Sanz - Appl. Math. Opt. 2009 [2]
- Existence of a radially symmetric solution.

Uniqueness

Open Question.

Characterization

Can we find some explicit solutions?

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- **Asymptotic Expansion.**
- Conclusions-The Disk Case.
- Numerical Results.
- A Descent Algorithm.
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Asymptotic Expansion

- Case when $\beta = \alpha + \varepsilon$, $\varepsilon > 0$ small
- $\sigma^\varepsilon = \alpha + \varepsilon\chi_B$

Theorem (Bellioh)

The first eigenvalue λ^ε of

$$-\operatorname{div}(\sigma^\varepsilon \nabla u^\varepsilon) = \lambda^\varepsilon u^\varepsilon \text{ in } \Omega, \quad (3.1)$$

$$u^\varepsilon = 0 \text{ on } \partial\Omega, \quad (3.2)$$

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

$$\int_{\Omega} (u^\varepsilon)^2 = 1 \quad (3.3)$$

is analytic with respect to ε .

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Theorem (Rellich)

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...Asymptotic Expansion

So, we can introduce the series expansion

$$u^\varepsilon = v_0 + \varepsilon v_1 + \dots, \quad (3.4)$$

$$\lambda^\varepsilon = \lambda_0 + \varepsilon \lambda_1 + \dots, \quad (3.5)$$

in equations (3.1)-(3.2) and gather terms of similar order in ε :

$$-\operatorname{div}(\alpha \nabla v_0) = \lambda_0 v_0 \quad \text{in } \Omega, \quad (3.6)$$

$$v_0 = 0 \quad \text{on } \partial\Omega. \quad (3.7)$$

$$-\operatorname{div}(\alpha \nabla v_1) - \lambda_0 v_1 = \operatorname{div}(\chi_B \nabla v_0) + \lambda_1 v_0 \quad \text{in } \Omega, \quad (3.8)$$

$$v_1 = 0 \quad \text{on } \partial\Omega. \quad (3.9)$$

Due to the Fredholm alternative, equation (3.8)-(3.9) has a solution if and only if

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

...Asymptotic Expansion

As

$$\int_{\Omega} v_0^2 = 1$$

we obtain

$$\lambda_1 = \int_B |\nabla v_0|^2. \quad (3.10)$$

Theorem (Conca, Laurain, M.)

For sufficiently small $\varepsilon > 0$

$$\operatorname{argmin}_{|B|=m} \lambda^\varepsilon(B) = \operatorname{argmin}_{|B|=m} \lambda_1(B) \quad (3.11)$$

Under additional hypotheses, the optimal solution for the problem (1.1) is of the form

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

$$\operatorname{argmin}_{|B|=m} \lambda^\varepsilon(B) = \operatorname{argmin}_{|B|=m} (\lambda_0 + \varepsilon \lambda_1(B) + \dots)$$

Conclusions - The Disk Case

- $\Omega = B(0, 1)$; 2- or 3- dimensional space.
- solution of evp (3.6)-(3.7) is radial $v_0(x) = w(|x|)$

$$r^2 w_0''(r) + (d-1)rw_0'(r) + r^2 \frac{\lambda_0}{\alpha} w_0(r) = 0, \quad (3.12)$$

$$w_0'(0) = 0, \quad w_0(1) = 0. \quad (3.13)$$

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

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

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where c^* is such that $|\{x : w_1(|x|) < c^*\}| = m$.

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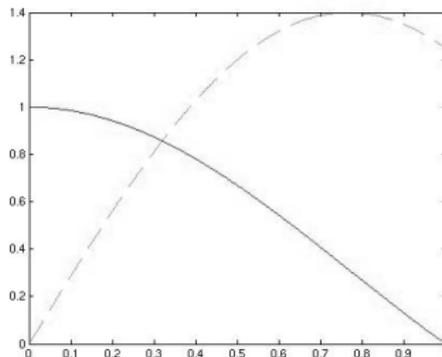
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Theorem

The solution of (1.1) is of two types. There exists m_c such that

- Type I: If $m \leq m_c$ then $B^* = B(0, (m/\pi)^{1/2})$ or,
- Type II: If $m > m_c$ then there exists ξ^0 and ξ^1 with $(m/\pi)^{1/2} < \xi^0 < \xi^1 < 1$ such that

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



Small Conductivity Gap-Other Domains

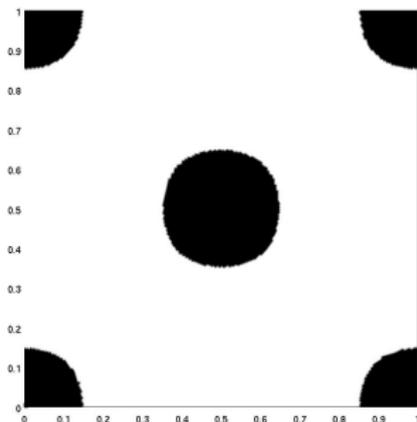


Figure: The optimal distribution in the square case.

...Small Conductivity Gap-Other Domains

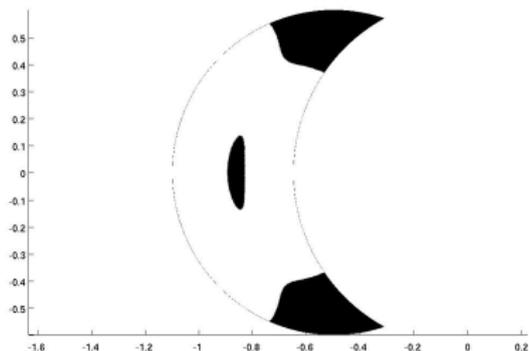


Figure: The optimal distribution in the crescent case.

...Small Conductivity Gap-Other Domains

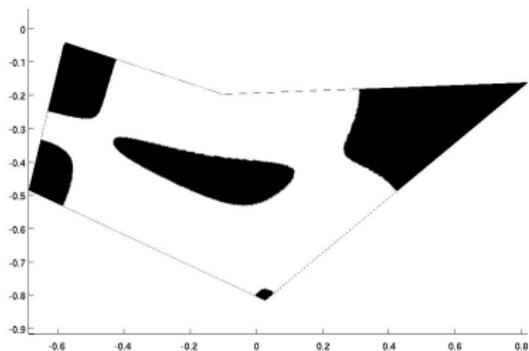


Figure: The optimal distribution in the polygon case.

...Small Conductivity Gap-Other Domains

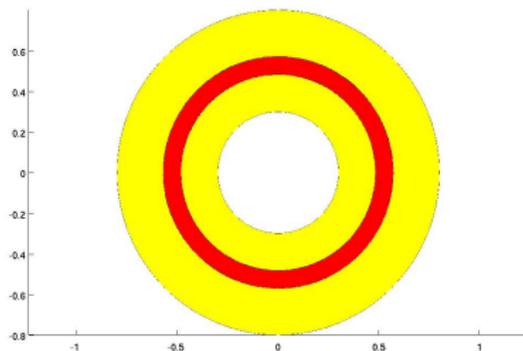


Figure: The optimal distribution in the ring case.

A Descent Algorithm-general α, β

Variational formulation for λ

$$\lambda = \min_{u \in H_0^1(\Omega)} \frac{\int_{\Omega} \sigma |\nabla u|^2}{\int_{\Omega} u^2} = \min_{u \in H_0^1(\Omega), \|u\|_2=1} \int_{\Omega} \sigma |\nabla u|^2. \quad (5.1)$$

A Descent Algorithm

- Initial measurable set B_0 , $|B_0| = m$.
- $m(B_0, c) = |\{x : |\nabla u_{B_0}(x)| \leq c\}|$. Non-decreasing $m(B_0, c) \rightarrow 0$ as $c \rightarrow 0$ whereas, $m(B_0, c) \rightarrow |\Omega|$ as $c \rightarrow \infty$.
- $c_0 := \inf\{c : m(B_0, c) \geq m\}$. (5.2)
- Under suitable conditions $|\{x : |\nabla u_{B_0}(x)| \leq c_0\}| = m$.
- Actualization $B_1 = \{x : |\nabla u_{B_0}(x)| \leq c_0\}$.

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Theorem

$\lambda(B_1) \leq \lambda(B_0)$; equality holds if and only if $B_1 = B_0$ almost everywhere (under extra hypotheses). If B_0 is optimal, then B_0 is almost everywhere equal to the level set $\{x : |\nabla u_{B_0}(x)| \leq c_0\}$.

- **The disk case.** $\Omega = B(0, R)$. The optimal set B^* should include the origin.
- **The ring or torus case.** If again we have radial symmetry, then the gradient of u vanishes at one point along a radius of the domain and by radial symmetry, the gradient of u vanishes on a whole circle whose center is the center of the ring or torus. This circle is in the optimal set.
- **Domains with corners in two dimensions.** In this case the optimal set B^* contains a neighbourhood of the corners with angle smaller than π while its complement $A^* = \Omega \setminus B^*$ contains a neighbourhood of the corners with angle greater than π .

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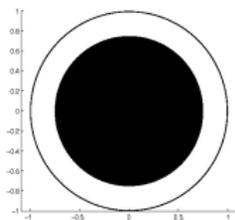


Figure: Initial domain $B_0 = B(0, 0.75)$

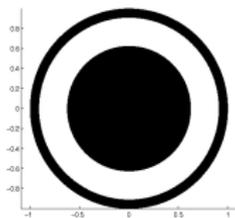


Figure: The optimal distribution in the disk case.

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