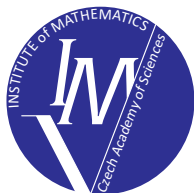


Guaranteed eigenvalue bounds for elliptic partial differential operators

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Reliable numerical methods



*To compute (approximate) solution is not sufficient.
We should provide an information about the error.*

Can we provide
a guaranteed upper bound?

$$\|u - u_h\| \leq \eta$$



Sinking of the Sleipner A off-shore platform in 1991, Norway. The failure resulted from inaccurate NASTRAN calculations.

Babuška, Verfürth, Ainsworth, Rannacher, Repin, ...



Eigenvalue problems

Laplace eigenvalue problem

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n && \text{in } \Omega \\ u_n &= 0 && \text{on } \partial\Omega \end{aligned}$$

Finite element method

- ▶ Very flexible (various domains, high order, various problems, ...)
- ▶ Converges with optimal speed
- ▶ Adaptive mesh refinement
- ▶ Nice theory

Guaranteed upper bound

$$\lambda_n \leq \lambda_{h,n}$$

Can we dream about anything else?



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Guaranteed upper bound

$$? \leq \lambda_n \leq \lambda_{h,n}$$

Can we dream about anything else? **Lower bounds!**



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Guaranteed upper bound

$$? \leq \lambda_n \leq \lambda_{h,n}$$

Can we dream about anything else? **Lower bounds!**

Guaranteed error bounds on eigenfunctions: $\|u_n - u_{h,n}\| \leq \eta$



1. Motivation
2. Theory
 - 2.1 Existence
 - 2.2 Min-max principle
 - 2.3 Optimal constants
3. Rayleigh–Ritz (Galerkin) method
4. Lower bounds on eigenvalues
 - 4.1 Weinstein's bound
 - 4.2 Lehmann–Goerisch method
 - 4.3 Interpolation constant based methods
5. Guaranteed bounds on eigenfunctions
6. Literature



2. Theory

2.1 Existence



Abstract formulation

Eigenvalue problem: Find eigenvalue λ_n and eigenfunction $u_n \in V \setminus \{0\}$ such that

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V.$$

- ▶ V is a Hilbert space.
- ▶ $a(\cdot, \cdot)$ and $b(\cdot, \cdot)$ are two bilinear forms on V .

Example

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n && \text{in } \Omega \\ u_n &= 0 && \text{on } \partial\Omega \end{aligned}$$

Weak formulation

$$(\nabla u_n, \nabla v) = \lambda_n (u_n, v) \quad \forall v \in V$$

- ▶ $V = H_0^1(\Omega)$
- ▶ $a(u, v) = (\nabla u, \nabla v)$
- ▶ $b(u, v) = (u, v)$

$$(u, v) = \int_{\Omega} uv \, dx$$

Hilbert–Schmidt theorem



$$Su_n = \mu_n u_n$$

Let

- ▶ V be a Hilbert space
- ▶ $S : V \rightarrow V$ be linear, bounded, compact, self-adjoint operator

Then

- ▶ there is (at most) countable sequence of nonzero real eigenvalues of S (repeated according to their multiplicity):
 $|\mu_1| \geq |\mu_2| \geq |\mu_3| \geq \dots > 0$,
and if the sequence is infinite then $\lim_{n \rightarrow \infty} \mu_n = 0$
- ▶ eigenfunctions u_n corresponding to these μ_n form a complete orthonormal system in \mathcal{M} and

$$V = (\ker S) \oplus \mathcal{M}$$

Note: $\mathcal{M} = \overline{\text{range } S}$



Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$: $a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$

- ▶ V is a real Hilbert space
- ▶ $a(\cdot, \cdot)$ is continuous, bilinear, symmetric, V -elliptic
- ▶ $b(\cdot, \cdot)$ is continuous, bilinear, symmetric, positive semidefinite
- ▶ $\|v\|_a = a(v, v)^{1/2}$ is the norm induced by $a(\cdot, \cdot)$
- ▶ $|v|_b = b(v, v)^{1/2}$ is the seminorm induced by $b(\cdot, \cdot)$
- ▶ $|\cdot|_b$ is **compact** with respect to $\|\cdot\|_a$,
i.e. from any sequence bounded in $\|\cdot\|_a$, we can extract a subsequence which is Cauchy in $|\cdot|_b$



Existence

Theorem. There exists (at most) countable sequence of eigenvalues

$$0 < \lambda_1 \leq \lambda_2 \leq \lambda_3 \leq \dots \rightarrow \infty$$

and the corresponding eigenfunctions can be normalized to satisfy

$$b(u_i, u_j) = \delta_{ij} \quad \forall i, j = 1, 2, \dots$$

Proof

- ▶ Solution operator $S : V \rightarrow V$: $a(Su, v) = b(u, v) \quad \forall v \in V$
- ▶ $a(u_n, v) = \lambda_n \underbrace{b(u_n, v)}_{a(Su_n, v)} \quad \forall v \in V \quad \Leftrightarrow \quad Su_n = \frac{1}{\lambda_n} u_n$
- ▶ Exercise: compactness of $|\cdot|_b$ with respect to $\|\cdot\|_a$ is equivalent to compactness of S
- ▶ Hilbert–Schmidt theorem: $\mu_1 \geq \mu_2 \geq \mu_3 \geq \dots > 0$, $\lambda_n = 1/\mu_n$ because $0 < \|u_n\|_a^2 = \lambda_n |u_n|_b^2$.



Existence

Theorem. There exists (at most) countable sequence of eigenvalues

$$0 < \lambda_1 \leq \lambda_2 \leq \lambda_3 \leq \dots \rightarrow \infty$$

and the corresponding eigenfunctions can be normalized to satisfy

$$b(u_i, u_j) = \delta_{ij} \quad \forall i, j = 1, 2, \dots$$

Note

$$\frac{1}{\lambda_i} a(u_i, u_j) = \delta_{ij} \quad \forall i, j = 1, 2, \dots$$



Orthonormal basis of eigenfunctions

Theorem. The space V can be decomposed as

$$V = \mathcal{K} \oplus \mathcal{M},$$

where $\mathcal{K} = \{v \in V : |v|_b = 0\}$ and $\mathcal{M} = \text{span}\{u_1, u_2, \dots\}$.

Moreover,

$$\begin{aligned} a(u, v) &= 0 \quad \forall u \in \mathcal{K}, \quad \forall v \in \mathcal{M}, \\ b(u, v) &= 0 \quad \forall u \in \mathcal{K}, \quad \forall v \in V. \end{aligned} \quad (*)$$

Proof

- ▶ $(*)$ follows from $|b(u, v)| \leq |u|_b |v|_b = 0$
- ▶ Hilbert–Schmidt theorem: $V = (\ker S) \oplus \mathcal{M}$
Now, $\ker S = \mathcal{K}$, because
 - (a) $u \in \mathcal{K} \Rightarrow 0 = b(u, v) = a(Su, v) \quad \forall v \in V$
 $\Rightarrow Su = 0 \Rightarrow u \in \ker S$
 - (b) $u \in \ker S \Rightarrow 0 = a(Su, u) = b(u, u) = |u|_b^2 \Rightarrow u \in \mathcal{K}$



Orthonormal basis of eigenfunctions

Theorem. The space V can be decomposed as

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where $\mathcal{K} = \{v \in V : |v|_b = 0\}$ and $\mathcal{M} = \text{span}\{u_1, u_2, \dots\}$.

Moreover,

$$\begin{aligned} a(u, v) &= 0 \quad \forall u \in \mathcal{K}, \quad \forall v \in \mathcal{M}, \\ b(u, v) &= 0 \quad \forall u \in \mathcal{K}, \quad \forall v \in V. \end{aligned} \quad (*)$$

Proof

- ▶ Express $v \in \mathcal{M}$ as $v = \sum_{n=1}^{\infty} c_n u_n$ and

$$a(u, v) = \sum_{n=1}^{\infty} c_n a(u, u_n) = \sum_{n=1}^{\infty} c_n \lambda_n b(u, u_n) \stackrel{(*)}{=} 0.$$



Parseval's identities



Theorem. For all $v \in V$, there are unique $v^{\mathcal{K}} \in \mathcal{K}$ and $v^{\mathcal{M}} \in \mathcal{M}$ such that

$$v = v^{\mathcal{K}} + v^{\mathcal{M}}, \quad v^{\mathcal{M}} = \sum_{n=1}^{\infty} c_n u_n, \quad c_n = b(v^{\mathcal{M}}, u_n) = b(v, u_n)$$

$$|v|_b^2 = \sum_{n=1}^{\infty} |b(v, u_n)|^2,$$

$$\|v\|_a^2 = \|v^{\mathcal{K}}\|_a^2 + \|v^{\mathcal{M}}\|_a^2 \quad \text{with} \quad \|v^{\mathcal{M}}\|_a^2 = \sum_{n=1}^{\infty} \lambda_n |b(v, u_n)|^2.$$

Proof

- ▶ $v = v^{\mathcal{K}} + v^{\mathcal{M}} = v^{\mathcal{K}} + \sum_{n=1}^{\infty} c_n u_n$
- ▶ $|v|_b^2 = b(v, v^{\mathcal{K}} + \sum_{n=1}^{\infty} c_n u_n) = \sum_{n=1}^{\infty} c_n b(v, u_n)$
- ▶ $\|v\|_a^2 = \|v^{\mathcal{M}}\|_a^2 + \|v^{\mathcal{K}}\|_a^2$ and $\|v^{\mathcal{M}}\|_a^2 = \sum_{n=1}^{\infty} \lambda_n c_n^2$





Example 1: Dirichlet Laplacian

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n && \text{in } \Omega \\ u_n &= 0 && \text{on } \partial\Omega \end{aligned}$$

Weak formulation: Find $\lambda_n \in \mathbb{R}$, $u_n \in H_0^1(\Omega) \setminus \{0\}$:

$$(\nabla u_n, \nabla v) = \lambda_n (Iu_n, Iv) \quad \forall v \in H_0^1(\Omega),$$

where $I : H_0^1(\Omega) \rightarrow L^2(\Omega)$ is the identity operator.

- ▶ $V = H_0^1(\Omega)$
- ▶ $a(u, v) = (\nabla u, \nabla v) \dots$ cont., bilin., sym., V -elliptic
- ▶ $b(u, v) = (Iu, Iv) \dots$ cont., bilin., sym., pos. def.
- ▶ **Compactness:** I is a compact operator by Rellich theorem.
Definition: I is compact if from a sequence $\{v_j\} \subset H_0^1(\Omega)$ bounded in $\|\nabla v\|_{L^2(\Omega)} \leq C$ we can extract a subsequence such that $\{Iv_j\}$ is Cauchy in $L^2(\Omega)$.



Example 1: Dirichlet Laplacian

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

Exact solution for an interval $\Omega = (0, L)$

$$\lambda_n = \frac{n^2 \pi^2}{L^2}, \quad u_n(x) = \sin \frac{n\pi x}{L}, \quad n = 1, 2, 3, \dots$$

Easy to verify

$$u_n'(x) = \frac{n\pi}{L} \cos \frac{n\pi x}{L}$$

$$u_n''(x) = -\frac{n^2 \pi^2}{L^2} \sin \frac{n\pi x}{L} = -\frac{n^2 \pi^2}{L^2} u_n(x)$$

Is it complete?



Example 1: Dirichlet Laplacian

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

Exact solution for a square $\Omega = (0, \pi)^2$

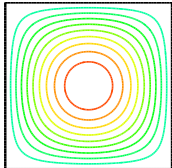
$$\lambda_{k,\ell} = k^2 + \ell^2, \quad u_{k,\ell}(x, y) = \sin(kx) \sin(\ell y), \quad k, \ell = 1, 2, \dots$$

$\lambda_1 = 2$ ($k = 1, \ell = 1$)	$\lambda_6 = 10$ ($k = 1, \ell = 3$)
$\lambda_2 = 5$ ($k = 2, \ell = 1$)	$\lambda_7 = 13$ ($k = 3, \ell = 2$)
$\lambda_3 = 5$ ($k = 1, \ell = 2$)	$\lambda_8 = 13$ ($k = 2, \ell = 3$)
$\lambda_4 = 8$ ($k = 2, \ell = 2$)	$\lambda_9 = 17$ ($k = 4, \ell = 1$)
$\lambda_5 = 10$ ($k = 3, \ell = 1$)	$\lambda_{10} = 17$ ($k = 1, \ell = 4$)

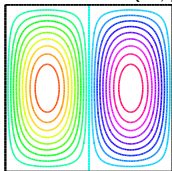


Example: Square

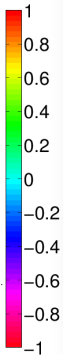
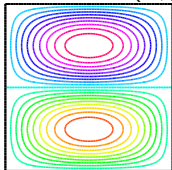
$$\lambda_1 = 2, u_1(x, y) = \sin(x) \sin(y)$$



$$\lambda_2 = 5, u_2(x, y) = \sin(2x) \sin(y)$$



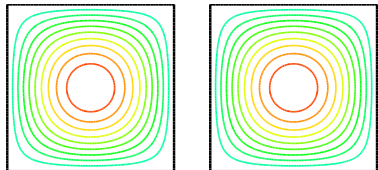
$$\lambda_3 = 5, u_3(x, y) = \sin(x) \sin(2y)$$



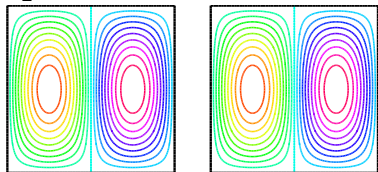


Example: Two squares

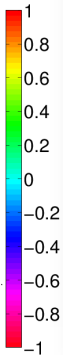
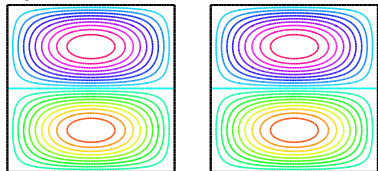
$$\lambda_1 = 2$$



$$\lambda_2 = 5$$



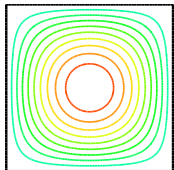
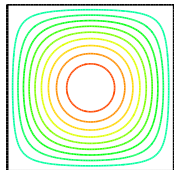
$$\lambda_3 = 5$$



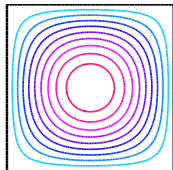
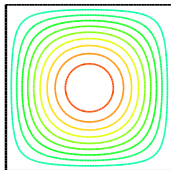


Example: Two squares

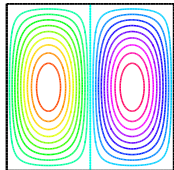
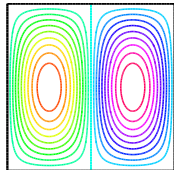
$\lambda_1 = 2$



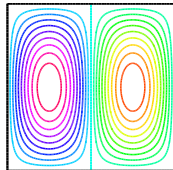
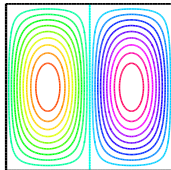
$\lambda_2 = 2$



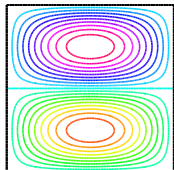
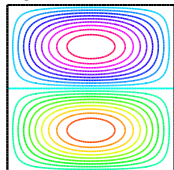
$\lambda_3 = 5$



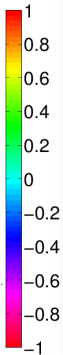
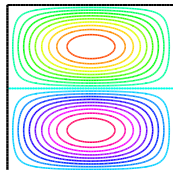
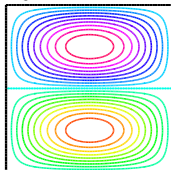
$\lambda_4 = 5$



$\lambda_5 = 5$



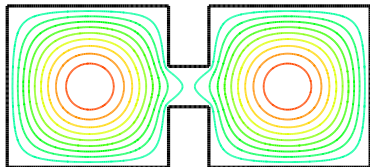
$\lambda_6 = 5$



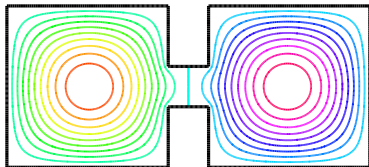
Example: Dumbbell



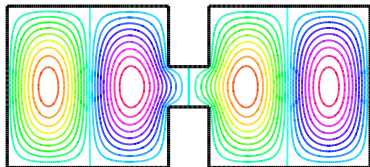
$\lambda_1 \approx 1.9558$



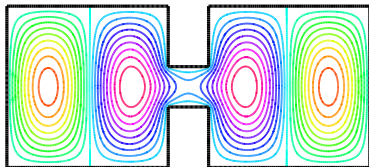
$\lambda_2 \approx 1.9607$



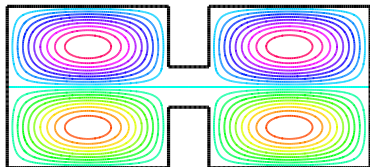
$\lambda_4 \approx 4.8299$



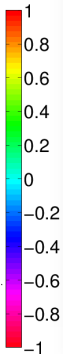
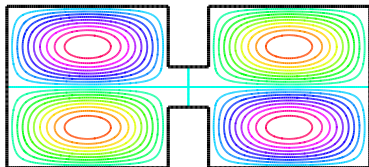
$\lambda_3 \approx 4.8008$



$\lambda_5 \approx 4.9968$



$\lambda_6 \approx 4.9968$





2. Theory

2.2 Min-max principle



Minimum principle

$$\text{Rayleigh quotient: } R(v) = \frac{a(v, v)}{b(v, v)} = \frac{\|v\|_a^2}{|v|_b^2}$$

Theorem. Numbers $0 < \lambda_1 \leq \lambda_2 \leq \dots$ and functions $u_1, u_2, \dots \in V \setminus \{0\}$ are eigenpairs of

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

if and only if

$$\lambda_1 = \min_{v \in V, |v|_b \neq 0} R(v) \quad u_1 = \arg \min_{v \in V, |v|_b \neq 0} R(v),$$

$$\lambda_n = \min_{v \in \mathcal{M}_{n-1}^\perp} R(v) \quad u_n = \arg \min_{v \in \mathcal{M}_{n-1}^\perp} R(v),$$

where $\mathcal{M}_{n-1} = \text{span}\{u_1, u_2, \dots, u_{n-1}\}$,

$$\mathcal{M}_{n-1}^\perp = \{v \in \mathcal{M} : b(v, u_i) = 0, \forall i = 1, 2, \dots, n-1\}$$

$$= \{v \in V : b(v, u_i) = 0, \forall i = 1, 2, \dots, n-1$$

and $|v|_b \neq 0\}$.



Minimum principle

Proof. (Including $n = 1$).

\Rightarrow Let $a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$.

Then $u_n \in \mathcal{M}_{n-1}^\perp$, $\lambda_n = R(u_n)$, and thus $\inf_{\mathcal{M}_{n-1}^\perp} R(v) \leq \lambda_n$.

If $v \in \mathcal{M}_{n-1}^\perp$ then $v^{\mathcal{K}} = 0$, $c_i = b(v, u_i) = 0$ for $i = 1, \dots, n-1$, and

$$R(v) = \frac{\|v\|_a^2}{|v|_b^2} = \frac{\sum_{i=n}^{\infty} \lambda_i c_i^2}{\sum_{i=n}^{\infty} c_i^2} \geq \lambda_n \frac{\sum_{i=n}^{\infty} c_i^2}{\sum_{i=n}^{\infty} c_i^2} = \lambda_n$$

\Leftarrow The minimum is attained: $\exists u_n \in \mathcal{M}_{n-1}^\perp : \lambda_n = R(u_n)$.

Let $t \in \mathbb{R}$, $v \in \mathcal{M}_{n-1}^\perp$ and $\varphi(t) = R(u_n + tv)$.

Derivative $\varphi'(0)$ exists and

$$\varphi'(0) = \frac{2}{|u_n|_b} \left(a(u_n, v) - \frac{\|u_n\|_a^2}{|u_n|_b^2} b(u_n, v) \right)$$

Since $\varphi(t)$ has a minimum at $t = 0$, we have $\varphi'(0) = 0$.

If $v = u_i$, $i = 1, 2, \dots, n-1$, then

$$b(u_n, u_i) = 0 \text{ and } a(u_n, u_i) = \lambda_i b(u_n, u_i) = 0.$$



(Courant–Fischer–Weyl) Min-max principle

Theorem.

$$\lambda_n = \min_{v \in \mathcal{M}_{n-1}^\perp} R(v) = \min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v)$$

where $\mathcal{V}^{(n)}$ is the set of all n -dimensional subspaces of \mathcal{M} .
Moreover, the minimum is attained for $E = \text{span}\{u_1, \dots, u_n\}$.

Proof. (Induction over n .)

$n = 1$: Since $R(\alpha v) = R(v)$ for all $\alpha \neq 0$, we have

$$\min_{E \in \mathcal{V}^{(1)}} \max_{v \in E} R(v) = \min_{v \in \mathcal{M}} R(v) = \min_{v \in V, |v|_b \neq 0} R(v)$$



(Courant–Fischer–Weyl) Min-max principle

Theorem.

$$\lambda_n = \min_{v \in \mathcal{M}_{n-1}^\perp} R(v) = \min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v)$$

where $\mathcal{V}^{(n)}$ is the set of all n -dimensional subspaces of \mathcal{M} .

Moreover, the minimum is attained for $E = \text{span}\{u_1, \dots, u_n\}$.

Proof. (Induction over n .)

$n > 1$: Let $\tilde{\mathcal{V}}^{(n)} \subset \mathcal{V}^{(n)}$ be a set of all spaces

$\tilde{E}^z = \text{span}\{u_1, \dots, u_{n-1}, z\}$, where $b(z, u_i) = 0$ for $i = 1, \dots, n-1$.

$$\min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v) \leq \min_{\tilde{E}^z \in \tilde{\mathcal{V}}^{(n)}} \max_{v \in \tilde{E}^z} R(v) = \min_{z \in \mathcal{M}_{n-1}^\perp} \max_{v \in \tilde{E}^z} R(v) \stackrel{(!)}{=} \min_{z \in \mathcal{M}_{n-1}^\perp} R(z)$$

To prove (!), let $v \in \tilde{E}^z$, $|v|_b = |z|_b = 1$. Thus,

$v = \alpha z + \sum_{i=1}^{n-1} c_i u_i$, $|v|_b^2 = \alpha^2 + \sum_{i=1}^{n-1} c_i^2 = 1$, and

$$R(v) = \|v\|_a^2 = \alpha^2 \|z\|_a^2 + \sum_{i=1}^{n-1} c_i^2 \|u_i\|_a^2 \leq \left(\alpha^2 + \sum_{i=1}^{n-1} c_i^2 \right) \|z\|_a^2 = R(z),$$

because $z \in \mathcal{M}_{i-1}^\perp$ for all $i = 1, 2, \dots, n-1$ and $R(u_i) \leq R(z)$.



(Courant–Fischer–Weyl) Min-max principle

Theorem.

$$\lambda_n = \min_{v \in \mathcal{M}_{n-1}^\perp} R(v) = \min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v)$$

where $\mathcal{V}^{(n)}$ is the set of all n -dimensional subspaces of \mathcal{M} .

Moreover, the minimum is attained for $E = \text{span}\{u_1, \dots, u_n\}$.

Proof. (Induction over n .)

$n > 1$: (cont'd)

Let $E \in \mathcal{V}^{(n)}$.

There exists $z \in E : |z|_b \neq 0$ and $b(z, u_i) = 0$ for $i = 1, 2, \dots, n-1$.

$$\max_{v \in E} R(v) \geq R(z) \geq \min_{z \in \mathcal{M}_{n-1}^\perp} R(z)$$





Example 2: Neumann Laplacian

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n && \text{in } \Omega \\ \frac{\partial u_n}{\partial \nu} &= 0 && \text{on } \partial\Omega \end{aligned}$$

Weak formulation: Find $\lambda_n \in \mathbb{R}$, $u_n \in H^1(\Omega) \setminus \{0\}$:

$$(\nabla u_n, \nabla v) = \lambda_n (u_n, v) \quad \forall v \in H^1(\Omega)$$

Problem: $u_0 \equiv 1$, $\lambda_0 = 0$

\Rightarrow bilinear form $a(u, v) = (\nabla u, \nabla v)$ is not $H^1(\Omega)$ -elliptic.

- ▶ $V = \{v \in H^1(\Omega) : \int_{\Omega} v = 0\}$
- ▶ $a(u, v) = (\nabla u, \nabla v) \dots$ cont., bilin., sym., V -elliptic
- ▶ $b(u, v) = (u, v) \dots$ cont., bilin., sym., pos. def.
- ▶ **Compactness:** by Rellich theorem.

Example 2: Neumann Laplacian



$$\begin{aligned} -\Delta u_n &= \lambda_n u_n && \text{in } \Omega \\ \frac{\partial u_n}{\partial \nu} &= 0 && \text{on } \partial\Omega \end{aligned}$$

Exact solution for a square $\Omega = (0, \pi)^2$

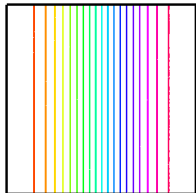
$$\lambda_{k,\ell} = k^2 + \ell^2, \quad u_{k,\ell}(x, y) = \cos(kx) \cos(\ell y), \quad k, \ell = 0, 1, 2, \dots$$

$\lambda_0 = 0$ ($k = 0, \ell = 0$)	$\lambda_5 = 4$ ($k = 0, \ell = 2$)
$\lambda_1 = 1$ ($k = 1, \ell = 0$)	$\lambda_6 = 5$ ($k = 2, \ell = 1$)
$\lambda_2 = 1$ ($k = 0, \ell = 1$)	$\lambda_7 = 5$ ($k = 1, \ell = 2$)
$\lambda_3 = 2$ ($k = 1, \ell = 1$)	$\lambda_8 = 8$ ($k = 2, \ell = 2$)
$\lambda_4 = 4$ ($k = 2, \ell = 0$)	$\lambda_9 = 9$ ($k = 3, \ell = 0$)

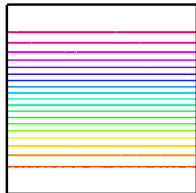
Example 2: Neumann Laplacian



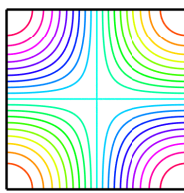
$$\lambda_1 = 1$$



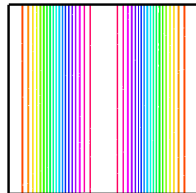
$$\lambda_2 = 1$$



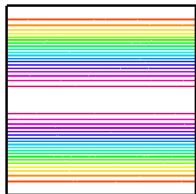
$$\lambda_3 = 2$$



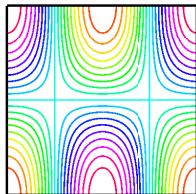
$$\lambda_4 = 4$$



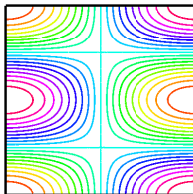
$$\lambda_5 = 4$$



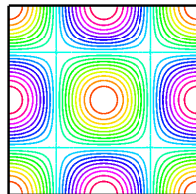
$$\lambda_6 = 5$$



$$\lambda_7 = 5$$



$$\lambda_8 = 8$$





Example 3: Steklov eigenvalue problem

$$\begin{aligned} -\Delta u_n + u_n &= 0 && \text{in } \Omega \\ \frac{\partial u_n}{\partial \nu} &= \lambda_n u_n && \text{on } \partial\Omega \end{aligned}$$

Weak formulation: Find $u_n \in H^1(\Omega)$, $\|u_n\|_{L^2(\partial\Omega)} \neq 0$, and $\lambda_n \in \mathbb{R}$:

$$(\nabla u_n, \nabla v) + (u_n, v) = \lambda_n (\gamma u_n, \gamma v)_{\partial\Omega} \quad \forall v \in H^1(\Omega)$$

- ▶ $V = H^1(\Omega)$, $V = \mathcal{K} \oplus \mathcal{M}$, $\mathcal{K} = \{v \in H^1(\Omega) : \gamma v = 0 \text{ on } \partial\Omega\}$
 $\mathcal{M} = \{v \in H^1(\Omega) : \gamma v \neq 0 \text{ on } \partial\Omega\}$
- ▶ $a(u, v) = (\nabla u, \nabla v) + (u, v) \dots$ cont., bilin., sym., V -elliptic
- ▶ $b(u, v) = (u, v)_{\partial\Omega} \dots$ cont., bilin., sym., pos. semidefinite
- ▶ **Compactness:**
Trace operator $\gamma : H^1(\Omega) \rightarrow L^2(\partial\Omega)$ is compact

[Kufner, John, Fučík 1997], [Biegert 2009]

Example 3: Steklov eigenvalue problem



$$\begin{aligned} -\Delta u_n + u_n &= 0 && \text{in } \Omega \\ \frac{\partial u_n}{\partial \nu} &= \lambda_n u_n && \text{on } \partial\Omega \end{aligned}$$

Exact solution for a square $\Omega = (-L, L)^2$

$$\lambda_1 = \frac{\sqrt{2}}{2} \tanh\left(\frac{\sqrt{2}}{2}L\right), \quad u_1(x, y) = \cosh\left(\frac{\sqrt{2}}{2}x\right) \cosh\left(\frac{\sqrt{2}}{2}y\right)$$

$$\lambda_2 = ?$$

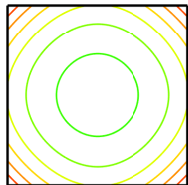
$$\lambda_3 = ?$$

$$\lambda_4 = \frac{\sqrt{2}}{2} \coth\left(\frac{\sqrt{2}}{2}L\right), \quad u_4(x, y) = \sinh\left(\frac{\sqrt{2}}{2}x\right) \sinh\left(\frac{\sqrt{2}}{2}y\right)$$

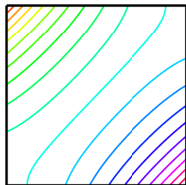
Example 3: Steklov eigenvalue problem ($L = \pi/2$)



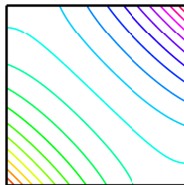
$$\lambda_1 = 0.5687$$



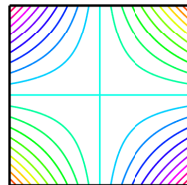
$$\lambda_2 = 0.7610$$



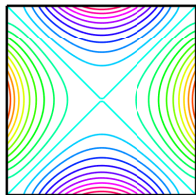
$$\lambda_3 = 0.7610$$



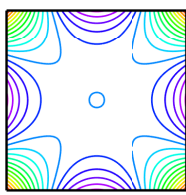
$$\lambda_4 = 0.8791$$



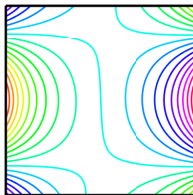
$$\lambda_5 = 1.739$$



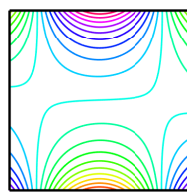
$$\lambda_6 = 1.739$$



$$\lambda_7 = 1.763$$



$$\lambda_8 = 1.763$$





2. Theory

2.3 Optimal constants



Optimal constants

Abstract eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$:

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Theorem

$$|v|_b \leq \lambda_1^{-1/2} \|v\|_a \quad \forall v \in V, \quad \text{with equality for } v = u_1.$$

Proof

Let $v \in V$, $|v|_b \neq 0$.

$$\lambda_1 = \min_{w \in V, |w|_b \neq 0} \frac{\|w\|_a^2}{|w|_b^2} \leq \frac{\|v\|_a^2}{|v|_b^2} \Leftrightarrow |v|_b^2 \leq \lambda_1^{-1} \|v\|_a^2$$

□



Optimal constants

Abstract eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$:

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Theorem

$$|v|_b \leq \lambda_1^{-1/2} \|v\|_a \quad \forall v \in V, \quad \text{with equality for } v = u_1.$$

Example 1: Dirichlet Laplacian.

$$V = H_0^1(\Omega), \quad \|v\|_a = \|\nabla v\|_{L^2(\Omega)} \quad |v|_b = \|v\|_{L^2(\Omega)}$$

Corollary 1. The optimal constant in Friedrichs inequality

$$\|v\|_{L^2(\Omega)} \leq C_F \|\nabla v\|_{L^2(\Omega)} \quad \forall v \in H_0^1(\Omega) \quad \text{is} \quad C_F = \lambda_1^{-1/2},$$

where λ_1 is the principal eigenvalue of the Dirichlet Laplacian.

$$\blacktriangleright \Omega = (0, L) \quad \Rightarrow \quad C_F = \frac{L}{\pi}$$

$$\blacktriangleright \Omega = (0, L_1) \times (0, L_2) \quad \Rightarrow \quad C_F = \frac{1}{\pi} \left(\frac{1}{L_1^2} + \frac{1}{L_2^2} \right)^{-1/2}$$



Optimal constants

Abstract eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$:

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Theorem

$$|v|_b \leq \lambda_1^{-1/2} \|v\|_a \quad \forall v \in V, \quad \text{with equality for } v = u_1.$$

Example 2: Neumann Laplacian.

$$V = \{v \in H^1(\Omega) : \int_{\Omega} v \, dx = 0\}, \quad \|v\|_a = \|\nabla v\|_{L^2(\Omega)}, \quad |v|_b = \|v\|_{L^2(\Omega)}$$

Corollary 2. The optimal constant in Poincaré inequality

$$\|v\|_{L^2(\Omega)} \leq C_P \|\nabla v\|_{L^2(\Omega)} \quad \forall v \in H^1(\Omega), \int_{\Omega} v \, dx = 0, \quad \text{is } C_P = \lambda_1^{-1/2},$$

where λ_1 is the principal eigenvalue of the Neumann Laplacian.

$$\blacktriangleright \Omega = (0, L_1) \times (0, L_2) \quad \Rightarrow \quad C_P = \frac{\max\{L_1, L_2\}}{\pi}$$



Optimal constants

Abstract eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$:

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Theorem

$$|v|_b \leq \lambda_1^{-1/2} \|v\|_a \quad \forall v \in V, \quad \text{with equality for } v = u_1.$$

Example 3: Steklov eigenvalue problem.

$$V = H^1(\Omega), \quad \|v\|_a^2 = \|\nabla v\|_{L^2(\Omega)}^2 + \|v\|_{L^2(\Omega)}^2, \quad |v|_b = \|v\|_{L^2(\partial\Omega)}$$

Corollary 3. The optimal constant in trace inequality

$$\|v\|_{L^2(\partial\Omega)} \leq C_T \|v\|_{H^1(\Omega)} \quad \forall v \in H^1(\Omega) \quad \text{is} \quad C_T = \lambda_1^{-1/2},$$

where λ_1 is the principal eigenvalue of the Steklov problem.

$$\blacktriangleright \Omega = (-L, L)^2 \quad \Rightarrow \quad C_T = (\sqrt{2} \coth(\sqrt{2}L/2))^{1/2}$$



3. Rayleigh–Ritz (Galerkin) method



Rayleigh–Ritz (Galerkin) method

Eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in V \setminus \{0\}$:

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Finite dimensional subspace: $V_h \subset V$, $\dim V_h = N < \infty$.

Discrete eigenvalue problem: Find $\lambda_{h,n} \in \mathbb{R}$, $u_{h,n} \in V_h \setminus \{0\}$:

$$a(u_{h,n}, v_h) = \lambda_{h,n} b(u_{h,n}, v_h) \quad \forall v_h \in V_h$$

Generalized eigenvalue problem for matrices:

$$A \mathbf{x}_n = \lambda_{h,n} B \mathbf{x}_n,$$

$$\text{where } u_{h,n} = \sum_{j=1}^N x_j \varphi_j, \quad A_{ij} = a(\varphi_j, \varphi_i), \quad B_{ij} = b(\varphi_j, \varphi_i)$$



Properties

Discrete eigenvalue problem: Find $\lambda_{h,n} \in \mathbb{R}$, $u_{h,n} \in V_h \setminus \{0\}$:

$$a(u_{h,n}, v_h) = \lambda_{h,n} b(u_{h,n}, v_h) \quad \forall v_h \in V_h$$

- ▶ $0 < \lambda_{h,1} \leq \lambda_{h,2} \leq \dots \leq \lambda_{h,N}$
- ▶ $\frac{1}{\lambda_{h,i}} a(u_{h,i}, u_{h,j}) = b(u_{h,i}, u_{h,j}) = \delta_{ij} \quad \forall i, j = 1, 2, \dots, N.$
- ▶ Minimum principle:

$$\lambda_{h,1} = \min_{v_h \in V_h, |v_h|_b \neq 0} R(v_h) \quad u_{h,1} = \arg \min_{v_h \in V_h, |v_h|_b \neq 0} R(v_h),$$

$$\lambda_{h,n} = \min_{v_h \in \mathcal{M}_{h,n-1}^\perp} R(v_h) \quad u_{h,n} = \arg \min_{v_h \in \mathcal{M}_{h,n-1}^\perp} R(v_h),$$

where $\mathcal{M}_{h,n-1}^\perp = \{v_h \in V_h : |v_h|_b \neq 0 \text{ and } b(v_h, u_{h,i}) = 0$
 $\forall i = 1, 2, \dots, n-1\}.$



Properties

Discrete eigenvalue problem: Find $\lambda_{h,n} \in \mathbb{R}$, $u_{h,n} \in V_h \setminus \{0\}$:

$$a(u_{h,n}, v_h) = \lambda_{h,n} b(u_{h,n}, v_h) \quad \forall v_h \in V_h$$

- ▶ $0 < \lambda_{h,1} \leq \lambda_{h,2} \leq \dots \leq \lambda_{h,N}$
- ▶ $\frac{1}{\lambda_{h,i}} a(u_{h,i}, u_{h,j}) = b(u_{h,i}, u_{h,j}) = \delta_{ij} \quad \forall i, j = 1, 2, \dots, N.$
- ▶ Min-max principle:

$$\lambda_{h,n} = \min_{E_h \in \mathcal{V}_h^{(n)}} \max_{v_h \in E_h} R(v_h)$$

where $\mathcal{V}_h^{(n)}$ is the set of all n -dimensional subspaces of V_h .

- ▶ Theorem.

$$\lambda_n \leq \lambda_{h,n}, \quad n = 1, 2, \dots, N$$

Proof.

$$\mathcal{V}_h^{(n)} \subset \mathcal{V}^{(n)} \quad \Rightarrow \quad \lambda_n = \min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v) \leq \lambda_{h,n} \quad \square$$



4. Lower bounds on eigenvalues

4.1 Weinstein's bound



Standard (conforming) approach:

Temple (1928), Weinstein (1937), Kato (1949),
Lehmann (1949), Goerisch (1985), ...

Nonconforming FEM:

Carstensen, Gedicke, Gallistl (2014), Xuefeng LIU (2015), ...

Many results: M.G. Armentano, G. Barrenechea, H. Behnke,
R.G. Duran, L. Grubišić, Jun Hu, J.R. Kuttler, Y.A. Kuznetsov,
Fubiao Lin, Qun Lin, M. Plum, S.I. Repin, V.G. Sigillito,
M. Vohralík, Hehu Xie, Yidu Yang, Zhimin Zhang, ... *many others*



Recall

Find $\lambda_n \in \mathbb{R}$ and $u_n \in V \setminus \{0\}$ such that

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

- ▶ V is a Hilbert space.
- ▶ $a(\cdot, \cdot)$ and $b(\cdot, \cdot)$ are two bilinear forms on V .
- ▶ $V = \mathcal{K} \oplus \mathcal{M}$
- ▶ $\mathcal{K} = \{v \in V : |v|_b = 0\}$
- ▶ $\mathcal{M} = \text{span}\{u_1, u_2, \dots\}$
- ▶ $v = v^{\mathcal{K}} + v^{\mathcal{M}}$
- ▶ $v^{\mathcal{M}} = \sum_{n=1}^{\infty} c_n u_n, \quad c_n = b(v^{\mathcal{M}}, u_n) = b(v, u_n)$
- ▶ $|v|_b^2 = \sum_{n=1}^{\infty} |b(v, u_n)|^2$
- ▶ $\|v\|_a^2 = \|v^{\mathcal{K}}\|_a^2 + \|v^{\mathcal{M}}\|_a^2 \quad \text{with} \quad \|v^{\mathcal{M}}\|_a^2 = \sum_{n=1}^{\infty} \lambda_n |b(v, u_n)|^2$



Weinstein's bound

Theorem

Let $\lambda_* \in \mathbb{R}$ and $u_* \in V$ with $|u_*|_b \neq 0$ be arbitrary and $w \in V$ be given by

$$a(w, v) = a(u_*, v) - \lambda_* b(u_*, v) \quad \forall v \in V.$$

Then

$$\min_j \frac{|\lambda_j - \lambda_*|^2}{\lambda_j} \leq \frac{\|w\|_a^2}{|u_*|_b^2}.$$

Proof: $w = w^{\mathcal{K}} + w^{\mathcal{M}}$

$$\begin{aligned} \|w^{\mathcal{M}}\|_a^2 &= \sum_{j=1}^{\infty} \lambda_j |b(w, u_j)|^2 = \sum_{j=1}^{\infty} \frac{|a(w, u_j)|^2}{\lambda_j} \\ &= \sum_{j=1}^{\infty} \frac{|a(u_*, u_j) - \lambda_* b(u_*, u_j)|^2}{\lambda_j} = \sum_{j=1}^{\infty} \frac{|\lambda_j - \lambda_*|^2}{\lambda_j} |b(u_*, u_j)|^2 \end{aligned}$$

Thus,

$$\|w\|_a^2 \geq \|w^{\mathcal{M}}\|_a^2 \geq \min_j \frac{|\lambda_j - \lambda_*|^2}{\lambda_j} \sum_{j=1}^{\infty} |b(u_*, u_j)|^2 \quad \square$$



Weinstein's bound

Corollary: Let λ_n has multiplicity m , i.e.,
 $\lambda_{n-1} \neq \lambda_n = \dots = \lambda_{n+m-1} \neq \lambda_{n+m}$. If

$$\sqrt{\lambda_{n-1}\lambda_n} \leq \lambda_* \leq \sqrt{\lambda_n\lambda_{n+m}} \quad (\text{closeness})$$

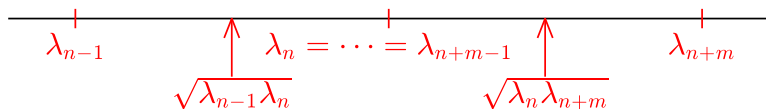
and

$$\|w\|_a \leq \eta$$

then

$$\ell_n \leq \lambda_n,$$

where $\ell_n = \frac{1}{4|u_*|_b^2} \left(-\eta + \sqrt{\eta^2 + 4\lambda_*|u_*|_b^2} \right)^2$.





Weinstein's bound

Corollary: Let λ_n has multiplicity m , i.e.,
 $\lambda_{n-1} \neq \lambda_n = \dots = \lambda_{n+m-1} \neq \lambda_{n+m}$. If

$$\sqrt{\lambda_{n-1}\lambda_n} \leq \lambda_* \leq \sqrt{\lambda_n\lambda_{n+m}} \quad (\text{closeness})$$

and

$$\|w\|_a \leq \eta$$

then

$$\ell_n \leq \lambda_n,$$

where $\ell_n = \frac{1}{4|u_*|_b^2} \left(-\eta + \sqrt{\eta^2 + 4\lambda_*|u_*|_b^2} \right)^2$.

Proof: Clearly,

$$\frac{(\lambda_n - \lambda_*)^2}{\lambda_n} = \min_j \frac{|\lambda_j - \lambda_*|^2}{\lambda_j} \leq \frac{\|w\|_a^2}{|u_*|_b^2} \leq \frac{\eta^2}{|u_*|_b^2}$$

and solve for λ_n .



Complementary upper bound on the residual

Laplace eigenvalue problem: Find λ_n and $u_n \in H_0^1(\Omega) \setminus \{0\}$:

$$(\nabla u_n, \nabla v) = \lambda_n (u_n, v) \quad \forall v \in H_0^1(\Omega)$$

Definition. Flux $\mathbf{q} \in \mathbf{H}(\text{div}, \Omega)$ is equilibrated if $-\text{div } \mathbf{q} = \lambda_* u_*$.

Theorem. If \mathbf{q} is an equilibrated flux then

$$\|\nabla w\|_0 \leq \eta = \|\nabla u_* - \mathbf{q}\|_0.$$

Proof: Let $v \in H_0^1(\Omega)$, then

$$\begin{aligned} (\nabla w, \nabla v) &= (\nabla u_*, \nabla v) - \lambda_* (u_*, v) - (\text{div } \mathbf{q}, v) - (\mathbf{q}, \nabla v) \\ &= (\nabla u_* - \mathbf{q}, \nabla v) - (\lambda_* u_* + \text{div } \mathbf{q}, v) \\ &\leq \|\nabla u_* - \mathbf{q}\|_0 \|\nabla v\|_0 \quad \square \end{aligned}$$

[Neittaanmäki, Repin 2004], [Repin 2008], [Braess, Schöberl, 2008],
[Ainsworth, Vejchodský 2011,2014], [Vohralík at al.]

Avoiding equilibration



Shifted eigenvalue problem:

$$\underbrace{(\nabla u_n, \nabla v) + \gamma(u_n, v)}_{a_\gamma(u_n, v)} = (\lambda_n + \gamma)(u_n, v) \quad \forall v \in H_0^1(\Omega)$$

Theorem. Let $\mathbf{q} \in \mathbf{H}(\text{div}, \Omega)$ and $\gamma > 0$. Then

$$\|\nabla w\|_0 \leq \|w\|_{a_\gamma} \leq \eta, \quad \eta^2 = \|\nabla u_* - \mathbf{q}\|_0^2 + \frac{1}{\gamma} \|\lambda_* u_* + \text{div } \mathbf{q}\|_0^2$$

Proof:

$$\begin{aligned} a_\gamma(w, v) &= (\nabla u_*, \nabla v) - \lambda_*(u_*, v) - (\text{div } \mathbf{q}, v) - (\mathbf{q}, \nabla v) \\ &= (\nabla u_* - \mathbf{q}, \nabla v) - (\lambda_* u_* + \text{div } \mathbf{q}, v) \\ &\leq \|\nabla u_* - \mathbf{q}\|_0 \|\nabla v\|_0 + \gamma^{-1/2} \|\lambda_* u_* + \text{div } \mathbf{q}\|_0 \gamma^{1/2} \|v\|_0 \\ &\leq (\|\nabla u_* - \mathbf{q}\|_0^2 + \gamma^{-1} \|\lambda_* u_* + \text{div } \mathbf{q}\|_0^2)^{1/2} (\|\nabla v\|_0^2 + \gamma \|v\|_0^2)^{1/2} \end{aligned}$$

Thus, $\|w\|_{a_\gamma}^2 \leq \|\nabla u_* - \mathbf{q}\|_0^2 + \gamma^{-1} \|\lambda_* u_* + \text{div } \mathbf{q}\|_0^2$ □



How to compute \mathbf{q} ?

Global flux reconstruction: Find $\mathbf{q}_h \in \mathbf{W}_h \subset \mathbf{H}(\text{div}, \Omega)$ minimizing

$$\eta^2 = \|\nabla u_* - \mathbf{q}_h\|_0^2 + \frac{1}{\gamma} \|\lambda_* u_* + \text{div } \mathbf{q}_h\|_0^2$$

FEM space:

$$V_h = \{v_h \in V : v_h|_K \in \mathbb{P}^1(K) \ \forall K \in \mathcal{T}_h\}$$

FEM approximation:

$$u_* = u_{h,n} \in V_h, \lambda_* = \lambda_{h,n}$$

Raviart–Thomas space:

$$\mathbf{RT}_1(K) = [\mathbb{P}^1(K)]^2 \oplus \mathbf{x}\mathbb{P}^1(K) \quad (\text{local})$$

$$\mathbf{W}_h = \{\mathbf{q}_h \in \mathbf{H}(\text{div}, \Omega) : \mathbf{q}_h|_K \in \mathbf{RT}_1(K) \ \forall K \in \mathcal{T}_h\} \quad (\text{global})$$



How to compute \mathbf{q} ?

Global flux reconstruction: Find $\mathbf{q}_h \in \mathbf{W}_h \subset \mathbf{H}(\text{div}, \Omega)$ minimizing

$$\eta^2 = \|\nabla u_* - \mathbf{q}_h\|_0^2 + \frac{1}{\gamma} \|\lambda_* u_* + \text{div } \mathbf{q}_h\|_0^2$$

Euler–Lagrange equations:

$$(\mathbf{q}_h, \mathbf{w}_h) + \frac{1}{\gamma} (\text{div } \mathbf{q}_h, \text{div } \mathbf{w}_h) = (\nabla u_*, \mathbf{w}_h) - \frac{1}{\gamma} (\lambda_* u_*, \text{div } \mathbf{w}_h)$$

$$\forall \mathbf{w}_h \in \mathbf{W}_h$$

Equivalent to linear system:

$$M\mathbf{y} = F,$$

$$\text{where } \mathbf{q}_h = \sum_j y_j \boldsymbol{\psi}_j, \quad M_{ij} = (\boldsymbol{\psi}_j, \boldsymbol{\psi}_i) + \frac{1}{\gamma} (\text{div } \boldsymbol{\psi}_j, \text{div } \boldsymbol{\psi}_i),$$
$$F_i = (\nabla u_*, \boldsymbol{\psi}_i) - \frac{1}{\gamma} (\lambda_* u_*, \text{div } \boldsymbol{\psi}_i)$$

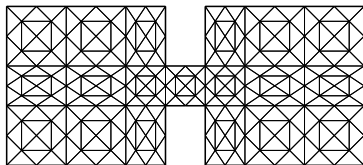


Example: dumbbell

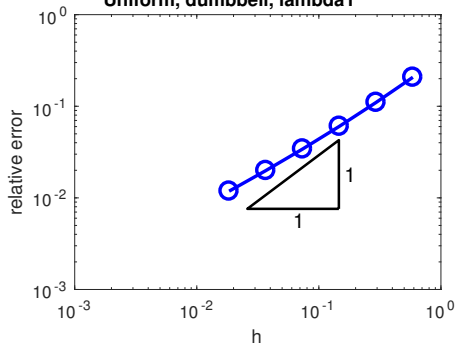
$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega = \text{dumbbell} \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\text{Rel. error: } \frac{|\lambda_n - \lambda_{h,n}|}{\lambda_n} \leq \frac{\lambda_{h,n} - \ell_n}{\ell_n}$$

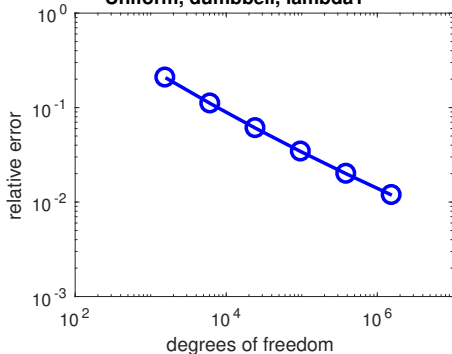
$$\gamma = 10^{-6}$$



Uniform, dumbbell, lambda1



Uniform, dumbbell, lambda1

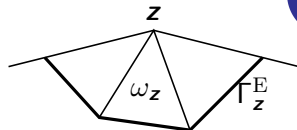




Local flux reconstruction

Flux reconstruction:

$$\mathbf{q}_h = \sum_{z \in \mathcal{N}_h} \mathbf{q}_z$$



Local problems: Find $\mathbf{q}_z \in \mathbf{W}_z$ minimizing

$$\|\varphi_z \nabla u_* - \mathbf{q}_z\|_{L^2(\omega_z)}^2 + \frac{1}{\gamma} \|\lambda_* \varphi_z u_* + \operatorname{div} \mathbf{q}_z\|_{L^2(\omega_z)}^2$$

Euler–Lagrange equations:

$$(\mathbf{q}_z, \mathbf{w}_h)_{\omega_z} + \frac{1}{\gamma} (\operatorname{div} \mathbf{q}_z, \operatorname{div} \mathbf{w}_h)_{\omega_z} = (\varphi_z \nabla u_*, \mathbf{w}_h)_{\omega_z} - \frac{1}{\gamma} (\lambda_* \varphi_z u_*, \operatorname{div} \mathbf{w}_h)_{\omega_z} \quad \forall \mathbf{w}_h \in \mathbf{W}_z$$

Patch of elements: $\omega_z = \bigcup \{K \in \mathcal{T}_h : z \in K\}$

Partition of unity: $\sum_{z \in \mathcal{N}_h} \varphi_z = 1$

$\mathbf{W}_z = \{\mathbf{q} \in \mathbf{H}(\operatorname{div}, \omega_z) : \mathbf{q}|_K \in \mathbf{RT}_1(K) \quad \forall K \subset \omega_z, \mathbf{q} \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^E\}$

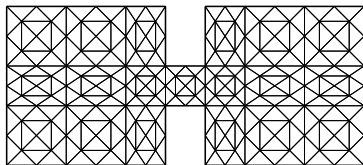


Example: dumbbell

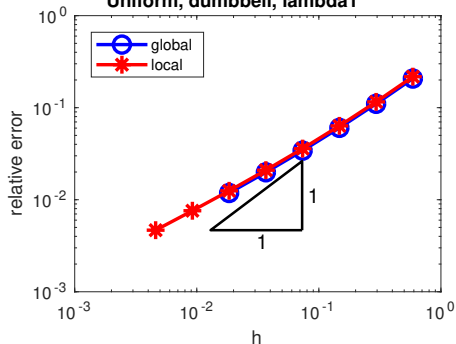
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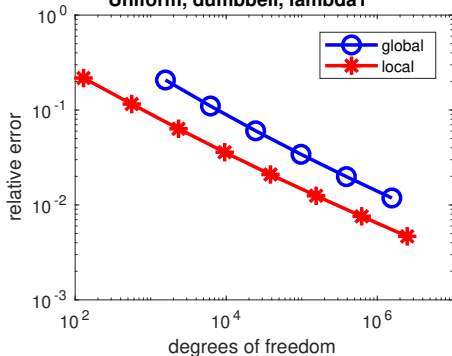
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Uniform, dumbbell, lambda1



Uniform, dumbbell, lambda1

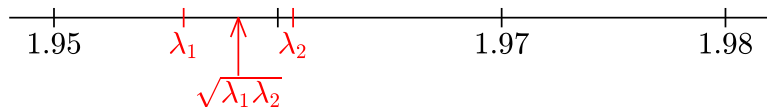




Closeness assumption for dumbbell

$$\sqrt{\lambda_{n-1}\lambda_n} \leq \lambda_* \leq \sqrt{\lambda_n\lambda_{n+m}} \Rightarrow \ell_n \leq \lambda_n$$

Exact eigenvalues: $\lambda_1 = 1.955793794588$, $\lambda_2 = 1.960683031595$



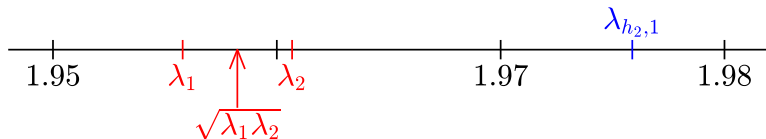
h	ℓ_1	$\lambda_{h,1}$	closeness
$h_1 = 1.1781$	1.6618	2.0228	no



Closeness assumption for dumbbell

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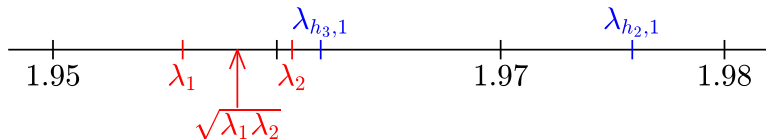
h	ℓ_1	$\lambda_{h,1}$	closeness
$h_1 = 1.1781$	1.6618	2.0228	no
$h_2 = 0.5890$	1.7711	1.9759	no



Closeness assumption for dumbbell

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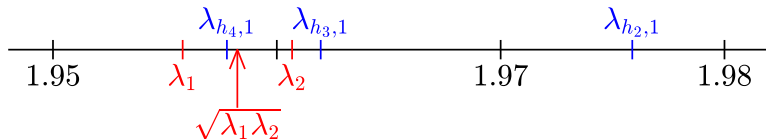
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$h_3 = 0.2945$	1.8449	1.9620	no



Closeness assumption for dumbbell

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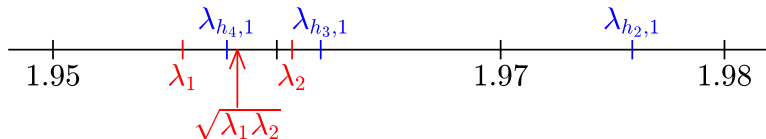
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$h_2 = 0.5890$	1.7711	1.9759	no
$h_3 = 0.2945$	1.8449	1.9620	no
$h_4 = 0.1473$	1.8899	1.9578	yes



Closeness assumption for dumbbell

$$\sqrt{\lambda_{n-1}\lambda_n} \leq \lambda_* \leq \sqrt{\lambda_n\lambda_{n+m}} \Rightarrow \ell_n \leq \lambda_n$$

Exact eigenvalues: $\lambda_1 = 1.955793794588$, $\lambda_2 = 1.960683031595$



h	ℓ_1	$\lambda_{h,1}$	closeness
$h_1 = 1.1781$	1.6618	2.0228	no
$h_2 = 0.5890$	1.7711	1.9759	no
$h_3 = 0.2945$	1.8449	1.9620	no
$h_4 = 0.1473$	1.8899	1.9578	yes
$h_5 = 0.0736$	1.9163	1.9565	yes
$h_6 = 0.0368$	1.9319	1.9560	yes
$h_7 = 0.0184$	1.9411	1.9559	yes

Weinstein's bound – summary



- ▶ easy to use
- ▶ it is a generalization of Bauer–Fike estimates for matrices
- ▶ good for general symmetric elliptic problems
- ▶ sub-optimal speed of convergence
- ▶ a priori information on spectrum needed for guaranteed lower bounds



4. Lower bounds on eigenvalues

4.2 Lehmann–Goerisch method



General setting:

Find $\lambda_n \in \mathbb{R}$ and $u_n \in V \setminus \{0\}$ such that

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Lehmann method



Theorem

Let $\tilde{\lambda}_N < \rho \leq \lambda_{N+1}$

- ▶ $\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_N \in V$ be linearly independent
- ▶ $A_{0,ij} = a(\tilde{u}_i, \tilde{u}_j)$
- ▶ $A_{1,ij} = b(\tilde{u}_i, \tilde{u}_j)$
- ▶ $w_i \in V : a(w_i, v) = b(\tilde{u}_i, v) \quad \forall v \in V$
 $A_{2,ij} = a(w_i, w_j)$

- ▶ $\mu_1 \leq \mu_2 \leq \dots \leq \mu_N : (\rho A_1 - A_0)\mathbf{x} = \mu(A_0 - 2\rho A_1 + \rho^2 A_2)\mathbf{x}$

Then $0 < \mu_1$ and

$$\rho - \frac{\rho}{1 + \mu_n} \leq \lambda_n, \quad n = 1, 2, \dots, N.$$

[Lehmann 1949, 1950]



Theorem

Let $\tilde{\lambda}_N < \rho \leq \lambda_{N+1}$

▶ $\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_N \in V$ be linearly independent

▶ $A_{0,ij} = a(\tilde{u}_i, \tilde{u}_j)$

▶ $A_{1,ij} = b(\tilde{u}_i, \tilde{u}_j)$

▶ X ... vector space

\mathcal{B} ... positive semidefinite symmetric bilinear form on X

$T : V \rightarrow X$... linear operator:

(a) $\mathcal{B}(Tu, Tv) = a(u, v) \quad \forall u, v \in V$

(b) $\hat{\mathbf{w}}_i \in X : \mathcal{B}(\hat{\mathbf{w}}_i, Tv) = b(\tilde{u}_i, v) \quad \forall v \in V$

(c) $\hat{A}_{2,ij} = \mathcal{B}(\hat{\mathbf{w}}_i, \hat{\mathbf{w}}_j)$

▶ $\hat{\mu}_1 \leq \hat{\mu}_2 \leq \dots \leq \hat{\mu}_N : (\rho A_1 - A_0)\hat{\mathbf{x}} = \hat{\mu}(A_0 - 2\rho A_1 + \rho^2 \hat{A}_2)\hat{\mathbf{x}}$

Then $0 < \hat{\mu}_1$ and

$$\rho - \frac{\rho}{1 + \hat{\mu}_n} \leq \lambda_n, \quad n = 1, 2, \dots, N.$$

Proof: Lehmann \Rightarrow Goerisch



It suffices to show that $\hat{A}_2 - A_2$ is positive semidefinite, because

$$\Rightarrow 0 < \hat{\mu}_i \leq \mu_i \text{ for all } i = 1, 2, \dots, N,$$

$$\Rightarrow \rho - \frac{\rho}{1 + \hat{\mu}_n} \leq \rho - \frac{\rho}{1 + \mu_n} \leq \lambda_n.$$

To show that $\hat{A}_2 - A_2$ is positive semidefinite:

Let $\mathbf{x} \in \mathbb{R}^N$, $\tilde{u} = \sum_{i=1}^N x_i \tilde{u}_i$, $w = \sum_{i=1}^N x_i w_i$, $\hat{\mathbf{w}} = \sum_{i=1}^N x_i \hat{\mathbf{w}}_i$, and

$$0 \leq \mathcal{B}(\hat{\mathbf{w}} - Tw, \hat{\mathbf{w}} - Tw) = \mathcal{B}(\hat{\mathbf{w}}, \hat{\mathbf{w}}) - 2 \underbrace{\mathcal{B}(\hat{\mathbf{w}}, Tw)}_{\stackrel{(b)}{=} b(\tilde{u}, w)} + \underbrace{\mathcal{B}(Tw, Tw)}_{\stackrel{(a)}{=} a(w, w)}.$$
$$= a(w, w)$$

Thus,

$$0 \leq \mathcal{B}(\hat{\mathbf{w}}, \hat{\mathbf{w}}) - a(w, w) \stackrel{(c)}{=} \mathbf{x}^T (\hat{A}_2 - A_2) \mathbf{x}.$$



Application to Laplace eigenvalue problem



$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega), \quad \Omega \subset \mathbb{R}^2$$

Setting

- ▶ $V = H_0^1(\Omega)$, $a(u, v) = (\nabla u, \nabla v) + \gamma(u, v)$, $b(u, v) = (u, v)$
- ▶ $X = [L^2(\Omega)]^3$
- ▶ $\mathcal{B}(\hat{u}, \hat{v}) = (\hat{u}_1, \hat{v}_1) + (\hat{u}_2, \hat{v}_2) + \gamma(\hat{u}_3, \hat{v}_3)$
- ▶ $Tu = \begin{pmatrix} \nabla u \\ u \end{pmatrix}$

Facts

(a) $\mathcal{B}(Tu, Tv) = a(u, v)$

Application to Laplace eigenvalue problem



$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega), \quad \Omega \subset \mathbb{R}^2$$

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- ▶ $Tu = \begin{pmatrix} \nabla u \\ u \end{pmatrix}$

Facts

$$(a) \quad \mathcal{B}(Tu, Tv) = a(u, v)$$

$$(b) \quad \mathcal{B}(\hat{w}_i, Tv) = b(\tilde{u}_i, v) \iff \hat{w}_i = \begin{pmatrix} \sigma_i \\ \hat{w}_{i,3} \end{pmatrix} \quad \sigma_i \in \mathbf{H}(\text{div}, \Omega)$$

$$(\sigma_i, \nabla v) + \gamma(\hat{w}_{i,3}, v) = (\tilde{u}_i, v) \quad \forall v \in V$$

$$-(\text{div } \sigma_i, v) + \gamma(\hat{w}_{i,3}, v) = (\tilde{u}_i, v) \quad \forall v \in V$$

$$\hat{w}_{i,3} = \frac{1}{\gamma}(\tilde{u}_i + \text{div } \sigma_i)$$

Application to Laplace eigenvalue problem



$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega), \quad \Omega \subset \mathbb{R}^2$$

Setting

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- ▶ $X = [L^2(\Omega)]^3$
- ▶ $\mathcal{B}(\hat{u}, \hat{v}) = (\hat{u}_1, \hat{v}_1) + (\hat{u}_2, \hat{v}_2) + \gamma(\hat{u}_3, \hat{v}_3)$
- ▶ $Tu = \begin{pmatrix} \nabla u \\ u \end{pmatrix}$

Facts

(a) $\mathcal{B}(Tu, Tv) = a(u, v)$

(b) $\mathcal{B}(\hat{w}_i, Tv) = b(\tilde{u}_i, v) \iff \hat{w}_i = \begin{pmatrix} \sigma_i \\ \frac{1}{\gamma}(\tilde{u}_i + \operatorname{div} \sigma_i) \end{pmatrix} \quad \sigma_i \in \mathbf{H}(\operatorname{div}, \Omega)$



$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega), \quad \Omega \subset \mathbb{R}^2$$

Setting

- ▶ $V = H_0^1(\Omega)$, $a(u, v) = (\nabla u, \nabla v) + \gamma(u, v)$, $b(u, v) = (u, v)$
- ▶ $X = [L^2(\Omega)]^3$
- ▶ $\mathcal{B}(\hat{u}, \hat{v}) = (\hat{u}_1, \hat{v}_1) + (\hat{u}_2, \hat{v}_2) + \gamma(\hat{u}_3, \hat{v}_3)$
- ▶ $Tu = \begin{pmatrix} \nabla u \\ u \end{pmatrix}$

Facts

(a) $\mathcal{B}(Tu, Tv) = a(u, v)$

(b) $\mathcal{B}(\hat{w}_i, Tv) = b(\tilde{u}_i, v) \iff \hat{w}_i = \begin{pmatrix} \sigma_i \\ \frac{1}{\gamma}(\tilde{u}_i + \operatorname{div} \sigma_i) \end{pmatrix} \quad \sigma_i \in \mathbf{H}(\operatorname{div}, \Omega)$

(c) $\hat{A}_{2,ij} = \mathcal{B}(\hat{w}_i, \hat{w}_j) \iff \hat{A}_{2,ij} = (\sigma_i, \sigma_j) + \frac{1}{\gamma}(\tilde{u}_i + \operatorname{div} \sigma_i, \tilde{u}_j + \operatorname{div} \sigma_j)$

Application to Laplace eigenvalue problem

Theorem (Lehmann–Goerisch)

Let $\tilde{\lambda}_N + \gamma < \rho \leq \lambda_{N+1} + \gamma$, $\gamma > 0$

- ▶ $\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_N \in V$ be linearly independent
 - ▶ $A_{0,ij} = (\nabla \tilde{u}_i, \nabla \tilde{u}_j) + \gamma(\tilde{u}_i, \tilde{u}_j)$
 - ▶ $A_{1,ij} = (\tilde{u}_i, \tilde{u}_j)$
 - ▶ $\sigma_1, \sigma_2, \dots, \sigma_N \in \mathbf{H}(\text{div}, \Omega)$ be arbitrary
- $$\hat{A}_{2,ij} = (\sigma_i, \sigma_j) + \frac{1}{\gamma}(\tilde{u}_i + \text{div } \sigma_i, \tilde{u}_j + \text{div } \sigma_j)$$

- ▶ $\hat{\mu}_1 \leq \hat{\mu}_2 \leq \dots \leq \hat{\mu}_N$: $(\rho A_1 - A_0)\hat{\mathbf{x}} = \hat{\mu}(A_0 - 2\rho A_1 + \rho^2 \hat{A}_2)\hat{\mathbf{x}}$

Then $0 < \hat{\mu}_1$ and

$$\ell_n = \rho - \gamma - \frac{\rho}{1 + \hat{\mu}_n} \leq \lambda_n, \quad n = 1, 2, \dots, N$$



How to find good $\hat{\mathbf{w}}_i$?

Observation: Let $\tilde{u}_i \approx u_i$ and $\tilde{\lambda}_i \approx \lambda_i$.

$$\Rightarrow a(w_i, v) = b(\tilde{u}_i, v) \approx \frac{1}{\tilde{\lambda}_i} a(\tilde{u}_i, v) \quad \forall v \in V$$

$$\Rightarrow w_i \approx \frac{1}{\tilde{\lambda}_i} \tilde{u}_i$$

Need

$$\Rightarrow \hat{A}_2 \approx A_2$$

$$\Rightarrow \mathcal{B}(\hat{\mathbf{w}}_i, \hat{\mathbf{w}}_j) \approx a(w_i, w_j) \stackrel{(a)}{=} \mathcal{B}(Tw_i, Tw_j)$$

$$\Rightarrow \hat{\mathbf{w}}_i \approx Tw_i \approx \frac{1}{\tilde{\lambda}_i} T\tilde{u}_i$$

Natural idea

make $\left| \frac{1}{\tilde{\lambda}_i} T\tilde{u}_i - \hat{\mathbf{w}}_i \right|_{\mathcal{B}}^2$ small

For Laplacian: Find $\sigma_{h,i} \in \mathbf{H}(\text{div}, \Omega)$ that

makes $\left\| \frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma} - \sigma_{h,i} \right\|_0^2 + \frac{1}{\gamma} \left\| \frac{\lambda_{h,i} u_{h,i}}{\lambda_{h,i} + \gamma} + \text{div } \sigma_{h,i} \right\|_0^2$ small



Choice of σ_i – global

Global minimization:

Find $\sigma_{h,i} \in \mathbf{W}_h \subset \mathbf{H}(\text{div}, \Omega)$, $i = 1, 2, \dots, N$, that minimizes

$$\left\| \frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma} - \sigma_{h,i} \right\|_0^2 + \frac{1}{\gamma} \left\| \frac{\lambda_{h,i} u_{h,i}}{\lambda_{h,i} + \gamma} + \text{div } \sigma_{h,i} \right\|_0^2$$

Euler–Lagrange equations:

$$(\sigma_{h,i}, \mathbf{w}_h) + \frac{1}{\gamma} (\text{div } \sigma_{h,i}, \text{div } \mathbf{w}_h) = \left(\frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma}, \mathbf{w}_h \right) - \frac{1}{\gamma} \left(\frac{\lambda_{h,i} u_{h,i}}{\lambda_{h,i} + \gamma}, \text{div } \mathbf{w}_h \right)$$

$$\forall \mathbf{w}_h \in \mathbf{W}_h$$

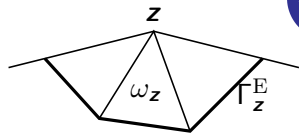
$$\mathbf{W}_h = \{ \sigma_h \in \mathbf{H}(\text{div}, \Omega) : \sigma_h|_K \in \mathbf{RT}_1(K) \quad \forall K \in \mathcal{T}_h \}$$



Choice of σ_j – local

Flux reconstruction:

$$\sigma_{h,i} = \sum_{z \in \mathcal{N}_h} \sigma_{z,i}$$



Local problems: Find $\sigma_{z,i} \in \mathbf{W}_z$, $i = 1, 2, \dots, N$ minimizing

$$\left\| \varphi_z \frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma} - \sigma_{z,i} \right\|_{0, \omega_z}^2 + \frac{1}{\gamma} \left\| \frac{\lambda_{h,i} \varphi_z u_{h,i}}{\lambda_{h,i} + \gamma} + \operatorname{div} \sigma_{z,i} \right\|_{0, \omega_z}^2$$

Euler–Lagrange equations:

$$\begin{aligned} & (\sigma_{z,i}, \mathbf{w}_h)_{\omega_z} + \frac{1}{\gamma} (\operatorname{div} \sigma_{z,i}, \operatorname{div} \mathbf{w}_h)_{\omega_z} \\ &= \left(\varphi_z \frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma}, \mathbf{w}_h \right)_{\omega_z} - \frac{1}{\gamma} \left(\frac{\varphi_z \lambda_{h,i} u_{h,i}}{\lambda_{h,i} + \gamma}, \operatorname{div} \mathbf{w}_h \right)_{\omega_z} \quad \forall \mathbf{w}_h \in \mathbf{W}_z \end{aligned}$$

Patch of elements: $\omega_z = \bigcup \{K \in \mathcal{T}_h : z \in K\}$

Partition of unity: $\sum_{z \in \mathcal{N}_h} \varphi_z = 1$

$\mathbf{W}_z = \{ \sigma \in \mathbf{H}(\operatorname{div}, \omega_z) : \sigma|_K \in \mathbf{RT}_1(K) \forall K \subset \omega_z, \sigma \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^E \}$

Comparison of flux reconstructions



Weinstein: Find $\mathbf{q}_{h,i} \in \mathbf{W}_h$ minimizing

$$\|\nabla u_{h,i} - \mathbf{q}_{h,i}\|_0^2 + \frac{1}{\gamma} \|\lambda_{h,i} u_{h,i} + \operatorname{div} \mathbf{q}_{h,i}\|_0^2$$

Lehmann–Goerisch: Find $\boldsymbol{\sigma}_{h,i} \in \mathbf{W}_h$ minimizing

$$\left\| \frac{\nabla u_{h,i}}{\lambda_{h,i} + \gamma} - \boldsymbol{\sigma}_{h,i} \right\|_0^2 + \frac{1}{\gamma} \left\| \frac{\lambda_{h,i} u_{h,i}}{\lambda_{h,i} + \gamma} + \operatorname{div} \boldsymbol{\sigma}_{h,i} \right\|_0^2$$

Thus,

$$\boldsymbol{\sigma}_{h,i} = \frac{\mathbf{q}_{h,i}}{\lambda_{h,i} + \gamma}$$

[Vejchodský 2018]

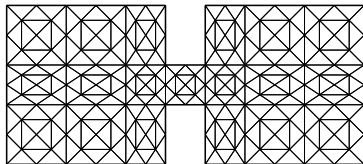


Example: dumbbell

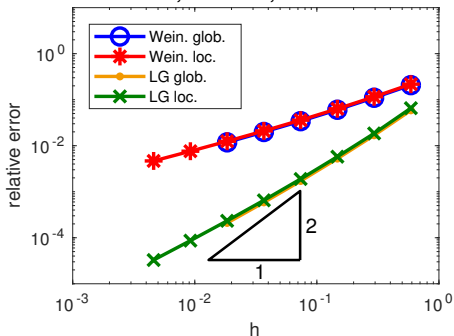
$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega = \text{dumbbell} \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\text{Rel. error: } \frac{|\lambda_n - \lambda_{h,n}|}{\lambda_n} \leq \frac{\lambda_{h,n} - \ell_n}{\ell_n}$$

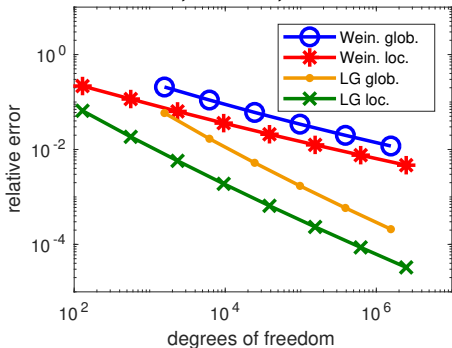
$$\gamma = 10^{-6}$$



Uniform, dumbbell, lambda1



Uniform, dumbbell, lambda1





How to get the a priori lower bound ρ ?

Monotonicity principle: If $V \subset \tilde{V}$ then $\mathcal{V}^{(n)} \subset \tilde{\mathcal{V}}^{(n)}$ and

$$\tilde{\lambda}_n = \min_{E \in \tilde{\mathcal{V}}^{(n)}} \max_{v \in E} R(v) \leq \min_{E \in \mathcal{V}^{(n)}} \max_{v \in E} R(v) = \lambda_n$$

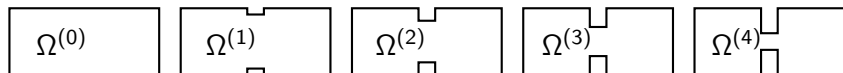
Example 1.

$$\Omega \subset \tilde{\Omega} \Rightarrow H_0^1(\Omega) \subset H_0^1(\tilde{\Omega}) \Rightarrow \tilde{\lambda}_n \leq \lambda_n$$

Example 2.

$$H_0^1(\Omega) \subset H^1(\Omega) \Rightarrow \lambda_n^{\text{Neumann}} \leq \lambda_n^{\text{Dirichlet}}$$

Homotopy



Analytically:	$\rho = 12.16$	$\rho = 11.39$	$\rho = 10.77$	$\rho = 9.988$
$12.16 \leq \lambda_{17}^{(0)}$	$\ell_{15} \doteq 11.39$	$\ell_{13} \doteq 10.77$	$\ell_{11} \doteq 9.988$	

[Plum 1990, 1991]

Adaptive mesh refinement



Recall the residual

$$w \in V : \quad (\nabla w, \nabla v) = (\nabla u_{h,i}, \nabla v) - \lambda_{h,i}(u_{h,i}, v) \quad \forall v \in V$$

Recall theorem:

$$\|\nabla w\|_0 \leq \eta, \quad \text{where } \eta^2 = \|\nabla u_{h,i} - \mathbf{q}_{h,i}\|_{L^2(\Omega)}^2 + \frac{1}{\gamma} \|\lambda_{h,i} u_{h,i} + \operatorname{div} \mathbf{q}_{h,i}\|_{L^2(\Omega)}^2$$

Local error indicators for mesh refinement:

$$\eta_K^2 = \|\nabla u_{h,i} - \mathbf{q}_{h,i}\|_{L^2(K)}^2 + \frac{1}{\gamma} \|\lambda_{h,i} u_{h,i} + \operatorname{div} \mathbf{q}_{h,i}\|_{L^2(K)}^2$$

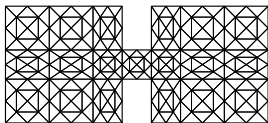
Note: Good for both Weinstein and Lehmann–Goerisch method:

$$\sigma_{h,i} = \frac{\mathbf{q}_{h,i}}{\lambda_{h,i} + \gamma}$$

Example: dumbbell

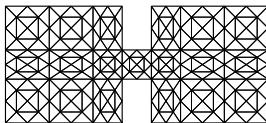


$$\begin{aligned} -\Delta u_j &= \lambda_j u_j && \text{in } \Omega \\ u_j &= 0 && \text{on } \partial\Omega \end{aligned}$$

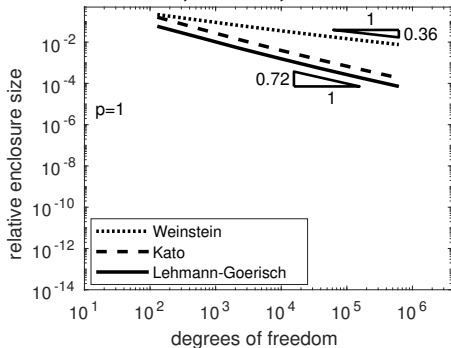


Example: dumbbell

$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



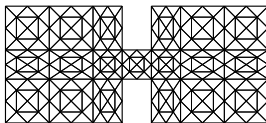
Uniform, dumbbell, lambda1



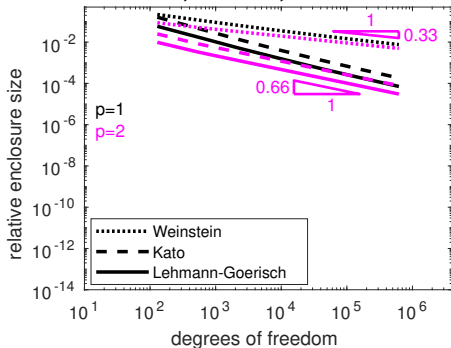
- ▶ relative enclosure size: $(\lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, $\rho := \ell_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell

$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



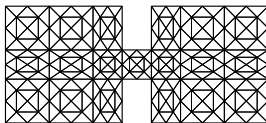
Uniform, dumbbell, lambda1



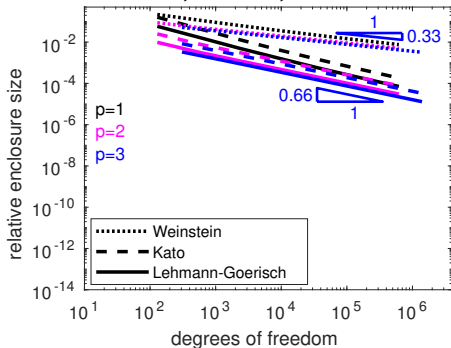
- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell

$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



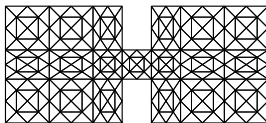
Uniform, dumbbell, lambda1



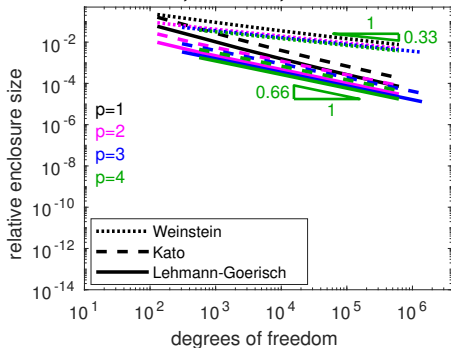
- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell

$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



Uniform, dumbbell, lambda1

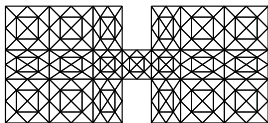


- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

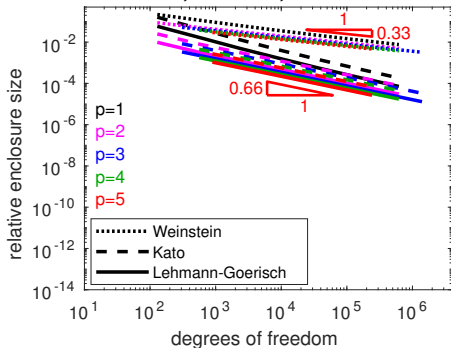
Example: dumbbell

$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1

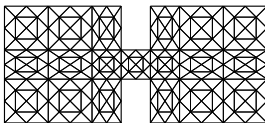


- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

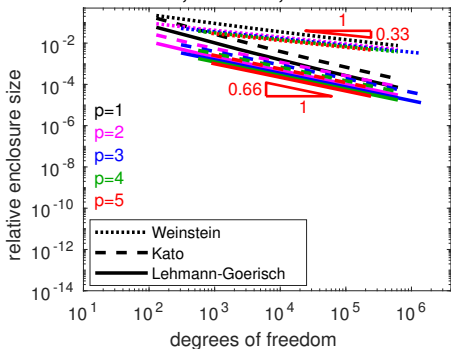
Example: dumbbell

$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

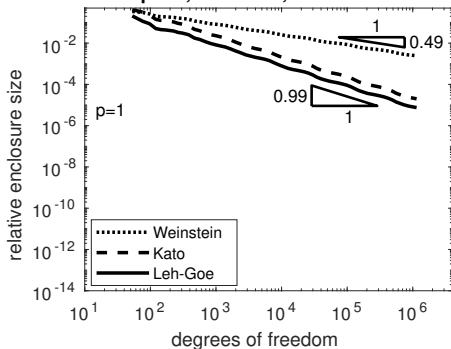
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

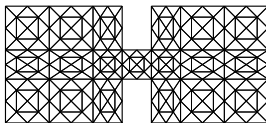


- ▶ relative enclosure size: $(\lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, $\rho := \ell_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

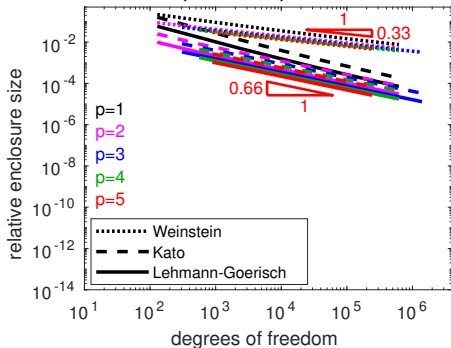
Example: dumbbell

$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

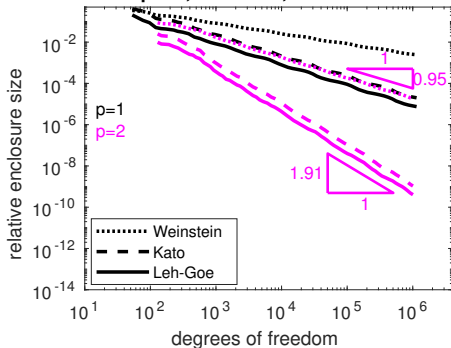
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

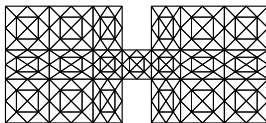


- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

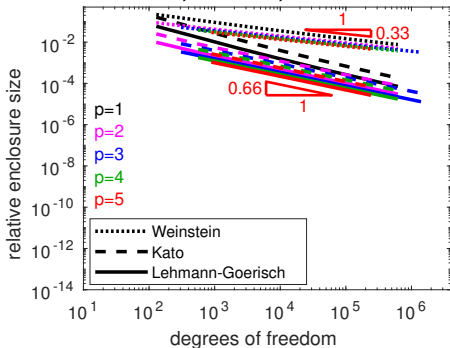
Example: dumbbell

$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

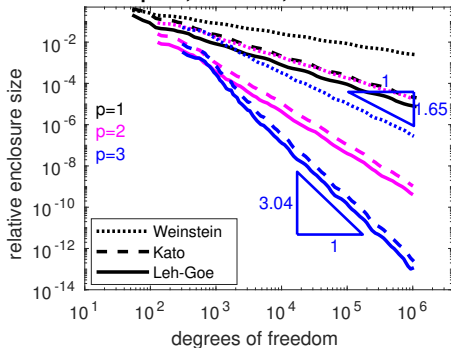
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1



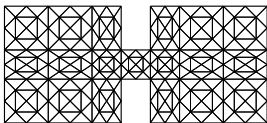
- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell

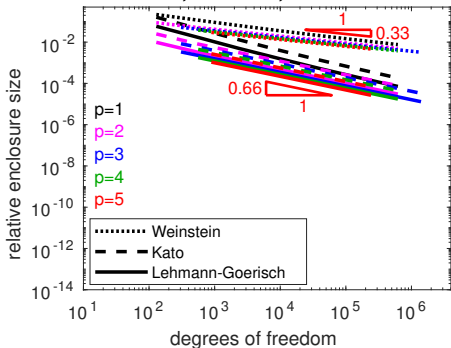


$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

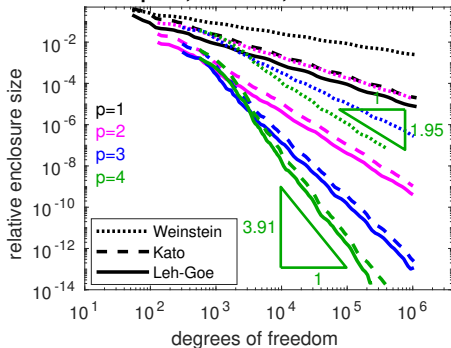
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1



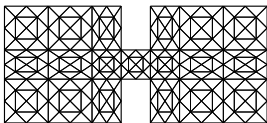
- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell

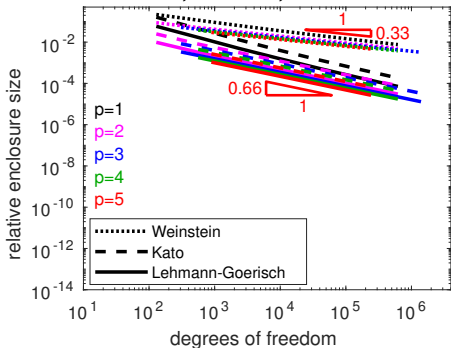


$$-\Delta u_i = \lambda_i u_i \quad \text{in } \Omega$$

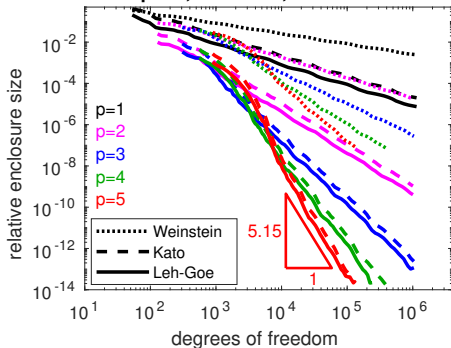
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

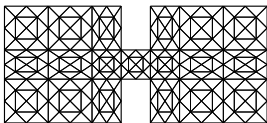


- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

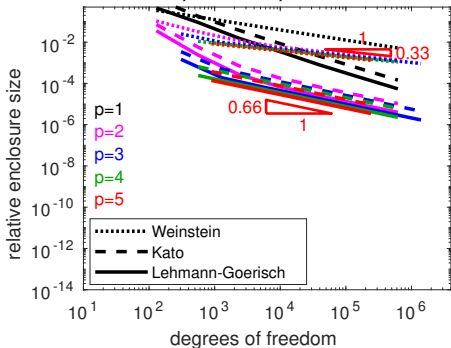
Example: dumbbell



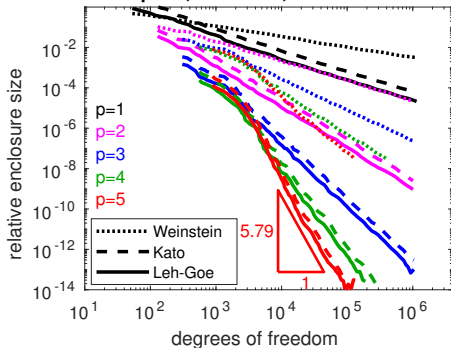
$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



Uniform, dumbbell, lambda5



Adaptive, dumbbell, lambda5

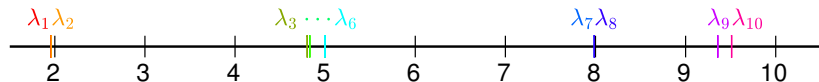
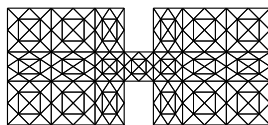


- ▶ relative enclosure size: $(\lambda_{h,i} - l_i)/l_i$
- ▶ $\gamma = 10^{-6}$, $\rho := l_{11}^{\text{Wein}} \approx 10.0017 \leq \lambda_{11}$

Example: dumbbell



$$\begin{aligned} -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\ u_i &= 0 && \text{on } \partial\Omega \end{aligned}$$



Computed bounds ($p = 5$, adaptive):

$$1.9557937945883 \leq \lambda_1 \leq 1.9557937945884$$

$$1.9606830315950 \leq \lambda_2 \leq 1.9606830315951$$

$$4.8007611240339 \leq \lambda_3 \leq 4.8007611240345$$

$$4.8298952545005 \leq \lambda_4 \leq 4.8298952545010$$

$$4.9968370972489 \leq \lambda_5 \leq 4.9968370972490$$

$$4.9968509041015 \leq \lambda_6 \leq 4.9968509041016$$

$$7.9869672921028 \leq \lambda_7 \leq 7.9869672921038$$

$$7.9870343068216 \leq \lambda_8 \leq 7.9870343068227$$



- ▶ optimal speed of convergence
- ▶ implementation based on standard FEM
- ▶ adaptivity for free
- ▶ naturally generalize to higher orders
- ▶ good for a wide class of problems
- ▶ an a priori lower bound on some eigenvalue is needed



4. Lower bounds on eigenvalues

4.3 Interpolation constant

based methods

[Carstensen, Gallistl, Gedicke 2014], [Liu 2015]



Nonconforming approximation

Eigenvalue problem: Find λ_n and $u_n \in V \setminus \{0\}$ such that

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V$$

Finite dimensional space: $\dim V_h = N < \infty$, but it can be $V_h \not\subset V$.

Discrete eigenvalue problem: Find $\lambda_{h,n} \in \mathbb{R}$, $u_{h,n} \in V_h \setminus \{0\}$:

$$a(u_{h,n}, v_h) = \lambda_{h,n} b(u_{h,n}, v_h) \quad \forall v_h \in V_h$$

Definition:

$$V(h) = V + V_h = \{v + v_h : v \in V, v_h \in V_h\}$$

Extensions of bilinear forms:

$$a_h, b_h : V(h) \times V(h) \rightarrow \mathbb{R}$$

$$a_h(u, v) = a(u, v) \quad \text{and} \quad b_h(u, v) = b(u, v) \quad \forall u, v \in V$$

$a_h(\cdot, \cdot)$ is symmetric and $V(h)$ -elliptic

$b_h(\cdot, \cdot)$ is symmetric and positive semidefinite on $V(h)$

Notation: $a = a_h$ and $b = b_h$



Lemma 1 (Discrete Friedrichs inequality).

$$|v_h|_b \leq \lambda_{h,1}^{-1/2} \|v_h\|_a \quad \forall v_h \in V_h$$

Proof. $\lambda_{h,1} = \min_{w_h \in V_h} \frac{\|w_h\|_a^2}{|w_h|_b^2} \leq \frac{\|v_h\|_a^2}{|v_h|_b^2}$



Elliptic projection: $P_h : V(h) \rightarrow V_h$

$$a(u - P_h u, v_h) = 0 \quad \forall v_h \in V_h$$

Lemma 2.

$$\|u\|_a^2 = \|P_h u\|_a^2 + \|u - P_h u\|_a^2$$

Proof.

$$\begin{aligned} \|u - P_h u\|_a^2 &= \|u\|_a^2 - 2a(u, P_h u) + \|P_h u\|_a^2 \\ a(u, P_h u) &= a(P_h u, P_h u) = \|P_h u\|_a^2 \end{aligned}$$





Lower bound

Theorem. Let $|u - P_h u|_b \leq C_h \|u - P_h u\|_a$ for all $u \in V$. Then

$$\frac{\lambda_{h,n}}{1 + \lambda_{h,n} C_h^2} \leq \lambda_n, \quad n = 1, 2, \dots, N.$$

Proof (for λ_1 only). Let $v \in V$.

$$\begin{aligned} |v|_b &\leq |P_h v|_b + |v - P_h v|_b \\ &\leq \lambda_{h,1}^{-1/2} \|P_h v\|_a + C_h \|v - P_h v\|_a \\ &\leq \left(\lambda_{h,1}^{-1} + C_h^2 \right)^{1/2} \left(\|P_h v\|_a^2 + \|v - P_h v\|_a^2 \right)^{1/2} \\ &= \left(\frac{1 + \lambda_{h,1} C_h^2}{\lambda_{h,1}} \right)^{1/2} \|v\|_a \end{aligned}$$

$$\lambda_1 = \min_{v \in V} \frac{\|v\|_a^2}{|v|_b^2} \geq \frac{\lambda_{h,1}}{1 + \lambda_{h,1} C_h^2}$$

Crouzeix–Raviart (CR) elements

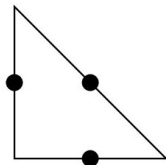


Laplace eigenvalue problem: Find $\lambda_n \in \mathbb{R}$, $u_n \in H_0^1(\Omega) \setminus \{0\}$:

$$(\nabla u_n, \nabla v) = \lambda_n (u_n, v) \quad \forall v \in H_0^1(\Omega)$$

CR space: $v_h \in V_h^{\text{CR}}$ if

- ▶ $v_h|_K \in \mathbb{P}^1(K)$
- ▶ v_h is continuous at midpoints of interior edges
- ▶ $v_h = 0$ at midpoints of boundary edges



CR eigenvalue problem: Find $\lambda_{h,i}^{\text{CR}} \in \mathbb{R}$, $u_{h,i}^{\text{CR}} \in V_h^{\text{CR}} \setminus \{0\}$:

$$(\nabla u_{h,i}^{\text{CR}}, \nabla v_h) = \lambda_{h,i}^{\text{CR}} (u_{h,i}^{\text{CR}}, v_h) \quad \forall v_h \in V_h^{\text{CR}}.$$



Crouzeix–Raviart interpolation

Let e_i , $i = 1, 2, 3$, be edges of triangle K .

Definition: $\Pi_h : H^1(K) \rightarrow \mathbb{P}^1(K)$ such that

$$\int_{e_i} u - \Pi_h u \, ds = 0 \quad \forall i = 1, 2, 3.$$

Note: If m_i is a midpoint of e_i then $\Pi_h u(m_i) = \frac{1}{|e_i|} \int_{e_i} u \, ds$.

Lemma. $\Pi_h = P_h$

Proof.

Let $u \in H^1(\Omega) \oplus V_h^{\text{CR}}$ and $v_h \in V_h^{\text{CR}}$.

$$\begin{aligned} a(u - \Pi_h u, v_h) &= \sum_{K \in \mathcal{T}_h} \int_K \nabla(u - \Pi_h u) \cdot \nabla v_h \\ &= \sum_{K \in \mathcal{T}_h} \left(\sum_{i=1}^3 \int_{e_i} (u - \Pi_h u) \underbrace{\frac{\partial v_h}{\partial \mathbf{n}}}_{=\text{const.}} \, ds - \int_K (u - \Pi_h u) \underbrace{\Delta v_h}_{=0} \, dx \right) = 0 \end{aligned}$$



The value of C_h

Interpolation error estimate:

$$\|u - \Pi_h u\|_{L^2(\Omega)} \leq C_h \|\nabla u - \nabla \Pi_h u\|_{L^2(\Omega)}$$

Local interpolation error estimate:

$$\|u - \Pi_h u\|_{L^2(K)} \leq C_h(K) \|\nabla u - \nabla \Pi_h u\|_{L^2(K)}$$

Lemma.

$$C_h \leq \max_{K \in \mathcal{T}_h} C_h(K)$$

Proof.

$$\begin{aligned} \|u - \Pi_h u\|_{L^2(\Omega)}^2 &= \sum_{K \in \mathcal{T}_h} \|u - \Pi_h u\|_{L^2(K)}^2 \leq \sum_{K \in \mathcal{T}_h} C_h^2(K) \|\nabla u - \nabla \Pi_h u\|_{L^2(K)}^2 \\ &\leq \max_{K \in \mathcal{T}_h} C_h^2(K) \|\nabla u - \nabla \Pi_h u\|_{L^2(\Omega)}^2 \end{aligned}$$

Explicit estimates of C_h



Interval

- ▶ $C_h = h/\pi$

Triangle

- ▶ $C_h = 0.4396h$ [Carstensen, Gedicke 2014]

- ▶ $C_h = 0.2983h$ [Carstensen, Gallistl 2014]

- ▶ $C_h = 0.1893h$ [Liu 2015]

Tetrahedron

- ▶ $C_h = 0.3804h$ [Liu 2015]



Explicit estimate of C_h for an interval

Setting: $\Omega = (\alpha, \beta)$, $V = H_0^1(\alpha, \beta)$,
 $a(u, v) = \int_{\alpha}^{\beta} u'v' dx$, $b(u, v) = \int_{\alpha}^{\beta} uv dx$

Partition: $\alpha = z_0 < z_1 < \dots < z_N = \beta$

Elements: $K_i = [z_{i-1}, z_i]$, $i = 1, 2, \dots, N$,
 $h_i = z_i - z_{i-1}$, $h = \max_{i=1, \dots, N} h_i$

CR space: $V_h = \{v \in H_0^1(\alpha, \beta) : v|_{K_i} \in \mathbb{P}^1(K_i), i = 1, 2, \dots, N\}$

Interpolation: $\Pi_h : H_0^1(\alpha, \beta) \rightarrow V_h$
 $(\Pi_h u)(x_i) = u(x_i)$, $i = 0, \dots, N$

Lemma.

$$\|u - \Pi_h u\|_{L^2(\Omega)} \leq \frac{h}{\pi} \|u' - (\Pi_h u)'\|_{L^2(\Omega)}$$

Proof.

$$\min_{v \in H^1(K_i)} R(v - \Pi_h v) = \min_{w \in H_0^1(K_i)} R(w) = R\left(\sin \frac{\pi(x - z_i)}{h_i}\right) = \pi^2/h_i^2$$



Upper bound

Interpolation to continuous functions: $\mathcal{I} : V_h^{\text{CR}} \rightarrow \tilde{V}_h \subset H^1(\Omega)$

Examples:

- ▶ Oswald quasi-interpolation [Oswald 1994]
- ▶ Interpolation to refined mesh [Carstensen, Merdon 2013]

Upper bound

- ▶ \mathcal{T}_h^* is the red refinement of \mathcal{T}_h
- ▶ $u_{h,i}^* = \mathcal{I}_{\text{CM}} \tilde{u}_{h,i}^{\text{CR}}$ for $i = 1, 2, \dots, m$
- ▶ $\mathbf{S}, \mathbf{Q} \in \mathbb{R}^{m \times m}$ with entries $\mathbf{S}_{j,k} = (\nabla u_{h,j}^*, \nabla u_{h,k}^*)$ and $\mathbf{Q}_{j,k} = (u_{h,j}^*, u_{h,k}^*)$
- ▶ $\mathbf{S} \mathbf{y}_i = \Lambda_i^* \mathbf{Q} \mathbf{y}_i, \quad i = 1, 2, \dots, m$
- ▶ $\Lambda_1^* \leq \Lambda_2^* \leq \dots \leq \Lambda_m^*$
- ▶ $\lambda_i \leq \Lambda_i^*$ for $i = 1, 2, \dots, m$

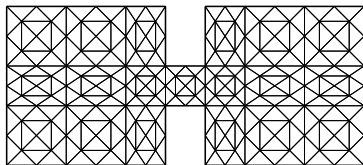


Example: dumbbell

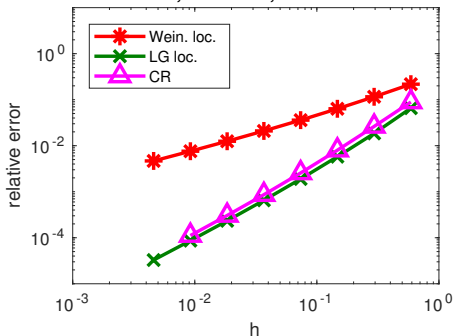
$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega = \text{dumbbell} \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\text{Rel. error: } \frac{|\lambda_n - \lambda_{h,n}|}{\lambda_n} \leq \frac{\lambda_{h,n} - \ell_n}{\ell_n}$$

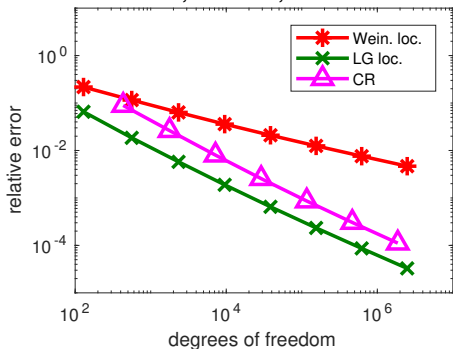
$$\gamma = 10^{-6}$$



Uniform, dumbbell, lambda1



Uniform, dumbbell, lambda1



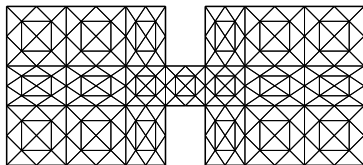


Example: dumbbell

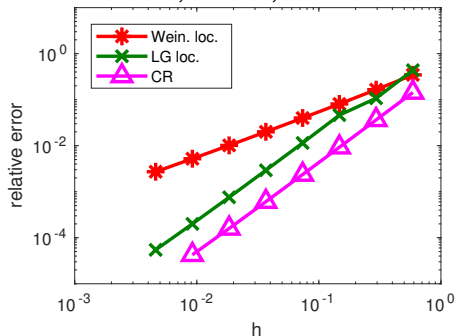
$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega = \text{dumbbell} \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\text{Rel. error: } \frac{|\lambda_n - \lambda_{h,n}|}{\lambda_n} \leq \frac{\lambda_{h,n} - \ell_n}{\ell_n}$$

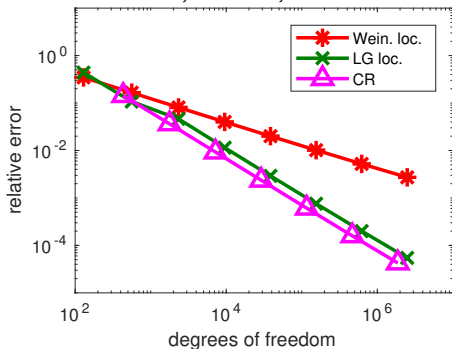
$$\gamma = 10^{-6}$$



Uniform, dumbbell, lambda5



Uniform, dumbbell, lambda5



Interpolation constant based method – summary



- ▶ no a priori information needed
- ▶ optimal speed of convergence
- ▶ easy to implement
- ▶ interpolation constant known in special cases only
- ▶ adaptivity is not for free
- ▶ higher order variant is not available



5. Guaranteed bounds on eigenfunctions

[work in progress, collaboration with X. Liu]

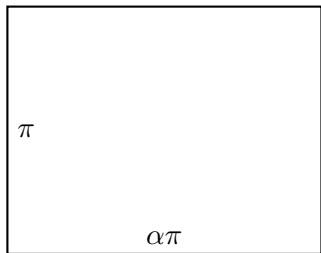
Laplace eigenvalue problem in a rectangle



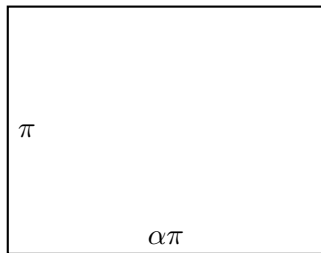
Laplace eigenvalue problem in a rectangle



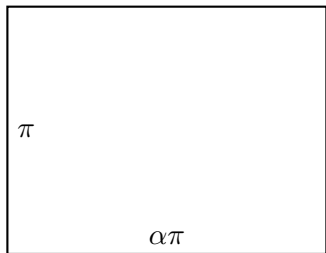
$$\alpha = 1.27$$



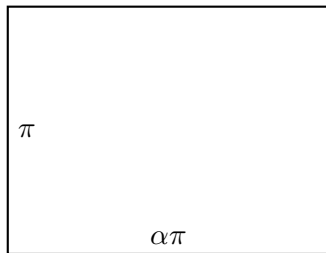
$$\alpha = 1.28$$



$$\alpha = 1.29$$



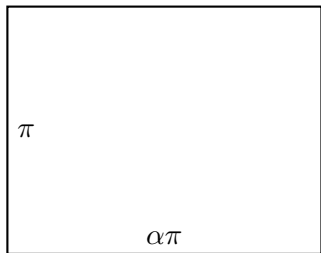
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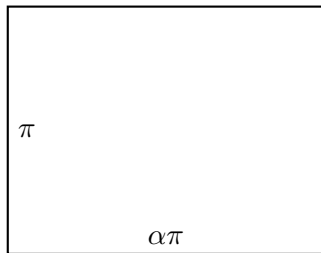


Laplace eigenvalue problem in a rectangle

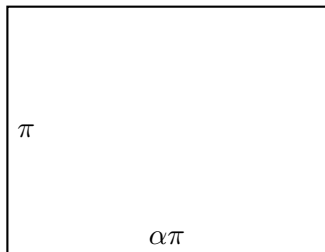
$$\alpha = 1.27, \lambda_4 = 6.4800$$



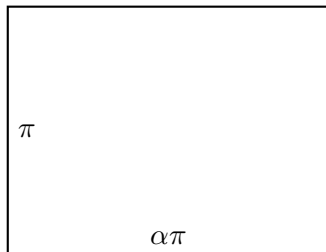
$$\alpha = 1.28$$



$$\alpha = 1.29$$



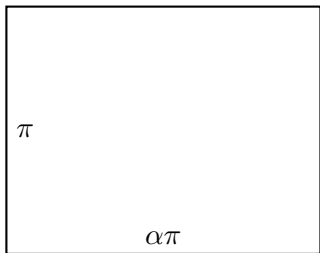
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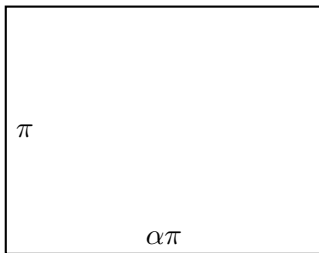


Laplace eigenvalue problem in a rectangle

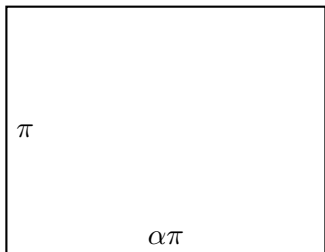
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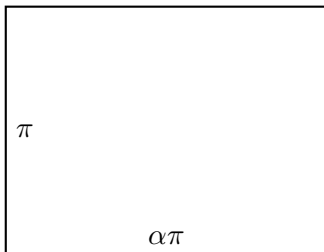
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$$\alpha = 1.29$$



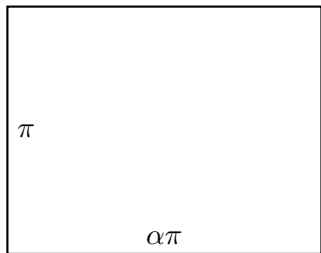
$$\alpha = 1.30$$



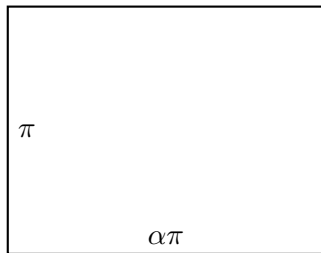
Laplace eigenvalue problem in a rectangle



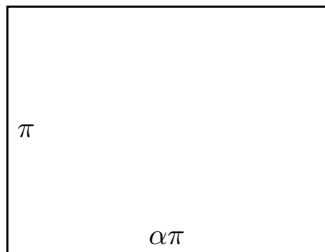
$$\alpha = 1.27, \lambda_4 = 6.4800$$



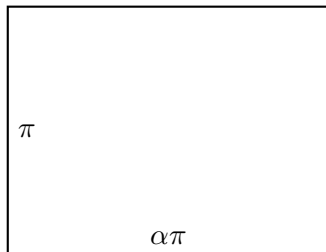
$$\alpha = 1.28, \lambda_4 = 6.4414$$



$$\alpha = 1.29, \lambda_4 = 6.4037$$



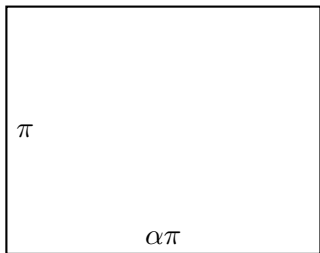
$$\alpha = 1.30$$



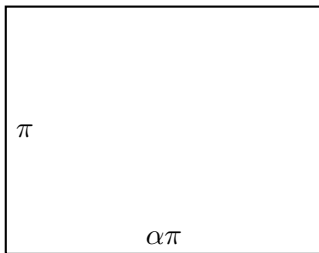
Laplace eigenvalue problem in a rectangle



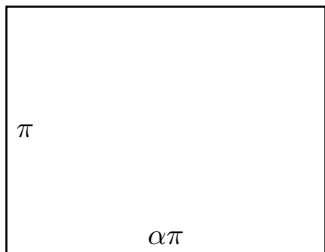
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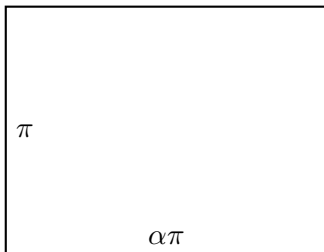
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$$\alpha = 1.29, \lambda_4 = 6.4037$$



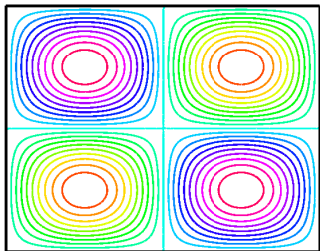
$$\alpha = 1.30, \lambda_4 = 6.3254$$



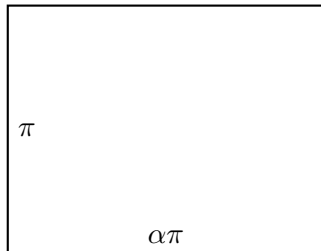


Laplace eigenvalue problem in a rectangle

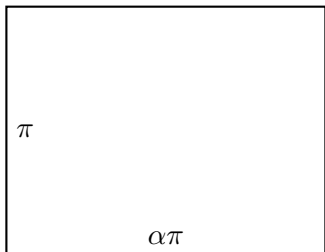
$$\alpha = 1.27, \lambda_4 = 6.4800$$



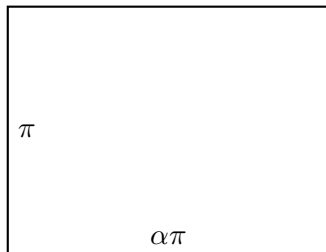
$$\alpha = 1.28, \lambda_4 = 6.4414$$



$$\alpha = 1.29, \lambda_4 = 6.4037$$



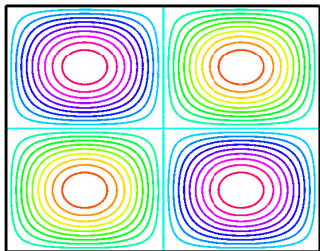
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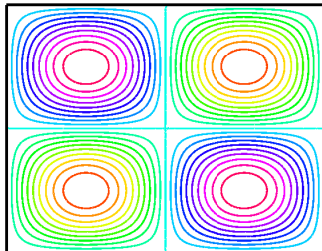


Laplace eigenvalue problem in a rectangle

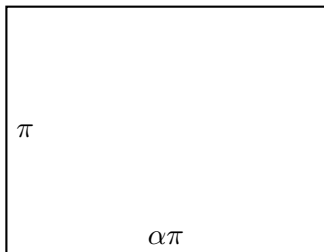
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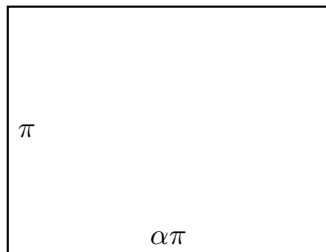
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$$\alpha = 1.29, \lambda_4 = 6.4037$$



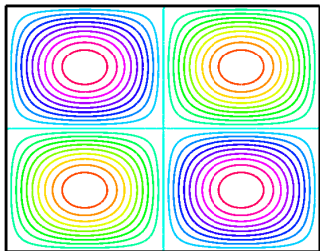
$$\alpha = 1.30, \lambda_4 = 6.3254$$



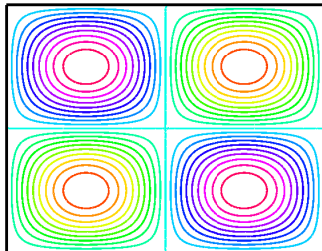


Laplace eigenvalue problem in a rectangle

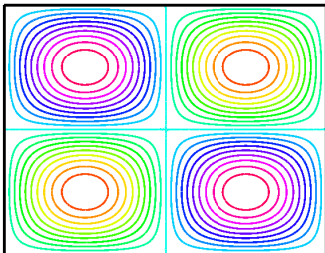
$$\alpha = 1.27, \lambda_4 = 6.4800$$



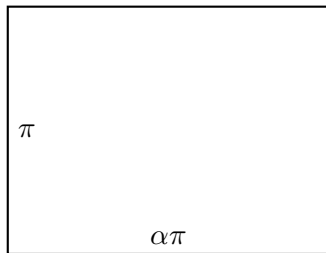
$$\alpha = 1.28, \lambda_4 = 6.4414$$



$$\alpha = 1.29, \lambda_4 = 6.4037$$



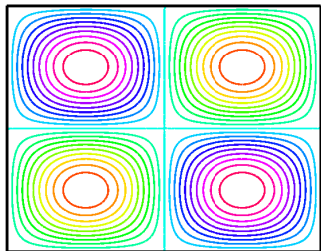
$$\alpha = 1.30, \lambda_4 = 6.3254$$



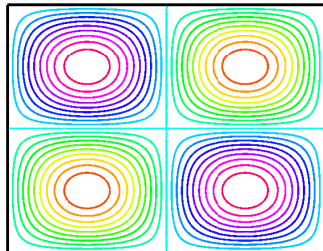
Laplace eigenvalue problem in a rectangle



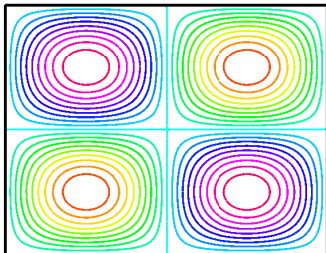
$$\alpha = 1.27, \lambda_4 = 6.4800$$



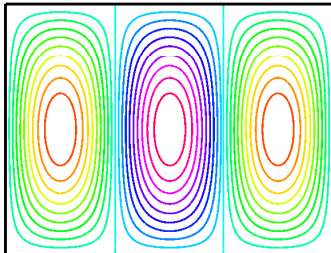
$$\alpha = 1.28, \lambda_4 = 6.4414$$



$$\alpha = 1.29, \lambda_4 = 6.4037$$



$$\alpha = 1.30, \lambda_4 = 6.3254$$

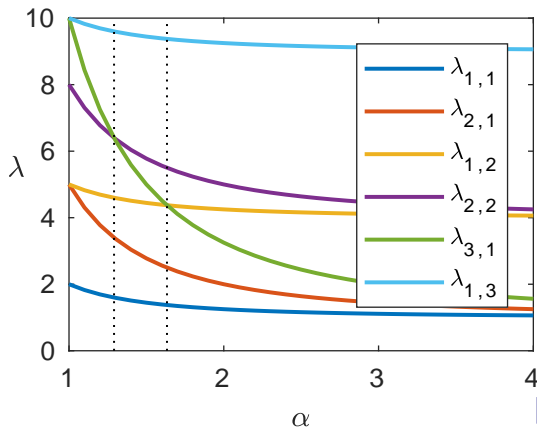


Laplace eigenvalue problem in a rectangle



$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega &= (0, \alpha\pi) \times (0, \pi) \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

Dependence of eigenvalues on α



$$\lambda_{k,m} = \frac{k^2}{\alpha^2} + m^2$$
$$u_{k,m} = \sin \frac{kx}{\alpha} \sin(my)$$

$$\alpha^* = \sqrt{5/3}$$
$$\approx 1.2910$$

[Trefethen, Betcke 2006]



Error bounds on eigenfunctions

Problem

- ▶ Eigenfunctions may be ill-posed \Rightarrow spaces of eigenfunctions
- ▶ Directed distance of spaces $\delta(E, E_h)$ [Meyer 2000]

Assume

- ▶ $\lambda_n, \lambda_{n+1}, \dots, \lambda_N$ (cluster)
- ▶ $E = \text{span}\{u_n, u_{n+1}, \dots, u_N\}$ (space of eigenfunctions)
- ▶ $E_h = \text{span}\{u_{h,n}, u_{h,n+1}, \dots, u_{h,N}\}$ (its approximation)
- ▶ $\ell_i \leq \lambda_i \leq \lambda_{h,i}$ (two sided bounds on eigenvalues)

\Rightarrow

Compute an upper bound on $\delta(E, E_h)$



Directed distance of spaces

Definition

Let E and E_h be two subspaces of a Hilbert space V then

$$\delta(E, E_h) = \max_{\substack{v \in E \\ \|v\|=1}} \min_{v_h \in E_h} \|v - v_h\|$$

Properties

- ▶ if $\dim E = \dim E_h$ then $\delta(E, E_h) = \delta(E_h, E)$
- ▶ $\delta^2(E, E_h) = 1 - \min_{\substack{v \in E \\ \|v\|=1}} \max_{\substack{v_h \in E_h \\ \|v_h\|=1}} |(v, v_h)|^2$

Example

Let $E = \text{span}\{u\}$ and $E_h = \text{span}\{u_h\}$ then

$$\delta^2(E, E_h) = 1 - \frac{|(u, u_h)|^2}{\|u\|^2 \|u_h\|^2} = 1 - \cos^2 \alpha = \sin^2 \alpha$$

$$\|u - u_h\|^2 = \|u\|^2 + \|u_h\|^2 - 2\|u\|\|u_h\|\sqrt{1 - \delta^2(E, E_h)}$$



Eigenvalue problem:

Find $\lambda_n > 0$ and $u_n \in V \setminus \{0\}$ such that

$$a(u_n, v) = \lambda_n b(u_n, v) \quad \forall v \in V.$$

Consider

- ▶ $E = \text{span}\{u_n, u_{n+1}, \dots, u_N\}$, $b(u_i, u_j) = \delta_{ij}$, $a(u_i, u_j) = \lambda_i \delta_{ij}$
- ▶ $E_h = \text{span}\{u_{h,n}, u_{h,n+1}, \dots, u_{h,N}\}$

Goal

$$\text{Upper bound on } \delta(E, E_h) = \max_{\substack{v \in E \\ \|v\|_a=1}} \min_{v_h \in E_h} \|v - v_h\|_a$$

Lehmann-like estimate of eigenfunctions



Theorem

Let $\lambda_{n-1} \leq \xi < \lambda_n$, $\lambda_N < \rho \leq \lambda_{N+1}$, $\theta \geq \max_{i=n, \dots, N} \left(\frac{\xi + \rho}{\lambda_i} - \frac{\xi \rho}{\lambda_i^2} \right)$

▶ $u_{h,n}, u_{h,n+1}, \dots, u_{h,N} \in V$ be linearly independent

▶ $A_{0,ij} = a(u_{h,i}, u_{h,j})$

▶ $A_{1,ij} = b(u_{h,i}, u_{h,j})$

▶ $w_i \in V : a(w_i, v) = b(u_{h,i}, v) \quad \forall v \in V$

$A_{2,ij} = a(w_i, w_j)$

▶ μ_{\min} be the smallest eigenvalue of $[(\xi + \rho)A_1 - \xi \rho A_2] \mathbf{x} = \mu A_0 \mathbf{x}$

Then

$$\delta^2(E, E_h) \leq \frac{\theta - \mu_{\min}}{\theta - 1}$$

Lehmann–Goerisch-like estimate of eigenfunctions



Theorem

Let $\lambda_{n-1} \leq \xi < \lambda_n$, $\lambda_N < \rho \leq \lambda_{N+1}$, $\theta \geq \max_{i=n, \dots, N} \left(\frac{\xi + \rho}{\lambda_i} - \frac{\xi \rho}{\lambda_i^2} \right)$

- ▶ $u_{h,n}, u_{h,n+1}, \dots, u_{h,N} \in V$ be linearly independent
- ▶ $A_{0,ij} = a(u_{h,i}, u_{h,j})$
- ▶ $A_{1,ij} = b(u_{h,i}, u_{h,j})$
- ▶ X ... vector space
- ▶ \mathcal{B} ... positive semidefinite symmetric bilinear form on X
- ▶ $T : V \rightarrow X$... linear operator:
 - $\mathcal{B}(Tu, Tv) = a(u, v) \quad \forall u, v \in V$
 - $\hat{\mathbf{w}}_i \in X : \mathcal{B}(\hat{\mathbf{w}}_i, Tv) = b(\tilde{u}_i, v) \quad \forall v \in V$
 - $\hat{A}_{2,ij} = \mathcal{B}(\hat{\mathbf{w}}_i, \hat{\mathbf{w}}_j)$
- ▶ $\hat{\mu}_{\min}$ be the smallest eigenvalue of $\left[(\xi + \rho)A_1 - \xi\rho\hat{A}_2 \right] \mathbf{x} = \hat{\mu}A_0\mathbf{x}$

Then

$$\delta^2(E, E_h) \leq \frac{\theta - \hat{\mu}_{\min}}{\theta - 1}$$



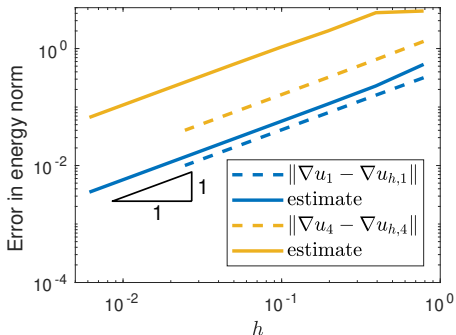
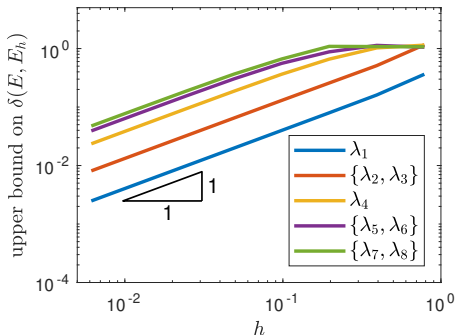
Example

Laplace eigenvalue problem in a square

$$\begin{aligned} -\Delta u_n &= \lambda_n u_n & \text{in } \Omega &= (0, \pi)^2 \\ u_n &= 0 & \text{on } \partial\Omega \end{aligned}$$

Exact eigenvalues

$$\lambda_1 = 2, \quad \lambda_2 = \lambda_3 = 5, \quad \lambda_4 = 8, \quad \lambda_5 = \lambda_6 = 10, \quad \lambda_7 = \lambda_8 = 13$$





6. Literature



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- ▶ T. Kato, *On the upper and lower bounds of eigenvalues*, J. Phys. Soc. Japan 4 (1949) 334–339.
- ▶ N.J. Lehmann, *Beiträge zur numerischen Lösung linearer Eigenwertprobleme. I and II*, ZAMM Z. Angew. Math. Mech. 29 (1949) 341–356 and 30 (1950) 1–16.
- ▶ F. Goerisch, H. Haunhorst, *Eigenwertschranken für Eigenwertaufgaben mit partiellen Differentialgleichungen*, ZAMM Z. Angew. Math. Mech. 65 (3) (1985) 129–135.



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My contributions

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Appendix



Weinstein bound:

- ▶ λ_* , u_* , \mathbf{q} can be arbitrary
- ▶ $\eta^2 = \|\nabla u_* - \mathbf{q}\|_0^2 + \frac{1}{\gamma} \|\lambda_* u_* + \operatorname{div} \mathbf{q}\|_0^2$
must be evaluated exactly (*)

Lehmann–Goerisch method:

- ▶ \tilde{u}_i , σ_i can be arbitrary
- ▶ $(A_0 - \rho A_1)\hat{\mathbf{x}} = \hat{\mu}(A_0 - 2\rho A_1 + \rho^2 \hat{A}_2)\hat{\mathbf{x}}$
must be solved exactly (*)

Interpolation constant based method:

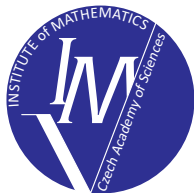
- ▶ $\lambda_{h,i}^{\text{CR}}$ must be computed exactly (*)

(*) or bounded by **interval arithmetic!**

Thank you for your attention

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