

Flux reconstructions and lower bounds on eigenvalues

Tomáš Vejchodský (vejchod@math.cas.cz)

Institute of Mathematics
Czech Academy of Sciences



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- ▶ Introduction
- ▶ Lehmann–Goerisch method
- ▶ Flux reconstruction
- ▶ Numerical examples



Eigenvalue problem

Find $u_i \in V$, $u_i \neq 0$, and $\lambda_i \in \mathbb{R}$ such that

$$a(u_i, v) = \lambda_i b(u_i, v) \quad \forall v \in V.$$

Setting

- ▶ V ... Hilbert space
- ▶ a ... V -elliptic, symmetric bilinear form on V
- ▶ b ... symmetric positive (semi)definite bilinear form on V
- ▶ Let $|\cdot|_b$ be compact with respect to $\|\cdot\|_a$. (Any sequence bounded in $\|\cdot\|_a$ contains a Cauchy subsequence in $|\cdot|_b$.)

Facts

- ▶ $0 < \lambda_1 \leq \lambda_2 \leq \dots$ countable sequence of eigenvalues
- ▶ $b(u_i, u_j) = \delta_{ij} \quad \forall i, j = 1, 2, \dots$
- ▶ $|v|_b^2 = \sum_{i=1}^{\infty} |b(v, u_i)|^2$ for all $v \in V$
- ▶ $\|v\|_a^2 = \sum_{i=1}^{\infty} \lambda_i |b(v, u_i)|^2$ for all $v \in V$ (if b positive definite)

Example: Laplace eigenvalue problem



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



Example: Laplace eigenvalue problem

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

Weak formulation

$$\lambda_i > 0, u_i \in V : \quad (\nabla u_i, \nabla v) = \lambda_i (u_i, v) \quad \forall v \in V$$

Notation:

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



Example: Laplace eigenvalue problem

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

Weak formulation

$$\lambda_i > 0, u_i \in V : \quad (\nabla u_i, \nabla v) = \lambda_i (u_i, v) \quad \forall v \in V$$

Finite element method

$$\Lambda_{h,i} > 0, u_{h,i} \in V_h : \quad (\nabla u_{h,i}, \nabla v_h) = \Lambda_{h,i} (u_{h,i}, v_h) \quad \forall v_h \in V_h$$

Notation:

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

$$V_h = \{v_h \in V : v_h|_K \in P_p(K) \quad \forall K \in \mathcal{T}_h\}$$



Theorem

Let

- ▶ $\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_N \in V$... linearly independent
- ▶ $A_{0,ij} = a(\tilde{u}_i, \tilde{u}_j)$
- ▶ $A_{1,ij} = b(\tilde{u}_i, \tilde{u}_j)$
- ▶ $\Lambda_1 \leq \Lambda_2 \leq \dots \leq \Lambda_N$ be eigenvalues of $A_0 \mathbf{x}_n = \Lambda_n A_1 \mathbf{x}_n$

Then

$$\lambda_n \leq \Lambda_n \quad \forall n = 1, 2, \dots, N.$$



Standard (conforming) approach:

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

Nonconforming FEM:

Carstensen (2013), Gedicke (2014), Gallistl (2013),
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, Xuefeng Liu, M. Plum, S.I. Repin,
V.G. Sigillito, M. Vohralík, Hehu Xie, Yidu Yang, Zhimin Zhang,
... *many others*



Theorem

Let $\Lambda_N < \rho \leq \lambda_{N+1}$

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

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

Then

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



Theorem

Let $\Lambda_N < \rho \leq \lambda_{N+1}$

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

Observation

$$\begin{aligned} a(w_i, v) &= b(\tilde{u}_i, v) \\ &\approx \frac{1}{\Lambda_{h,i}} a(\tilde{u}_i, v) \\ &\quad \forall v \in V \\ &\Rightarrow w_i \approx \frac{1}{\Lambda_{h,i}} \tilde{u}_i \end{aligned}$$

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

Then

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



Theorem

Let $\Lambda_N < \rho \leq \lambda_{N+1}$

- ▶ $\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_N \in V$ 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: $\mathcal{B}(Tu, Tv) = a(u, v) \quad \forall u, v \in V$
- ▶ $\hat{\mathbf{w}}_i \in X : \mathcal{B}(\hat{\mathbf{w}}_i, Tv) = b(\tilde{u}_i, v) \quad \forall v \in V$
- ▶ $\hat{A}_{2,ij} = \mathcal{B}(\hat{\mathbf{w}}_i, \hat{\mathbf{w}}_j)$
- ▶ $\hat{\mu}_1 \leq \hat{\mu}_2 \leq \dots \leq \hat{\mu}_N : (A_0 - \rho A_1)\hat{\mathbf{x}} = \hat{\mu}(A_0 - 2\rho A_1 + \rho^2 \hat{A}_2)\hat{\mathbf{x}}$

Then

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

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



Need

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

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

$$\Rightarrow \hat{\mathbf{w}}_i \approx Tw_i \approx \frac{1}{\Lambda_{h,i}} T\tilde{u}_i \quad (\text{using Observation } w_i \approx \frac{1}{\Lambda_{h,i}} \tilde{u}_i)$$

Natural idea

make $|\hat{\mathbf{w}}_i - \frac{1}{\Lambda_{h,i}} T\tilde{u}_i|_{\mathcal{B}}^2$ small



Example: Laplace eigenvalue problem

$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega)$$

Setting

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

Facts

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



Example: Laplace eigenvalue problem

$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega)$$

Setting

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

Facts

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

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

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

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

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



Example: Laplace eigenvalue problem

$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega)$$

Setting

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

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

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



Example: Laplace eigenvalue problem

$$(\nabla u_i, \nabla v) + \gamma(u_i, v) = (\lambda_i + \gamma)(u_i, v) \quad \forall v \in H_0^1(\Omega)$$

Setting

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

Facts

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

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

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



Example: Laplace eigenvalue problem

Theorem (Lehmann–Goerisch)

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

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

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

Then

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



Choice of σ_j

$$\text{minimize } |\hat{\mathbf{w}}_j - \frac{1}{\Lambda_i + \gamma} T \tilde{u}_j|_{\mathcal{B}}^2 = \left\| \sigma_j - \frac{\nabla \tilde{u}_j}{\Lambda_i + \gamma} \right\|_0^2 + \frac{1}{\gamma} \left\| \operatorname{div} \sigma_j + \frac{\Lambda_j \tilde{u}_j}{\Lambda_i + \gamma} \right\|_0^2$$

(i) Constraint minimization: [Behnke, Mertins, Plum, Wieners 2000]

$$\text{minimize } \left\| \sigma_j - \frac{\nabla \tilde{u}_j}{\Lambda_i + \gamma} \right\|_0^2 \text{ over } \sigma_j \in \mathbf{W}_h \subset \mathbf{H}(\operatorname{div}, \Omega)$$

$$\text{under the constraint: } \operatorname{div} \sigma_j + \frac{\Lambda_j \tilde{u}_j}{\Lambda_i + \gamma} = 0$$

\Leftrightarrow

Find $\sigma_{h,i} \in \mathbf{W}_h$, $q_{h,i} \in Q_h$, $i = 1, 2, \dots, N$

$$(\sigma_{h,i}, \mathbf{w}_h) + (q_{h,i}, \operatorname{div} \mathbf{w}_h) = \left(\frac{\nabla \tilde{u}_j}{\Lambda_i + \gamma}, \mathbf{w}_h \right) \quad \forall \mathbf{w}_h \in \mathbf{W}_h$$

$$(\operatorname{div} \sigma_{h,i}, \varphi_h) = \left(-\frac{\Lambda_j \tilde{u}_j}{\Lambda_i + \gamma}, \varphi_h \right) \quad \forall \varphi_h \in Q_h$$

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

$$Q_h = \{ q_h \in L^2(\Omega) : q_h|_K \in P_p(K) \quad \forall K \in \mathcal{T}_h \}$$



Choice of σ_j

$$\text{minimize } |\hat{\mathbf{w}}_j - \frac{1}{\Lambda_j + \gamma} T \tilde{u}_j|_{\mathcal{B}}^2 = \left\| \sigma_j - \frac{\nabla \tilde{u}_j}{\Lambda_j + \gamma} \right\|_0^2 + \frac{1}{\gamma} \left\| \operatorname{div} \sigma_j + \frac{\Lambda_j \tilde{u}_j}{\Lambda_j + \gamma} \right\|_0^2$$

(ii) Unconstraint minimization:

Find $\sigma_{h,j} \in \mathbf{W}_h$, $i = 1, 2, \dots, N$

$$(\sigma_{h,i}, \mathbf{w}_h) + \frac{1}{\gamma} (\operatorname{div} \sigma_{h,i}, \operatorname{div} \mathbf{w}_h) = \left(\frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma}, \mathbf{w}_h \right) - \frac{1}{\gamma} \left(\frac{\Lambda_i \tilde{u}_i}{\Lambda_i + \gamma}, \operatorname{div} \mathbf{w}_h \right)$$

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

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



Choice of σ_j

$$\text{minimize } |\hat{\mathbf{w}}_j - \frac{1}{\Lambda_i + \gamma} T \tilde{u}_j|_{\mathcal{B}}^2 = \left\| \sigma_j - \frac{\nabla \tilde{u}_j}{\Lambda_i + \gamma} \right\|_0^2 + \frac{1}{\gamma} \left\| \text{div } \sigma_j + \frac{\Lambda_i \tilde{u}_j}{\Lambda_i + \gamma} \right\|_0^2$$

(iii) Local constraint minimization:

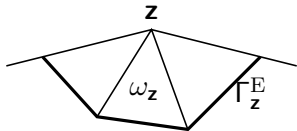
[Braess, Schöberl 2000], [Ern, Vohralík 2013]

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

where $\sigma_{z,i} \in \mathbf{W}_z$ minimizes $\left\| \sigma_{z,i} - \psi_z \frac{\nabla \tilde{u}_j}{\Lambda_i + \gamma} \right\|_{0, \omega_z}^2$

under the constraint: $\text{div } \sigma_{z,i} + \frac{\psi_z \Lambda_i \tilde{u}_j}{\Lambda_i + \gamma} + \frac{\nabla \psi_z \cdot \nabla \tilde{u}_j}{\Lambda_i + \gamma} = 0$ in ω_z

Partition of unity: $\sum_{z \in \mathcal{N}_h} \psi_z \equiv 1$ in Ω



$$\mathbf{W}_z = \{ \sigma_z \in \mathbf{H}(\text{div}, \omega_z) : \sigma_z|_K \in \mathbf{RT}_p(K) \ \forall K \in \mathcal{T}_z \text{ and } \sigma_z \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^E \}$$

$$Q_z = \{ q_z \in L^2(\omega_z) : q_z|_K \in P_p(K) \ \forall K \in \mathcal{T}_z \}$$



Choice of σ_i

$$\text{minimize } |\hat{\mathbf{w}}_i - \frac{1}{\Lambda_i + \gamma} T \tilde{u}_i|_{\mathcal{B}}^2 = \left\| \sigma_i - \frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma} \right\|_0^2 + \frac{1}{\gamma} \left\| \text{div } \sigma_i + \frac{\Lambda_i \tilde{u}_i}{\Lambda_i + \gamma} \right\|_0^2$$

(iii) Local constraint minimization:

[Braess, Schöberl 2000], [Ern, Vohralík 2013]

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

Find $\sigma_{z,i} \in \mathbf{W}_z$, $q_{z,i} \in Q_z$, $i = 1, 2, \dots, N$

$$\begin{aligned} (\sigma_{z,i}, \mathbf{w}_h)_{\omega_z} + (q_{z,i}, \text{div } \mathbf{w}_h)_{\omega_z} &= \left(\psi_z \frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma}, \mathbf{w}_h \right)_{\omega_z} & \forall \mathbf{w}_h \in \mathbf{W}_z \\ (\text{div } \sigma_{z,i}, \varphi_h)_{\omega_z} &= \left(-\frac{\psi_z \Lambda_i \tilde{u}_i}{\Lambda_i + \gamma}, \varphi_h \right)_{\omega_z} + \left(\frac{\nabla \psi_z \cdot \nabla \tilde{u}_i}{\Lambda_i + \gamma}, \varphi_h \right)_{\omega_z} \\ & & \forall \varphi_h \in Q_z \end{aligned}$$

$$\mathbf{W}_z = \{ \sigma_z \in \mathbf{H}(\text{div}, \omega_z) : \sigma_z|_K \in \mathbf{RT}_p(K) \ \forall K \in \mathcal{T}_z \text{ and } \sigma_z \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^{\text{E}} \}$$

$$Q_z = \{ q_z \in L^2(\omega_z) : q_z|_K \in P_p(K) \ \forall K \in \mathcal{T}_z \}$$



Choice of σ_i

$$\text{minimize } |\hat{\mathbf{w}}_i - \frac{1}{\Lambda_i + \gamma} T \tilde{u}_i|_{\mathcal{B}}^2 = \left\| \sigma_i - \frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma} \right\|_0^2 + \frac{1}{\gamma} \left\| \text{div } \sigma_i + \frac{\Lambda_i \tilde{u}_i}{\Lambda_i + \gamma} \right\|_0^2$$

(iv) Local unconstrained minimization:

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

$$\text{minimizing } \left\| \sigma_{z,i} - \psi_z \frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma} \right\|_{0, \omega_z}^2 + \frac{1}{\gamma} \left\| \text{div } \sigma_{z,i} + \frac{\psi_z \Lambda_i \tilde{u}_i}{\Lambda_i + \gamma} \right\|_{0, \omega_z}^2$$

$$\begin{aligned} (\sigma_{z,i}, \mathbf{w}_h)_{\omega_z} + \frac{1}{\gamma} (\text{div } \sigma_{z,i}, \text{div } \mathbf{w}_h)_{\omega_z} \\ = \left(\psi_z \frac{\nabla \tilde{u}_i}{\Lambda_i + \gamma}, \mathbf{w}_h \right)_{\omega_z} - \frac{1}{\gamma} \left(\frac{\psi_z \Lambda_i \tilde{u}_i}{\Lambda_i + \gamma}, \text{div } \mathbf{w}_h \right)_{\omega_z} \quad \forall \mathbf{w}_h \in \mathbf{W}_z \end{aligned}$$

$$\mathbf{W}_z = \{ \sigma_z \in \mathbf{H}(\text{div}, \omega_z) : \sigma_z|_K \in \mathbf{RT}_p(K) \quad \forall K \in \mathcal{T}_z \text{ and } \sigma_z \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^E \}$$



Remark 1. Weinstein and Kato bounds

Set $\eta_i = \|\nabla u_{h,i} - (\Lambda_{h,i} + \gamma)\sigma_{h,i}\|_0$ $i = 1, 2, \dots, N$
 $\sigma_{h,i}$ computed by (i) or (iii)

$$\text{Weinstein bound: } \ell_i^W = \frac{1}{4} \left(-\eta_i + \sqrt{\eta_i^2 + 4\Lambda_{h,i}} \right)^2$$

$$\text{Kato bound: } \ell_i^K = \Lambda_{h,i} \left(1 + \nu \Lambda_{h,i} \sum_{j=i}^N \frac{\eta_j^2}{\Lambda_{h,j}^2 (\nu - \Lambda_{h,j})} \right)^{-1}$$

where $\Lambda_{h,N} < \nu$

Theorem 1.

If $\sqrt{\lambda_{i-1}\lambda_i} \leq \Lambda_{h,i} \leq \sqrt{\lambda_i\lambda_{i+1}}$ then $\ell_i^W \leq \lambda_i$.

Theorem 2.

If $\nu \leq \lambda_{N+1}$ then $\ell_i^K \leq \lambda_i$ for all $i = 1, 2, \dots, N$.

[Vejchodský, Šebestová 2017]

Remark 2. Adaptive mesh refinement



Residual

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

Theorem

$$\|\nabla \varrho_i\|_0 \leq \eta_i, \quad \text{where } \eta_i = \|\nabla u_{h,i} - (\Lambda_{h,i} + \gamma)\sigma_{h,i}\|_0.$$

Local error indicators for mesh refinement

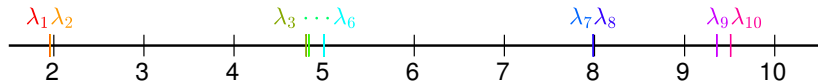
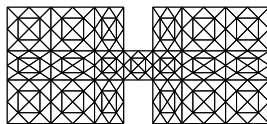
$$\eta_{i,K} = \|\nabla u_{h,i} - (\Lambda_{h,i} + \gamma)\sigma_{h,i}\|_{0,K} \quad \forall K \in \mathcal{T}_h$$

$\sigma_{h,i}$ computed by (i) or (iii)



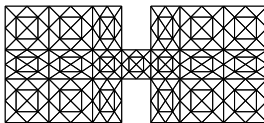
Example: Dumbbell shaped domain

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

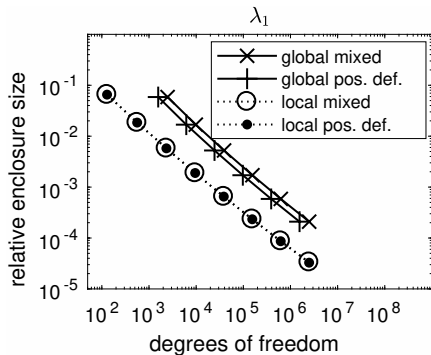
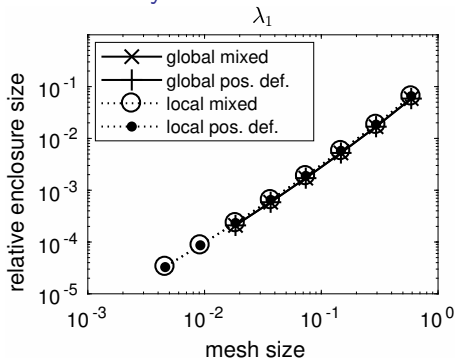


Example: Dumbbell shaped domain

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



Uniformly refined meshes:

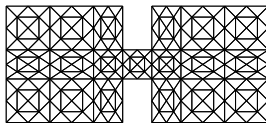


- ▶ $(\Lambda_{h,i} - \ell_i) / \ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

Example: Dumbbell shaped domain

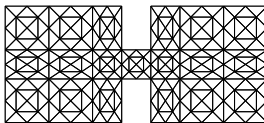


$$\begin{aligned} -\Delta u_j &= \lambda_j u_j && \text{in } \Omega \\ u_j &= 0 && \text{on } \partial\Omega \end{aligned}$$

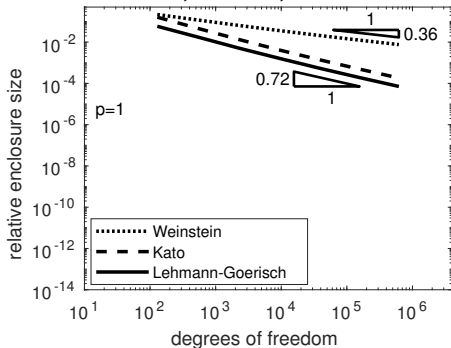


Example: Dumbbell shaped domain

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



Uniform, dumbbell, lambda1

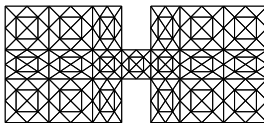


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

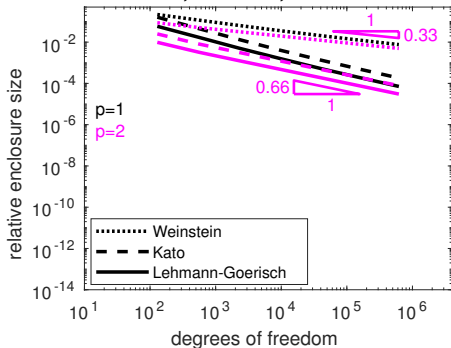
Example: Dumbbell shaped domain

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

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



Uniform, dumbbell, lambda1

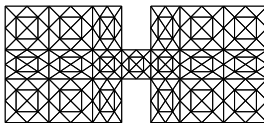


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

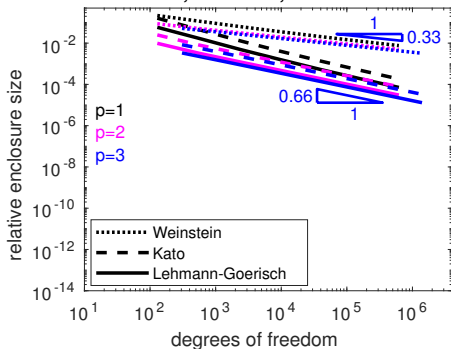
Example: Dumbbell shaped domain

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

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



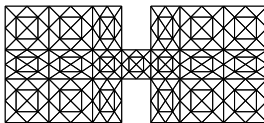
Uniform, dumbbell, lambda1



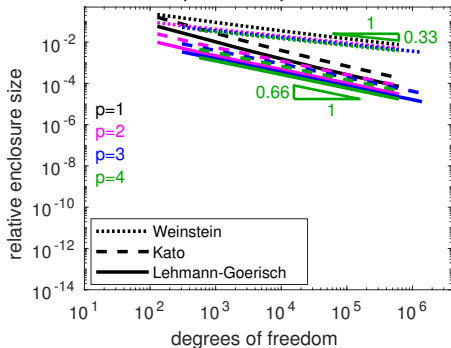
- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

Example: Dumbbell shaped domain

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



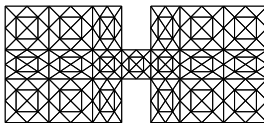
Uniform, dumbbell, lambda1



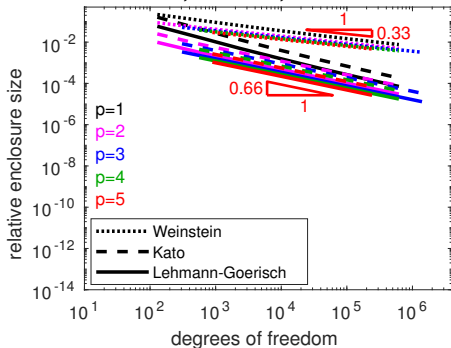
- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

Example: Dumbbell shaped domain

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



Uniform, dumbbell, lambda1

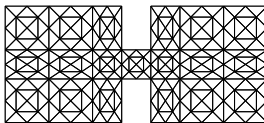


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

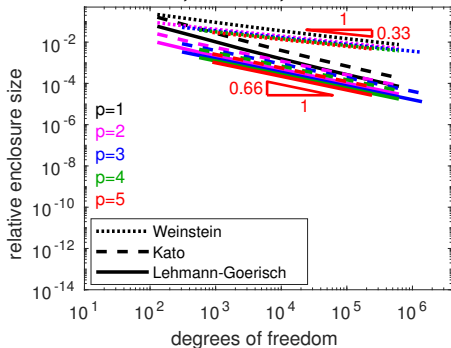


Example: Dumbbell shaped domain

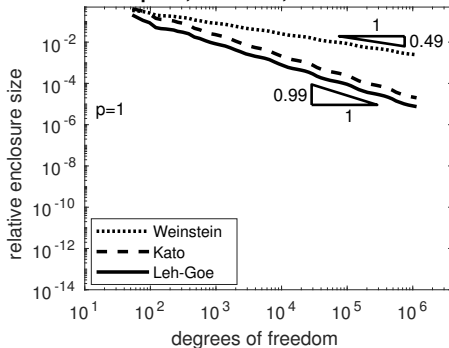
$$\begin{aligned} -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\ u_i &= 0 && \text{on } \partial\Omega \end{aligned}$$



Uniform, dumbbell, lambda1



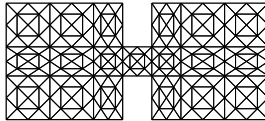
Adaptive, dumbbell, lambda1



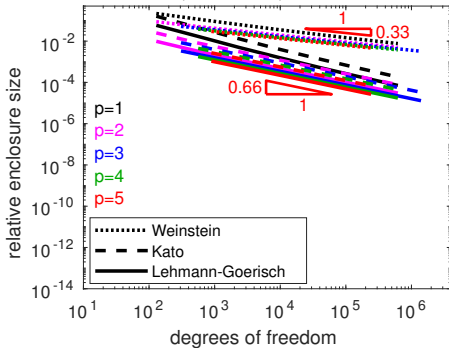
- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

Example: Dumbbell shaped domain

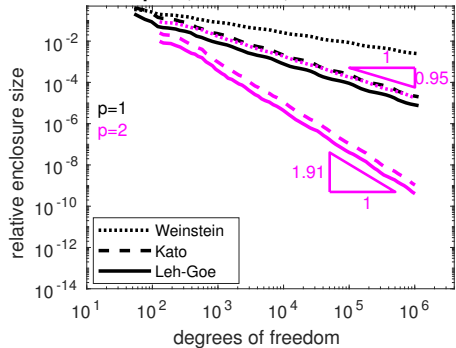
$$\begin{aligned}
 -\Delta u_i &= \lambda_i u_i && \text{in } \Omega \\
 u_i &= 0 && \text{on } \partial\Omega
 \end{aligned}$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

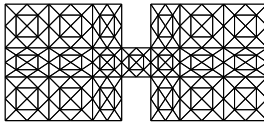


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

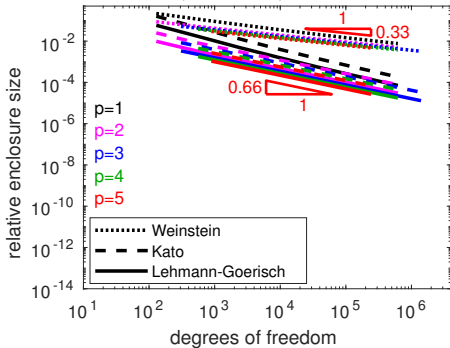
Example: Dumbbell shaped domain

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

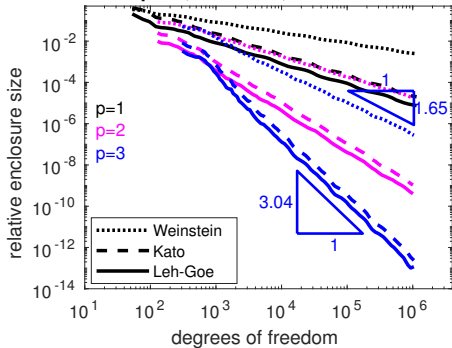
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

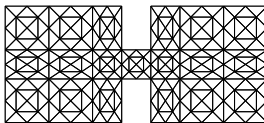


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

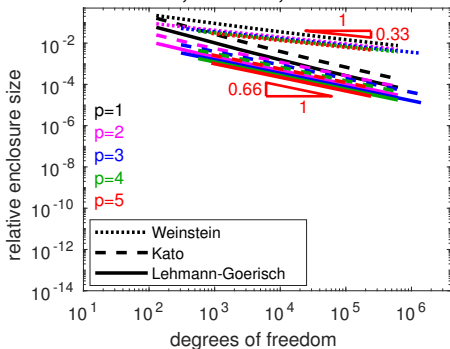
Example: Dumbbell shaped domain

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

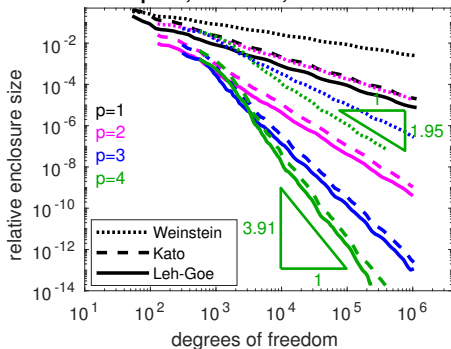
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

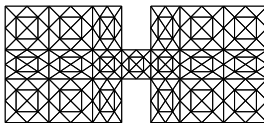


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

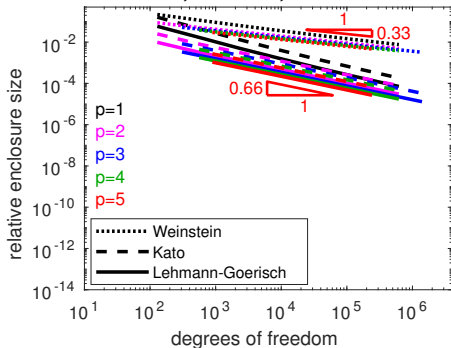
Example: Dumbbell shaped domain

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

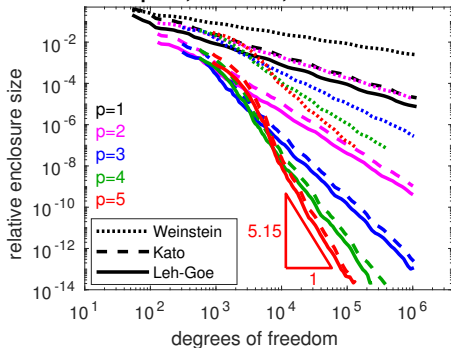
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

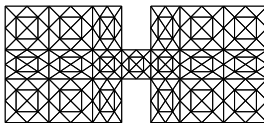


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method

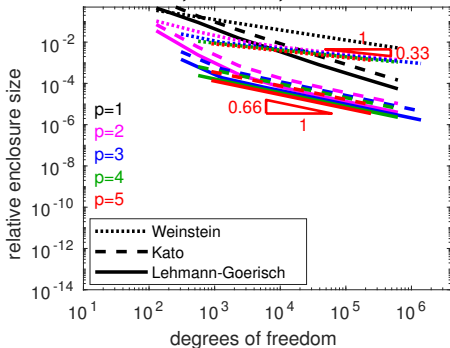
Example: Dumbbell shaped domain

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

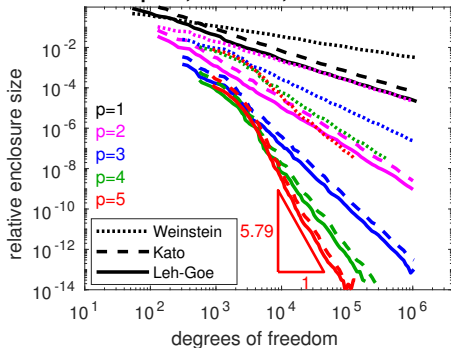
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda5



Adaptive, dumbbell, lambda5

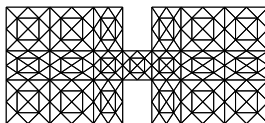


- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i)/\ell_i$
- ▶ $\gamma = 10^{-6}$, a priori lower bound ρ by Weinstein's method



Example: Dumbbell shaped domain

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



Tight pairs of eigenvalues:

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

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

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

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



- ▶ Flux reconstructions for source problems
⇒ good for eigenvalue problems
- ▶ Savings in memory requirements
- ▶ Parallelization

Flexibility:

- ▶ General symmetric elliptic operators
- ▶ Higher-order approximations
- ▶ Adaptivity