

Higher order lower bounds on eigenvalues of symmetric elliptic operators

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Lower bounds on eigenvalues



Laplace eigenvalue problem

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

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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:

$$V = H_0^1(\Omega)$$

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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:

$$V = H_0^1(\Omega)$$

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

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Upper bound:

$$\lambda_i \leq \Lambda_{h,i}$$

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Can we compute lower bound?

$$\ell_i \leq \lambda_i \leq \Lambda_{h,i} \quad \Rightarrow \quad |\Lambda_{h,i} - \lambda_i| \leq \Lambda_{h,i} - \ell_i$$

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Old problem:

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

Many results: M.G. Armentano, G. Barrenechea, H. Behnke,
C. Carstensen, R.G. Duran, D. Galistl, J. Gedicke, 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*

Nonconforming:

$$\frac{\lambda_{h,i}^{\text{CR}}}{1 + 0.1893^2 h^2 \lambda_{h,i}^{\text{CR}}} \leq \lambda_i \quad (\text{Crouzeix–Raviart, triangles})$$

- ▶ no a priori information on spectrum needed
- ▶ first order only

[Carstensen, Gallistl 2013], [Carstensen, Gedicke 2014]
[Xuefeng Liu 2015]

Conforming:

- ▶ a priori information on spectrum needed
- ▶ higher order versions

[Behnke, Mertins, Plum, Wieners 2000]
[Cancès, Dusson, Maday, Stamm, Vohralík 2017], [Vejchodský, Šebestová 2017]



Flux reconstruction in $\mathbf{H}(\text{div}, \Omega)$

$$\boldsymbol{\sigma}_{h,i} \approx \nabla u_i, \quad i = 1, 2, \dots, n$$

(a) Global problem:

$\boldsymbol{\sigma}_{h,i} \in \mathbf{W}_h$ minimizes $\|\nabla u_{h,i} - \boldsymbol{\sigma}_{h,i}\|_{L^2(\Omega)}^2$
under constraint: $-\text{div } \boldsymbol{\sigma}_{h,i} = \Lambda_{h,i} u_{h,i}$

Spaces:

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



Flux reconstruction in $\mathbf{H}(\text{div}, \Omega)$

$$\sigma_{h,i} \approx \nabla u_i, \quad i = 1, 2, \dots, n$$

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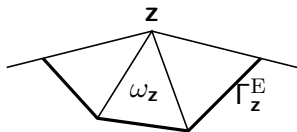
$$\text{under constraint: } -\text{div } \sigma_{h,i} = \Lambda_{h,i} u_{h,i}$$

(b) Local problems:

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

$$\sigma_{z,i} \in \mathbf{W}_z \text{ minimize } \|\psi_z \nabla u_{h,i} - \sigma_{z,i}\|_{L^2(\omega_z)}^2$$

$$\text{under constraint: } -\text{div } \sigma_{z,i} = \Lambda_{h,i} \Pi(\psi_z u_{h,i}) - \nabla \psi_z \cdot \nabla u_{h,i}$$



Spaces:

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

$$\mathbf{W}_z = \{\sigma_z \in \mathbf{H}(\text{div}, \omega_z) : \sigma_z|_K \in \mathbf{RT}_p(K) \forall K \in \mathcal{T}_z$$

$$\text{and } \sigma_z \cdot \mathbf{n}_z = 0 \text{ on } \Gamma_z^E\}$$



Weinstein and Kato bounds

Set $\eta_i = \|\nabla u_{h,i} - \sigma_{h,i}\|_{L^2(\Omega)}$ $i = 1, 2, \dots, n$

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

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]

Let $\gamma > 0$, $\nu \leq \lambda_{n+1}$, and $\tilde{\sigma}_{h,i} = (\Lambda_{h,i} + \gamma)^{-1} \sigma_{h,i}$

For $m = n, n-1, \dots, 2, 1$ do

- ▶ $\rho = \nu + \gamma$
- ▶ $\mathbf{M}_{ij} = (\nabla u_{h,i}, \nabla u_{h,j}) + (\gamma - \rho)(u_{h,i}, u_{h,j})$
- ▶ $\mathbf{N}_{ij} = (\nabla u_{h,i}, \nabla u_{h,j}) + (\gamma - 2\rho)(u_{h,i}, u_{h,j}) + \rho^2(\tilde{\sigma}_{h,i}, \tilde{\sigma}_{h,j}) + (\rho^2/\gamma)(u_{h,i} + \operatorname{div} \tilde{\sigma}_{h,i}, u_{h,j} + \operatorname{div} \tilde{\sigma}_{h,j})$
- ▶ $\mu_1 \leq \dots \leq \mu_m$: $\mathbf{M}\mathbf{y}_i = \mu_i \mathbf{N}\mathbf{y}_i$, $i = 1, 2, \dots, m$
- ▶ If \mathbf{N} is s.p.d. and if $\mu_{m+1-j} < 0$ then
$$\ell_{j,m}^* = \rho - \gamma - \rho / (1 - \mu_{m+1-j}) \leq \lambda_j, \quad j = 1, 2, \dots, m$$
- ▶ $\ell_m^{\text{LG}} = \max\{\ell_{m,i}^*, i = m, m+1, \dots, n\} \leq \lambda_m$
- ▶ $\nu = \ell_m^{\text{LG}}$

end for

Theorem

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



Residual

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

Theorem

$$\|\nabla w_i\|_{L^2(\Omega)} \leq \eta_i, \quad \text{where } \eta_i = \|\nabla u_{h,i} - \sigma_{h,i}\|_{L^2(\Omega)}.$$

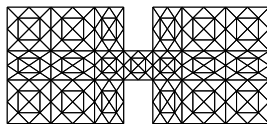
Local error indicators for mesh refinement

$$\eta_{i,K} = \|\nabla u_{h,i} - \sigma_{h,i}\|_{L^2(K)} \quad \forall K \in \mathcal{T}_h$$



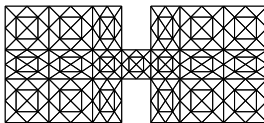
Example: Dumbbell shaped domain

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

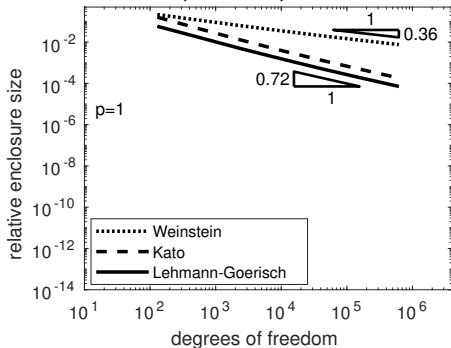


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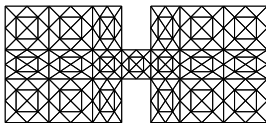
Uniform, dumbbell, lambda1



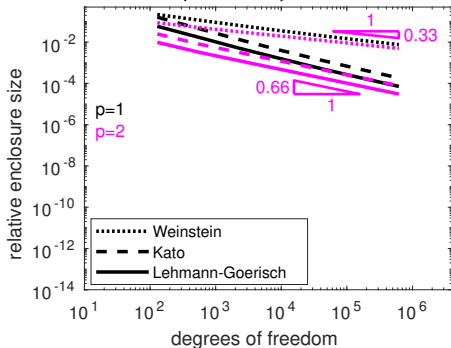
- ▶ relative enclosure size: $(\Lambda_{h,i} - \ell_i) / \ell_i$
- ▶ $\gamma = 10^{-6}$, $\nu := \ell_{11}^W \leq \lambda_{11} \approx 10.0017$

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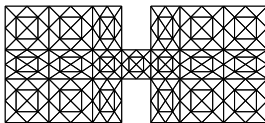
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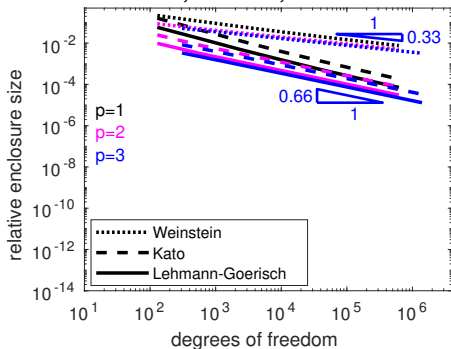
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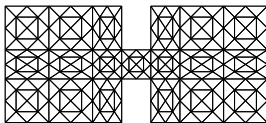


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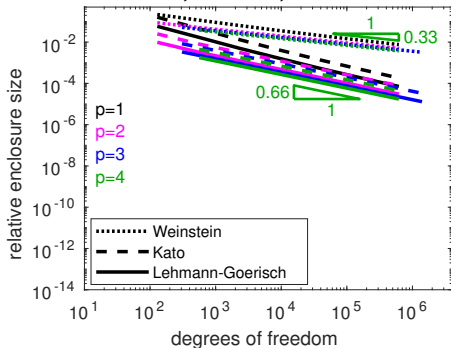
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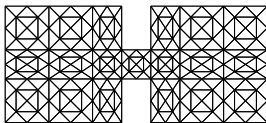
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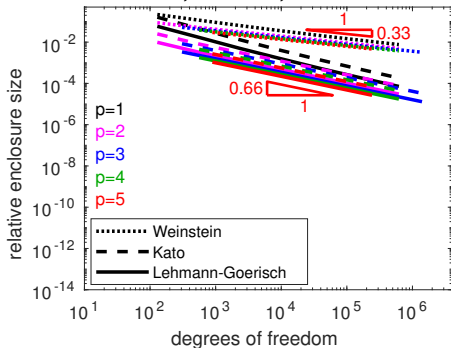
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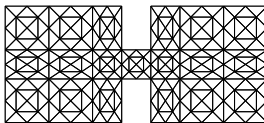
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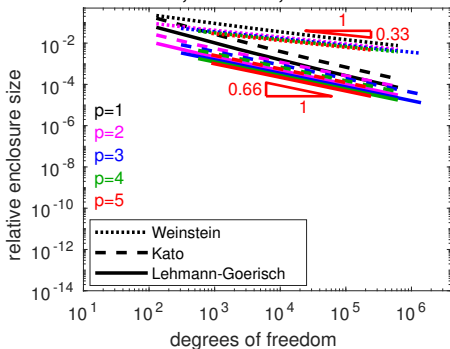
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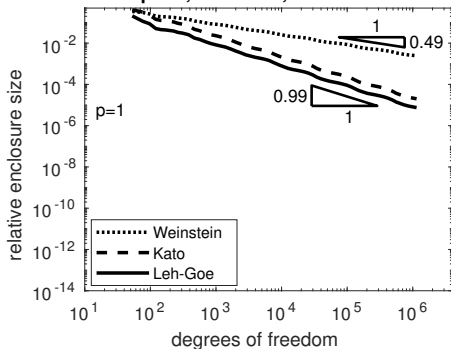
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Adaptive, dumbbell, lambda1

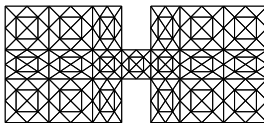


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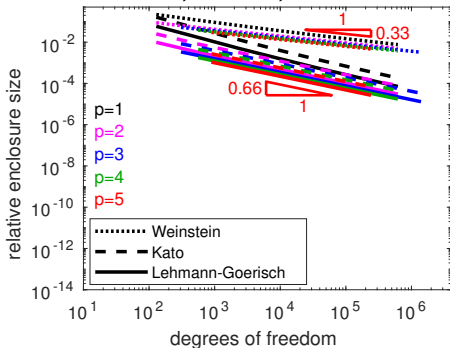
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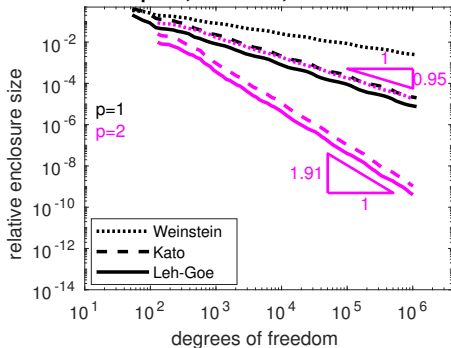
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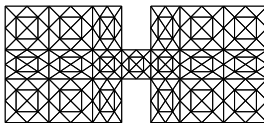


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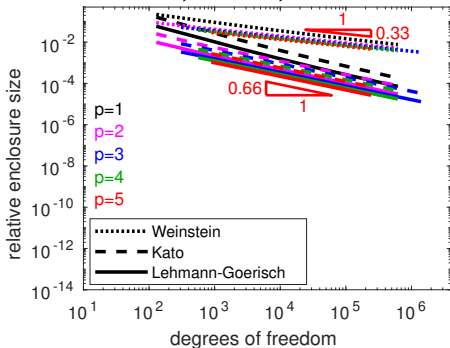


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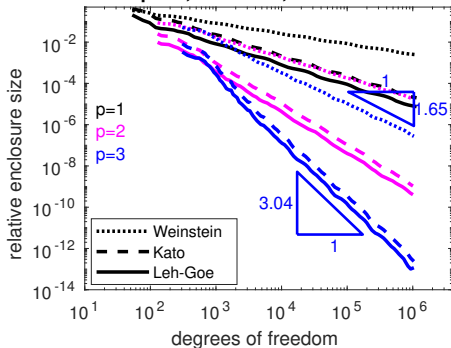
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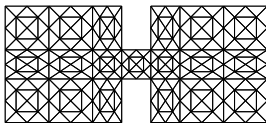
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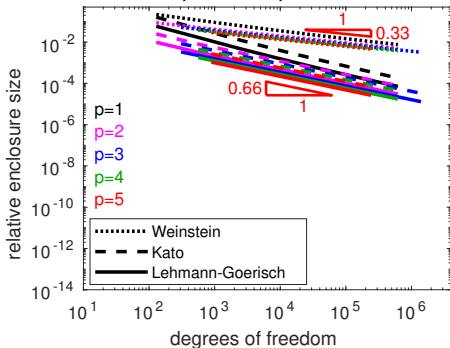
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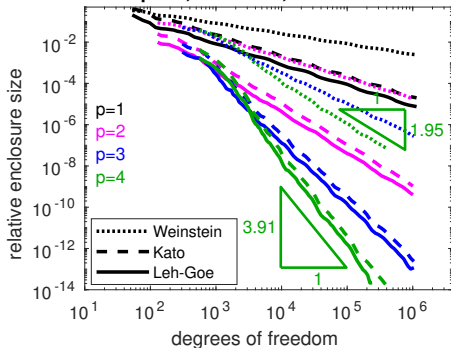
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Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

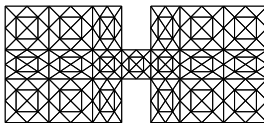


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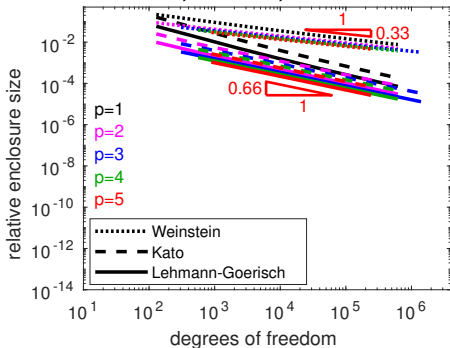
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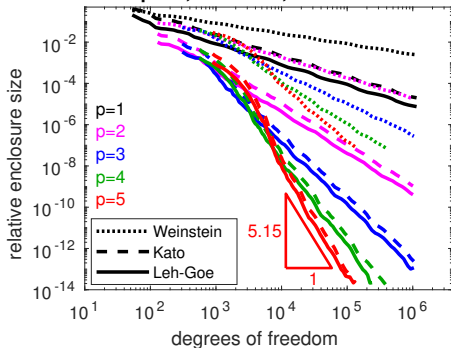
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Uniform, dumbbell, lambda1



Adaptive, dumbbell, lambda1

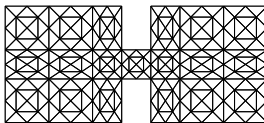


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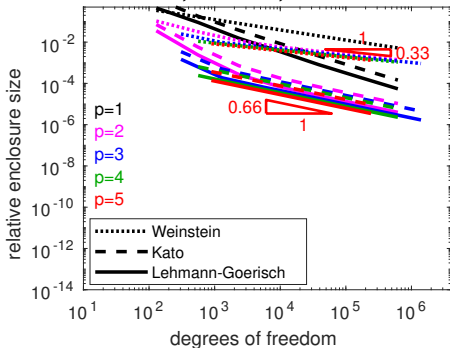
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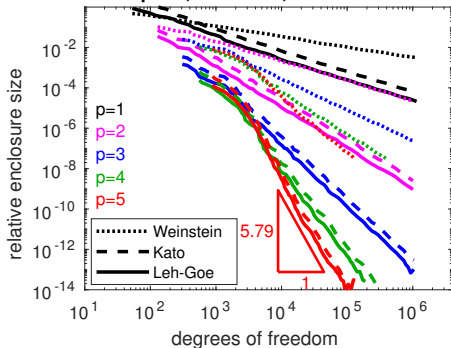
$$u_i = 0 \quad \text{on } \partial\Omega$$



Uniform, dumbbell, lambda5



Adaptive, dumbbell, lambda5

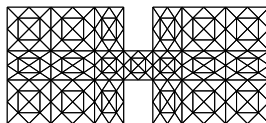


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Example: Dumbbell shaped domain

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Tight pairs of eigenvalues:

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

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

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

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

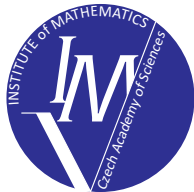


- ▶ Kato and Lehmann–Goerisch methods provide lower bounds with optimal rates of convergence even for higher-order approximations
- ▶ The same flux reconstruction can be used in all methods and it can be used for adaptive mesh refinement
- ▶ Weinstein bound has suboptimal convergence rate, but it is useful for a priori lower bounds

Thank you for your attention

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