

# Computing lower bounds of eigenvalues by the finite element method

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# Abstract eigenvalue problem



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

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

$a$  ... symmetric, continuous,  $V$ -elliptic

$b$  ... symmetric, continuous

Solution operator:

$$S : V \mapsto V$$

$Sz = y$ , where  $y \in V$ :

$$a(y, v) = b(z, v) \quad \forall v \in V$$

Assumption:  $S$  is compact

# Symmetric elliptic eigenvalue problem



Find  $\lambda_j \in \mathbb{R}$  and  $u_j \neq 0$ :

$$\begin{aligned} -\operatorname{div}(\mathcal{A}\nabla u_j) + cu_j &= \lambda_j \beta_1 u_j && \text{in } \Omega \\ (\mathcal{A}\nabla u_j) \cdot \mathbf{n} + \alpha u_j &= \lambda_j \beta_2 u_j && \text{on } \Gamma_N \\ u_j &= 0 && \text{on } \Gamma_D \end{aligned}$$

Weak formulation:

- ▶  $V = \{v \in H^1(\Omega) : v = 0 \text{ on } \Gamma_D\}$
- ▶  $a(u, v) = (\mathcal{A}\nabla u, \nabla v) + (cu, v) + (\alpha u, v)_{\Gamma_N}$
- ▶  $b(u, v) = (\beta_1 u, v) + (\beta_2 u, v)_{\Gamma_N}$

Assumptions:

- ▶  $\mathcal{A} \in \mathbb{R}^{d \times d}$  uniformly positive definite,  $c \geq 0$ ,  $\alpha \geq 0$
- ▶  $\mathcal{A}$ ,  $c$ ,  $\alpha$ ,  $\beta_1$ ,  $\beta_2$  piecewise constant



Find  $\lambda_i^h \in \mathbb{R}$  and  $u_i^h \in V^h$ ,  $u_i^h \neq 0$ :

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

where

$$\triangleright V^h = \{v^h \in V : v^h|_K \in P^1(K) \quad \forall K \in \mathcal{T}_h\}$$

Courant–Fischer–Weyl min-max principle:

$$\lambda_i \leq \lambda_i^h$$

Babuška, Osborn (1989):

$$|\lambda_i - \lambda_i^h| \leq Ch^2$$

$$\|u_i - u_i^h\| \leq Ch$$



Lower bound  $\underline{\lambda}_j^h \leq \lambda_j$  ?

(A) Kuttler, Sigillito (1978):

If  $S$  compact,  $\lambda_i^h \in \mathbb{R}$ ,  $u_i^h \in V$ , and  $|u_i^h|_b = 1$ .

$w_i \in V$ :  $a(w_i, v) = a(u_i^h, v) - \lambda_i^h b(u_i^h, v) \quad \forall v \in V$

Then

$$\min_j \left| \frac{\lambda_j - \lambda_i^h}{\lambda_j} \right| \leq |w_i|_b$$

(B) Friedrichs–Poincaré inequality:

$$|w_i|_b \leq C_{ab} \|w_i\|_a, \quad C_{ab} = \frac{1}{\sqrt{\lambda_1}}$$

(C) Complementarity estimate:

$$\|w_i\|_a \leq A_i + C_{ab} B_i$$

Hence

$$\min_j \left| \frac{\lambda_j - \lambda_i^h}{\lambda_j} \right| \leq \frac{1}{\sqrt{\lambda_1}} A_i + \frac{1}{\lambda_1} B_i$$



## Lower bound $\underline{\lambda}_i^h \leq \lambda_i$

Relative closeness assumption:

Let the (relatively) closest eigenvalue to  $\lambda_i^h$  be  $\lambda_i$ .

Case 1:  $i = 1$

$$\frac{\lambda_1^h - \lambda_1}{\lambda_1} = \min_j \left| \frac{\lambda_j - \lambda_1^h}{\lambda_j} \right| \leq \frac{1}{\sqrt{\lambda_1}} A_1 + \frac{1}{\lambda_1} B_1$$

$\Rightarrow$

$$\lambda_1^h \leq \lambda_1, \quad \underline{\lambda}_1^h = \frac{1}{4} \left( -A_1 + \sqrt{A_1^2 + 4(\lambda_1^h - B_1)} \right)^2$$

Case 2:  $i > 1$

$$\frac{\lambda_i^h - \lambda_i}{\lambda_i} = \min_j \left| \frac{\lambda_j - \lambda_i^h}{\lambda_j} \right| \leq \frac{1}{\sqrt{\lambda_1^h}} A_i + \frac{1}{\lambda_1^h} B_i$$

$\Rightarrow$

$$\underline{\lambda}_i^h \leq \lambda_i, \quad \underline{\lambda}_i^h = \left( 1 + \frac{A_i}{\sqrt{\lambda_1^h}} + \frac{B_i}{\lambda_1^h} \right)^{-1} \lambda_i^h$$



## Complementarity estimate:

Flux reconstruction  $\mathbf{q}_i^h$ :

- ▶  $\mathbf{q}_i^h \in \mathbf{H}(\text{div}, \Omega)$ ,
- ▶  $\mathbf{q}_i^h|_K \in \text{RT}_1(K) \quad \forall K \in \mathcal{T}_h$
- ▶  $\mathbf{q}_i^h$  solves local problems on patches  $\omega_a$
- ▶  $\text{div } \mathbf{q}_i^h = cu_i^h - \lambda_i^h \beta_1 u_i^h \quad \text{in } \Omega$
- ▶  $\mathbf{q}_i^h \cdot \mathbf{n} = -\alpha u_i^h + \lambda_i^h \beta_2 u_i^h \quad \text{on } \Gamma_N$

Theorem

$$w_i \in V: a(w_i, v) = a(u_i^h, v) - \lambda_i^h b(u_i^h, v) \quad \forall v \in V$$

Then

$$\|w_i\|_a \leq A_i + C_{ab} B_i,$$

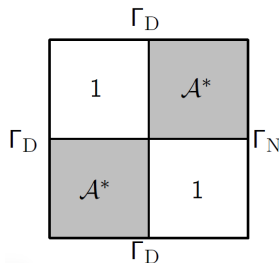
where

- ▶  $A_i = \|\nabla u_i^h - \mathcal{A}^{-1} \mathbf{q}_i^h\|_{\mathcal{A}}$
- ▶  $B_i = 0$

# Numerical example



$$\begin{aligned} -\operatorname{div}(\mathcal{A}\nabla u_i) &= \lambda_i u_i && \text{in } \Omega \\ (\mathcal{A}\nabla u_i) \cdot \mathbf{n} &= 0 && \text{on } \Gamma_N \\ u_i &= 0 && \text{on } \Gamma_D \end{aligned}$$



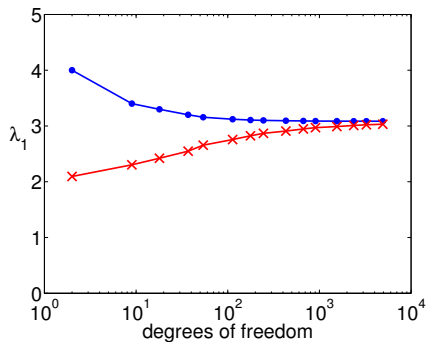
- ▶  $\Omega = (-1, 1)^2$
- ▶  $\mathcal{A} = \begin{cases} 1 & \text{for } xy \leq 0 \\ \mathcal{A}^* & \text{for } xy > 0 \end{cases}$
- ▶ Adaptive algorithm driven by  $\eta_K = \|\nabla u_i^h - \mathcal{A}^{-1} \mathbf{q}_i^h\|_{\mathcal{A}, K}$



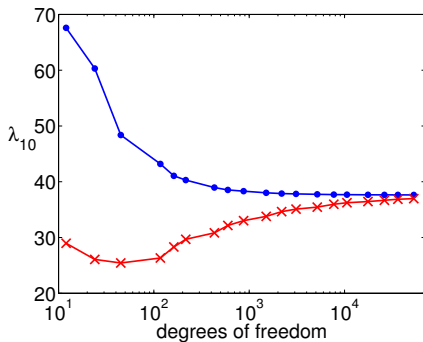
# Smooth case ( $\mathcal{A}^* = 1$ )



$3.03115 \leq \lambda_1 \leq 3.08512$



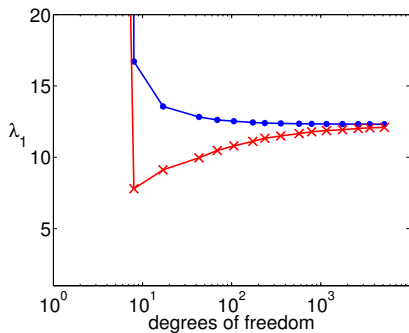
$36.9745 \leq \lambda_{10} \leq 37.6392$



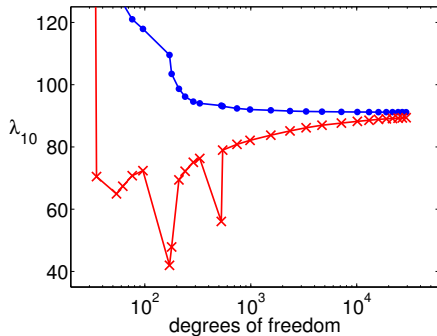
# Singularity ( $\mathcal{A}^* = 1000$ )



$12.0944 \leq \lambda_1 \leq 12.3214$



$89.3918 \leq \lambda_{10} \leq 91.1678$





## Highlights

- ▶ Upper bound by the (conforming) FEM
- ▶ Lower bound by postprocessing
- ▶ Mixed boundary conditions (includes Steklov eigenvalue problem)
- ▶ Efficiency of the error indicators
- ▶ Convergence of the adaptive algorithm

## Future work

- ▶ Indicator for the relative closeness assumption
- ▶ Estimates of the error in eigenvectors

Thank you for your attention

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