

Convergence of adaptive FEM for elliptic eigenvalue problems

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ZSS 08



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Model Problem

Let Ω be a polygonal domain bounded in \mathbb{R}^2

Problem: seek eigenpairs (λ, u) of the problem

$$\begin{cases} -\nabla \cdot \mathcal{A}(\nabla u) = \lambda \mathcal{B}u & \text{in } \Omega, \\ u = 0 & \text{on } \partial\Omega. \end{cases}$$

$$\underline{a} \leq \xi^T \mathcal{A}(x) \xi \leq \bar{a} \quad \text{for all } \xi \in \mathbb{R}^2 \text{ with } |\xi| = 1 \text{ and for all } x \in \Omega, \\ \underline{b} \leq \mathcal{B}(x) \leq \bar{b} \quad \text{for all } x \in \Omega.$$

Variational Problem: seek eigenpairs $(\lambda, u) \in \mathbb{R} \times H_0^1(\Omega)$ such that

$$a(u, v) = \lambda (u, v)_{0, \mathcal{B}\Omega} \quad \text{for all } v \in H_0^1(\Omega),$$

where

$$a(u, v) := \int_{\Omega} \mathcal{A} \nabla u \cdot \nabla v \, dx, \quad \| \| u \| \|_{\Omega} := a(u, u)^{1/2}.$$

$$(u, v)_{0, \mathcal{B}, \Omega} := \int_{\Omega} \mathcal{B} u v \, dx.$$

Meshes and Discrete Problems

Define:

- \mathcal{T}_n conforming and shape regular triangulation of Ω
- \mathcal{F}_n is the set of the edges of the triangles in the interior of \mathcal{T}_n ,
- V_n space of piecewise linear functions over \mathcal{T}_n .

Problem: seek eigenpairs $(\lambda_n, u_n) \in \mathbb{R} \times V_n$ such that

$$a(u_n, v_n) = \lambda_n (u_n, v_n)_{0, \mathcal{B}, \Omega} \quad \text{for all } v_n \in V_n.$$

Theorem (Reliability for eigenfunctions)

Let (λ, u) be a simple eigenvalue of the continuous problem and let (λ_n, u_n) be the correspondent computed eigenpair. Then we have for $e_n = u - u_n$ that

$$\|e_n\|_{\Omega} \leq C \eta_n + G_n,$$

where

$$G_n = \frac{1}{2}(\lambda + \lambda_n) \frac{(e_n, e_n)_{0,B,\Omega}}{\|e_n\|_{\Omega}}.$$

[R. Verfürth (1996). *A Review of Posteriori Error Estimation and Adaptive Mesh Refinement Techniques*. Wiley-Teubner]

Definition (Jump)

$$[g]_f(x) := \left(\lim_{\substack{\tilde{x} \in \tau_1(f) \\ \tilde{x} \rightarrow x}} g(\tilde{x}) - \lim_{\substack{\tilde{x} \in \tau_2(f) \\ \tilde{x} \rightarrow x}} g(\tilde{x}) \right), \quad \text{with } x \in f.$$

Definition (Residual)

$$R_I(u, \lambda)(x) := (\nabla \cdot \mathcal{A} \nabla u + \lambda B u)(x), \quad \text{with } x \in \text{int}(\tau), \quad \tau \in \mathcal{T}_n,$$

$$R_F(u)(x) := [\vec{n}_f \cdot \mathcal{A} \nabla u]_f(x), \quad \text{with } x \in \text{int}(f), \quad f \in \mathcal{F}_n.$$

$$\eta_n := \left\{ \sum_{\tau \in \mathcal{T}_n} H_\tau^2 \|R_I(u_n, \lambda_n)\|_{0,\tau}^2 + \sum_{f \in \mathcal{F}_n} H_f \|R_F(u_n)\|_{0,f}^2 \right\}^{1/2},$$

Theorem (Reliability for eigenfunctions)

Let (λ, u) be a simple eigenvalue of the continuous problem and let (λ_n, u_n) be the correspondent computed eigenpair. Then we have for $e_n = u - u_n$ that

$$\|e_n\|_{\Omega} \leq C \eta_n + G_n,$$

where

$$G_n = \frac{1}{2}(\lambda + \lambda_n) \frac{(e_n, e_n)_{0,B,\Omega}}{\|e_n\|_{\Omega}}.$$

Theorem (Reliability for eigenvalues)

Let (λ, u) be a simple eigenvalue of the continuous problem and let (λ_n, u_n) be the correspondent computed eigenpair. Then we have for $e_n = u - u_n$ that

$$|\lambda_n - \lambda| \leq C' \eta_n^2 + G'_n,$$

where

$$G'_n = \eta_n \frac{1}{2} (\lambda + \lambda_n) \frac{(e_n, e_n)_{0,B,\Omega}}{\|e_n\|_{\Omega}} + \frac{1}{2} (\lambda - \lambda_n) (e_n, e_n)_{0,B,\Omega}.$$

Theorem (Efficiency)

Let (λ, u) be a simple eigenpair of the continuous problem and let (λ_n, u_n) be the corresponding computed eigenpairs. Then we have that the global residual estimator is bounded by the energy norm of the error as:

$$\eta_n \leq C' \left(\|e_n\|_{\Omega} + \|H_{\tau}(\lambda_n u_n - \lambda u)\|_{0,B,\Omega} \right).$$

Marking Strategy 1

Set the parameter $0 < \theta < 1$:

mark the sides in a minimal subset $\hat{\mathcal{T}}_n$ of \mathcal{T}_n such that

$$\left(\sum_{\tau \in \hat{\mathcal{T}}_n} \eta_{\tau,n}^2 \right)^{1/2} \geq \theta \eta_n,$$

where

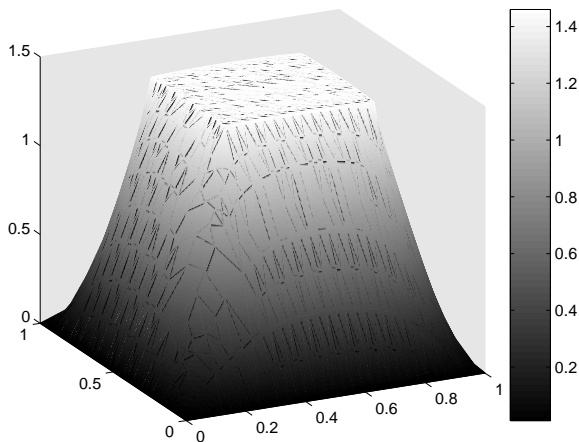
$$\eta_{\tau,n}^2 := H_{\tau}^2 \|R_I(u_n, \lambda_n)\|_{0,\tau}^2 + \sum_{f \in \partial\tau} \frac{1}{2} H_f \|R_F(u_n)\|_{0,f}^2.$$

Mesh Adaptivity Algorithm

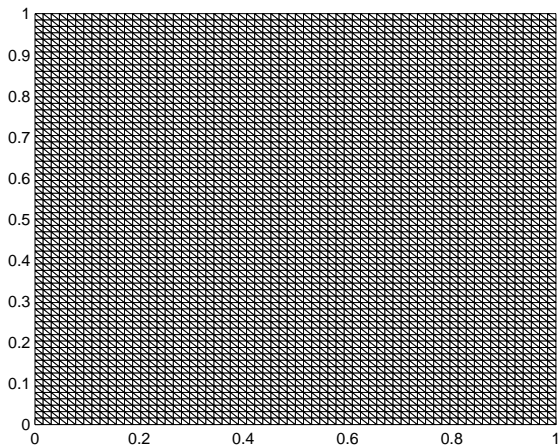
- 1 Set tol , $0 < \theta < 1$ and \mathcal{T}_0
- 2 Loop
- 3 Compute (λ_n, u_n) on \mathcal{T}_n
- 4 Marking strategy 1
- 5 Refine the mesh
- 6 End Loop if $\eta_n \leq \text{tol}$

Solution

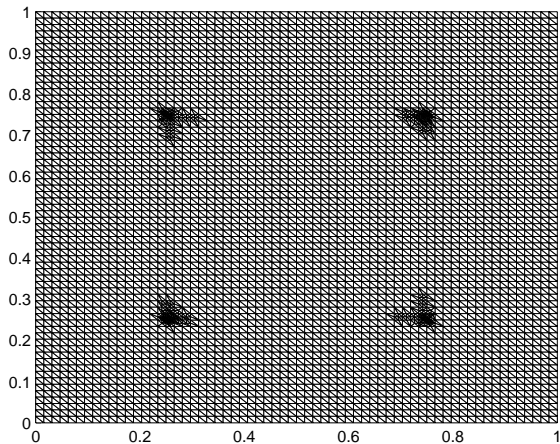
$$\mathcal{B} = 1, \quad \mathcal{A} = \{1, 10000\}.$$



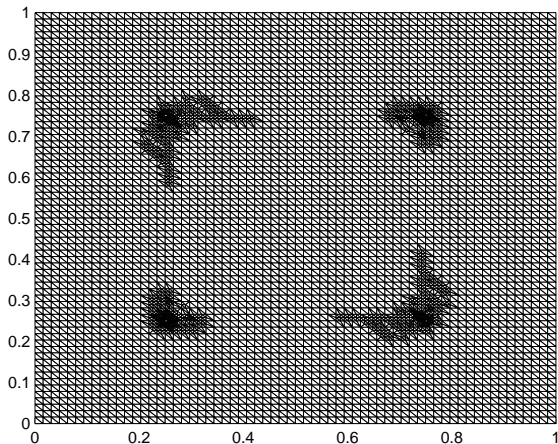
Uniform Mesh



Refined Mesh



Refined Mesh



$$\mathcal{B} = 1, \quad \mathcal{A} = \{1, 10000\}.$$

| Uniform | | | Adaptive | | | | | |
|-------------------------|-------|-----|-------------------------|-------|-----|-------------------------|-------|-----|
| | | | $\theta = 0.5$ | | | $\theta = 0.8$ | | |
| $ \lambda - \lambda_n $ | N | n | $ \lambda - \lambda_n $ | N | n | $ \lambda - \lambda_n $ | N | n |
| 1.1071 | 81 | 1 | 1.1071 | 81 | 1 | 1.1071 | 81 | 1 |
| 0.3521 | 289 | 2 | 0.2766 | 673 | 5 | 0.2244 | 799 | 3 |
| 0.1168 | 1089 | 3 | 0.0948 | 2080 | 8 | 0.0990 | 2235 | 4 |
| 0.0399 | 4225 | 4 | 0.0315 | 6039 | 11 | 0.0180 | 12375 | 6 |
| 0.0136 | 16641 | 5 | 0.0148 | 12607 | 13 | 0.0065 | 29148 | 7 |
| 0.0042 | 66049 | 6 | 0.0038 | 37126 | 16 | 0.0020 | 65387 | 8 |

Table: Comparison between uniform refinement and the adaptive method for the smallest eigenvalue of the generic elliptic eigenvalue problem with discontinuous coefficients.

[P. Morin, R. H. Nochetto and K. G. Siebert (2000). Data oscillation and convergence of adaptive FEM. *SIAM J. Numer. Anal.* 38]

Oscillations:

$$\text{osc}(v_n, \mathcal{T}_n) := \left(\sum_{\tau \in \mathcal{T}_n} \|H_\tau(v_n - P_n v_n)\|_\tau^2 \right)^{1/2},$$

where $(P_n v_n)|_\tau := \frac{1}{|\tau|} \int_\tau v_n$

Marking Strategy 2

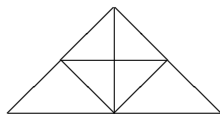
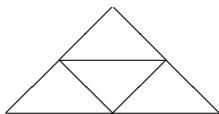
Set the parameter $0 < \tilde{\theta} < 1$:

mark the sides in a minimal subset $\tilde{\mathcal{T}}_n$ of \mathcal{T}_n such that

$$\text{osc}(u_n, \tilde{\mathcal{T}}_n) \geq \tilde{\theta} \text{osc}(u_n, \mathcal{T}_n).$$

Then we take the union of $\hat{\mathcal{T}}_n \cup \tilde{\mathcal{T}}_n$ and we refine all the elements in the union.

Bisection5



- 1 Split any edge
- 2 Split one of the new edge

PROS:

- 1 A new node on each edge
- 2 A new node in the interior of the element

Mesh Adaptivity Algorithm

- 1 Require $0 < \theta < 1$, $0 < \tilde{\theta} < 1$ and \mathcal{T}_0
- 2 Loop
- 3 Compute (λ_n, u_n) on \mathcal{T}_n
- 4 Marking strategy 1
- 5 Marking strategy 2
- 6 Refine the mesh
- 7 End Loop

Convergence for Linear Problems

$$-\nabla \cdot (\mathcal{A}\nabla u) = f.$$

[P. Morin, R. H. Nochetto and K. G. Siebert (2000). Data oscillation and convergence of adaptive FEM. *SIAM J. Numer. Anal.* 38]

- **osc decreases refining the mesh.**

$$\text{osc}(f, \mathcal{T}_{n+1}) \leq \alpha \text{osc}(f, \mathcal{T}_n), \quad \text{with } \alpha < 1.$$

- **The decay of osc trigs the convergence in the energy norm.**

$$a(u_n, u_n)^{1/2} \leq C_0 p^n, \quad \text{with } p < 1.$$

Convergence for Eigenvalue Problems

$$-\nabla \cdot (\mathcal{A} \nabla u) = \beta \lambda u.$$

[S. Giani and I. G. Graham (2007). A convergent adaptive method for elliptic eigenvalue problems. *submitted for publication*]

- The decay of osc depends on the behavior of the error.

$$\text{osc}(u_{n+1}, \mathcal{T}_{n+1}) \leq \tilde{\alpha} \text{osc}(u_n, \mathcal{T}_n) + C'' a(u - u_n, u - u_n)^{1/2}.$$

- The decay of the error depends on the behavior of osc .

$$\| \| u - u_{n+1} \| \|_{\Omega}^2 \leq \beta_n \| \| u - u_n \| \|_{\Omega}^2 + C''' \text{osc}(\lambda_n u_n, \mathcal{T}_n)^2.$$

- The error and the quantity osc are linked together.

If H_0 is small enough:

$$\| \| u - u_n \| \|_{\Omega} \leq C_0 p^n, \quad |\lambda - \lambda_n| \leq C_1 p^{2n},$$

$$\text{osc}(u_n, \mathcal{T}_n) \leq C_3 p^n, \quad \text{with } p < 1.$$