

Adaptive finite element methods in unbounded domains

T. Chaumont-Frelet^{*} and G. Gantner[†]

Robust adaptivity for nonlinear partial differential equations
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Introduction

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Solved questions (at least, partially)

Can we robustly estimate the error $\|\nabla(u - u_\ell)\|_\Omega$?

Is there a way to optimally design/produce the mesh \mathcal{T}_ℓ ?

A posteriori estimator

For all $K \in \mathcal{T}_\ell$ and $v_\ell \in \mathcal{P}_p(\mathcal{T}_\ell) \cap H_0^1(\Omega)$, we introduce

$$\eta_\ell^2(v_\ell, K) := h_K^2 \|f + \nabla \cdot (\mathbf{A} \nabla v_\ell)\|_K^2 + h_K \| \llbracket \mathbf{A} \nabla v_\ell \rrbracket \cdot \mathbf{n} \|_{\partial K \setminus \Gamma}^2.$$

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Reliability and efficiency

$$\|\nabla(\mathbf{u} - \mathbf{u}_\ell)\|_\Omega \lesssim \eta_\ell(\mathbf{u}_\ell, \mathcal{T}_\ell) \quad \eta_\ell(\mathbf{u}_\ell, K) \lesssim_p \|\nabla(\mathbf{u} - \mathbf{u}_\ell)\|_{\tilde{K}}$$

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These inequalities can be tightened to “ \leq and \lesssim_p ” using an equilibrated estimator.

Adaptive algorithm

Given an initial mesh \mathcal{T}_0 , for $\ell = 0, 1, \dots$, do

1. Compute the discrete solution u_ℓ associated with the mesh \mathcal{T}_ℓ .
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3. Find the elements $K \in \mathcal{M}_\ell \subset \mathcal{T}_\ell$ where $\eta_\ell(u_\ell, K)$ is large with Dörfler's marking.
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If there exists a sequence of meshes $\widehat{\mathcal{T}}_\ell$ reachable from \mathcal{T}_0 by finite NVBs, s.t.

$$\|\nabla(u - \widehat{u}_\ell)\|_\Omega \leq \widehat{C}|\widehat{\mathcal{T}}_\ell|^{-s}$$

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Optimal rate

If f is piecewise smooth and no anisotropic refinements are needed, then

$$\|\nabla(u - u_\ell)\|_\Omega \leq C|\mathcal{T}_\ell|^{-p/3},$$

which is the rate expected for smooth solutions, even if u is not smooth in general.

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Questions

How should we define the discrete solution u_ℓ ?

Can we robustly estimate the error $\|\nabla(u - u_\ell)\|_\Omega$?

Is there a way to optimally design/produce the mesh \mathcal{T}_ℓ ?

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We can then use the standard toolbox to approximate u^L by u_ℓ^L .

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Another issue is that it is hard to “balance” between ε and η in adaptive algorithms.

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This does implicitly introduce a truncation boundary Γ_ℓ .

However, we will never reference nor use the associated truncated PDE.

The truncation boundary Γ_ℓ does not need to be Lipschitz, and often is not.

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Main results in this talk

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Questions (with positive answers)

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Optimal rate

Under natural assumptions on f , if no anisotropic refinements are needed, then

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- 1 Precise setting
- 2 Residual based a posteriori estimator
- 3 Adaptive algorithm
- 4 Numerical examples

Precise setting

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Domain and coefficient

We assume that $\Omega \subset \mathbb{R}^3$ is a (possibly) unbounded polyhedron.
By this, we mean that it can be covered by (infinitely many) tetrahedra with diameters uniformly bounded from below.

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We assume that f and A are piecewise constant. This will be made precise later.

Precise setting
Variational formulation

For smooth compactly supported functions, the application

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The bilinear form $(\mathbf{A}\nabla\cdot, \nabla\cdot)_\Omega$ is naturally continuous and coercive over $\widehat{H}_0^1(\Omega)$.

If $v \in \widehat{H}_0^1(\Omega)$ we do have $\nabla v \in L^2(\Omega)$ and $v|_\Gamma = 0$.

We also have $v \rightarrow 0$ as $|\mathbf{x}| \rightarrow +\infty$ in an appropriate.

However, we do not need to have $v \in L^2(\Omega)$, so that $\widehat{H}_0^1(\Omega) \neq H_0^1(\Omega)$ in general.

The following “Poincaré–like” inequality is paramount.

Hardy's inequality

For all $v \in \widehat{H}_0^1(\mathbb{R}^3)$, we have

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$$\| |x|^{-1} v \|_{\mathbb{R}^3}^2 = \int_0^{+\infty} |v|^2 dr = - \int_0^{+\infty} \partial_r (|v|^2) r dr = -2 \int_0^{+\infty} v \partial_r v r dr.$$

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Then, using Cauchy–Schwarz inequality, we have

$$\int_0^{+\infty} v (\partial_r v r) dr \leq \| |x|^{-1} v \|_{\mathbb{R}^3} \| \partial_r v \|_{\mathbb{R}^3}.$$

Variational formulation

Assume that $|x|f \in L^2(\Omega)$. Then, there exists a unique $u \in \widehat{H}_0^1(\Omega)$ such that

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The proof follows from Lax–Milgram lemma.

Coercivity and continuity of the LHS are obvious by definition.

Continuity of the RHS follows from Hardy's inequality.

Precise setting

Infinite meshes

Let \mathcal{T}_ℓ be a collection of simplices and $\mathcal{T}_\ell^\dagger \subset \mathcal{T}_\ell$.

Shape regular mesh

We say that $\mathcal{T}_\ell \setminus \mathcal{T}_\ell^\dagger$ is a shape regular mesh with constant $\kappa \geq 1$ if the following is satisfied:

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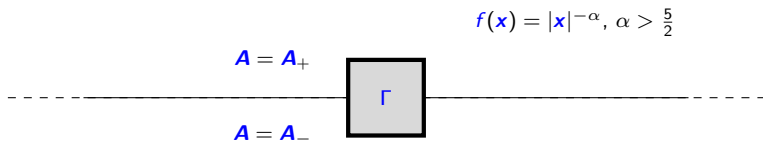
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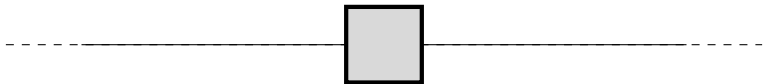
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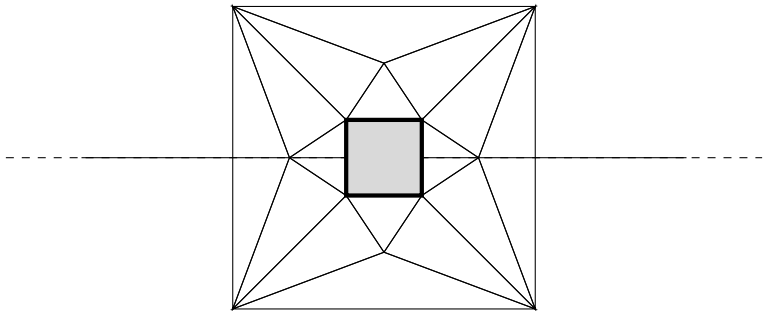
Throughout, we demand that f and \mathbf{A} are constant on each $K \in \mathcal{T}_\ell$.

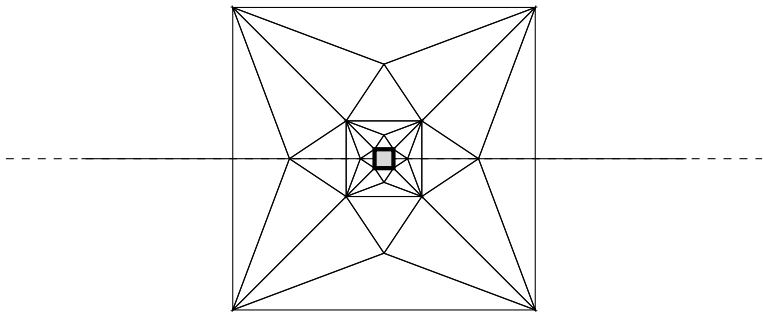
Precise setting

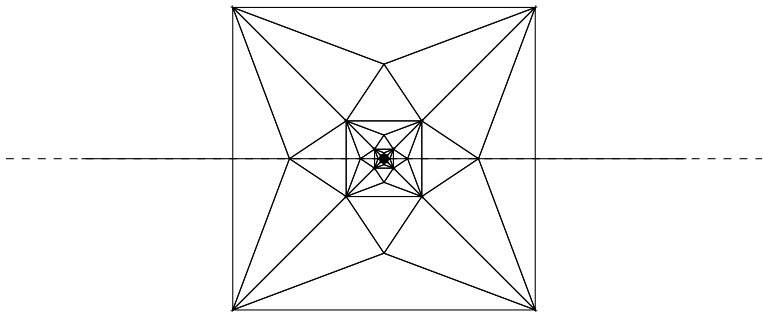
Example





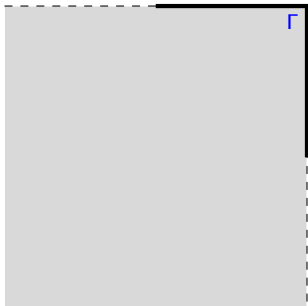


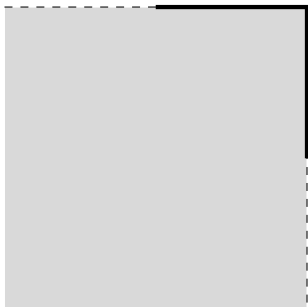


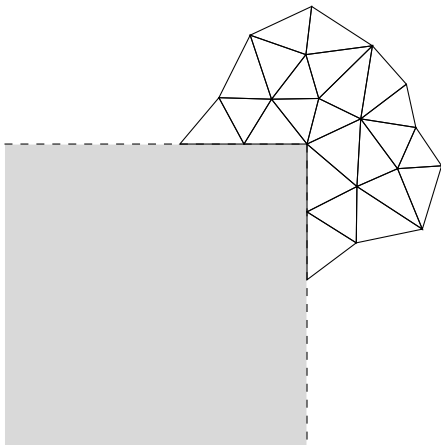


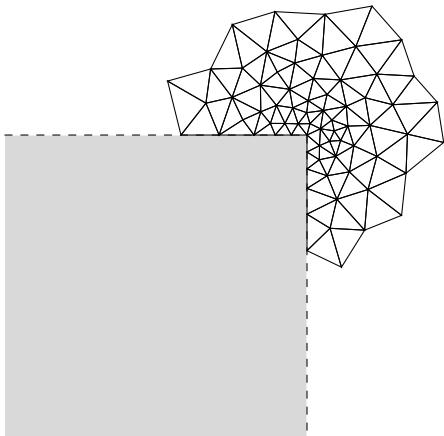
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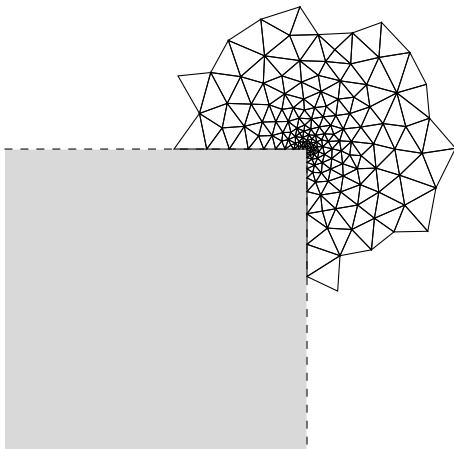
Another example

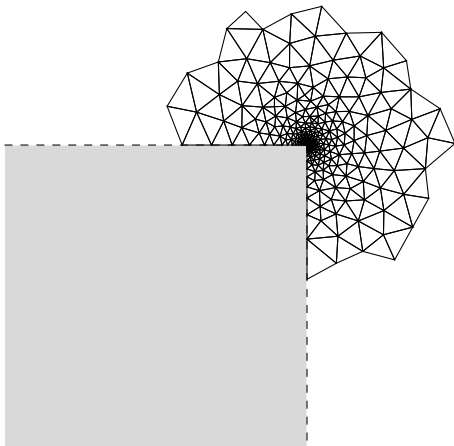












Residual based a posteriori estimator

Residual based a posteriori estimator
A posteriori error estimator

A posteriori estimator

For all $K \in \mathcal{T}_\ell$ and $v_\ell \in \mathcal{P}_p(\mathcal{T}_\ell \setminus \mathcal{T}_\ell^\dagger) \cap H_0^1(\Omega)$, we introduce

$$\eta_\ell^2(v_\ell, K) := h_K^2 \|f + \nabla \cdot (\mathbf{A} \nabla v_\ell)\|_K^2 + h_K \| \llbracket \mathbf{A} \nabla v_\ell \rrbracket \cdot \mathbf{n} \|_{\partial K \setminus \Gamma}^2.$$

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Here, our goal is to show the following properties.

Reliability and efficiency

$$\|\nabla(\mathbf{u} - \mathbf{u}_\ell)\|_\Omega \lesssim \eta_\ell(\mathbf{u}_\ell, \mathcal{T}_\ell) \quad \eta_\ell(\mathbf{u}_\ell, K) \lesssim_p \|\nabla(\mathbf{u} - \mathbf{u}_\ell)\|_K$$

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In fact, one can copy-paste the standard proof for the efficiency.
The interesting (and non-trivial) property is the reliability.

Residual based a posteriori estimator
Reliability

Why is it complicated?

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for a quasi-interpolation operator $J_\ell : \widehat{H}_0^1(\Omega) \rightarrow \mathcal{P}_p(\mathcal{T}_\ell \cap \mathcal{T}_\ell^\dagger) \cap H_0^1(\Omega)$.

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The solution hinge on

- (a) a careful use of a Scott-Zhang type J_ℓ ,
- (b) the specific shape-regularity assumptions we made,
- (c) Hardy's inequality.

For $v \in \widehat{H}_0^1(\Omega)$, using element-wise IBPs, we prove the following:

$$(\mathbf{A}\nabla(u - u_\ell), \nabla v)_\Omega = (f, v)_\Omega - (\mathbf{A}\nabla u_\ell, \nabla v)_\Omega$$

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By Galerkin orthogonality, we can replace v by $v - J_\ell v$, so that

$$(\mathbf{A}\nabla(u - u_\ell), \nabla v)_\Omega \leq \eta_\ell(u_\ell, \mathcal{T}_\ell) \left\{ \sum_{K \in \mathcal{T}_\ell} \left(h_K^{-2} \|v - J_\ell v\|_K^2 + h_K^{-1} \|v - J_\ell v\|_{\partial K}^2 \right) \right\}^{1/2}.$$

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 &= (f - \nabla_\ell \cdot (\mathbf{A}\nabla u_\ell), v)_{\mathcal{T}_\ell} - \langle \mathbf{A}\nabla u_\ell \cdot \mathbf{n}, v \rangle_{\partial\mathcal{T}_\ell} \\
 &= (f - \nabla_\ell \cdot (\mathbf{A}\nabla u_\ell), v)_{\mathcal{T}_\ell} - \frac{1}{2} \langle [\mathbf{A}\nabla u_\ell] \cdot \mathbf{n}, v \rangle_{\partial\mathcal{T}_\ell} \\
 &\leq \eta_\ell(u_\ell, \mathcal{T}_\ell) \left\{ \sum_{K \in \mathcal{T}_\ell} \left(h_K^{-2} \|v\|_K^2 + h_K^{-1} \|v\|_{\partial K}^2 \right) \right\}^{1/2}.
 \end{aligned}$$

By Galerkin orthogonality, we can replace v by $v - J_\ell v$, so that

$$(\mathbf{A}\nabla(u - u_\ell), \nabla v)_\Omega \leq \eta_\ell(u_\ell, \mathcal{T}_\ell) \left\{ \sum_{K \in \mathcal{T}_\ell} \left(h_K^{-2} \|v - J_\ell v\|_K^2 + h_K^{-1} \|v - J_\ell v\|_{\partial K}^2 \right) \right\}^{1/2}.$$

It remains to show that

$$\|\mathbf{h}^{-1}(v - J_\ell v)\|_\Omega + \|\nabla(v - J_\ell v)\|_\Omega \lesssim \|\nabla v\|_\Omega.$$

Quasi-interpolation estimate

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(c) For the active elements touching Γ_ℓ , we combine the two above arguments.

Adaptive algorithm

Adaptive algorithm

The algorithm

I now want to investigate the following adaptive algorithm.

Adaptive algorithm

Given an initial mesh \mathcal{T}_0 , for $\ell = 0, 1, \dots$, do

1. Compute the discrete solution u_ℓ associated with the mesh $\mathcal{T}_\ell \setminus \mathcal{T}_0$.
2. Compute the estimator associated estimators $\eta_\ell(u_\ell, K)$.
3. Find the elements $K \in \mathcal{M}_\ell \subset \mathcal{T}_\ell$ where $\eta_\ell(u_\ell, K)$ is large with Dörfler's marking.
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There are also concerns about whether NVB terminates for infinite meshes in Step 4.

Let us first investigate Steps 2 and 3.

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Otherwise, we need the following (weak) extra assumption.

Extra load term decay

We assume that there exists $\varepsilon > 0$ such that

$$|\mathbf{x}|^{1+\varepsilon} f \in L^2(\Omega),$$

and that (an upper bound of) $\| |\mathbf{x}|^{1+\varepsilon} f \|_\Omega$ is known.

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To do so, we observe that if $\mathcal{T}_\ell \setminus \mathcal{T}_0 \subset B_L$ for some $L \geq 0$, then

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As a result, we can find L large enough to exclude only negligible elements far away. That way, we can find a minimal set \mathcal{M}_ℓ , which is sufficient for optimal convergence.

Our algorithm will always mark a finite set of elements \mathcal{M}_ℓ .
However, these may lead to infinitely many refinements to enforce conformity.
In fact, we can build such counter examples, but there are “pathological”.

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Indeed, they often assume a finite and quasi-uniform mesh.

All in all, Step 4 terminates with finitely many operations.

Adaptive algorithm
Optimal convergence

We now wish to establish the following result.

Optimal convergence

If there exists a sequence of meshes $\widehat{\mathcal{T}}_\ell$ reachable from \mathcal{T}_0 by finite NVBs, s.t.

$$\|\nabla(\mathbf{u} - \widehat{\mathbf{u}}_\ell)\|_\Omega \leq \widehat{C} |\widehat{\mathcal{T}}_\ell \setminus \mathcal{T}_0|^{-s}$$

with $s \geq 0$, then the sequence produced by the adaptive algorithm satisfies

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As it turns out, they “just work” for infinite meshes.

We need to check the following “axioms”.

Axioms of adaptivity

A1 Stability on non-refined elements:

$$|\eta_e(\mathbf{v}_{e'}, \mathcal{T}_e \cap \mathcal{T}_{e'}) - \eta_e(\mathbf{v}_e, \mathcal{T}_e \cap \mathcal{T}_{e'})| \lesssim \|\nabla(\mathbf{v}_{e'} - \mathbf{v}_e)\|_{\Omega}.$$

A2 Reduction on refined elements: For some $q < 1$, we have

$$\eta_e(\mathbf{v}_e, \mathcal{T}_{e+1} \setminus \mathcal{T}_e) \leq q \eta_e(\mathbf{v}_e, \mathcal{T}_e \setminus \mathcal{T}_{e+1}).$$

A3 Discrete reliability:

This is a stronger localized version of reliability.

A4 Quasi-orthogonality:

This is always satisfied for symmetric Lax–Milgram problems with nested spaces.

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If $|\mathbf{x}|^{1+\varepsilon} f \in L^2(\mathbb{R}^3)$ and $|\mathbf{x}|^{|\alpha|+1} \partial^\alpha f \in L^2(\mathbb{R}^3)$ for $|\alpha| \leq p - 1$,

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$$\|\nabla(u - \hat{u}_\ell)\|_\Omega \leq C |\hat{\mathcal{T}}_\ell \setminus \mathcal{T}_0|^{-p/3+\nu},$$

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If Γ is bounded and $\mathbf{A} = \mathbf{I}$ outside a compact set, we can combine this result with a localization argument.

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for a well-chosen sequence of meshes $\widehat{\mathcal{T}}_\ell$.

This is satisfied, e.g., by $f = (1 + |\mathbf{x}|)^{-\mu}$ with $\mu > 5/2$.

If Γ is bounded and $\mathbf{A} = \mathbf{I}$ outside a compact set, we can combine this result with a localization argument.

We then obtain optimal rates under standard assumptions on the edges in Γ and \mathbf{A} . These correspond to the absence of necessary anisotropic refinements.

Fix $\nu > 0$. For an arbitrary $\tau > 0$, we are going to construct $\widehat{\mathcal{T}}_\ell$ such that

$$\|\nabla(u - \widehat{u}_\ell)\|_{\mathbb{R}^3} \leq \|\nabla(u - \widehat{\mathcal{T}}_\ell u)\|_{\mathbb{R}^3} \lesssim \tau$$

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Step 2: We perform a truncation. Due to Step 1, we have

$$\|\nabla u\|_{B_L^c} \leq L^{-\varepsilon} \| |\mathbf{x}|^\varepsilon \nabla u \|_\Omega \lesssim L^{-\varepsilon}.$$

We pick $L = \tau^{-1/\varepsilon}$, so that

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We subdivide N times the elements of $\mathcal{T}_0 \cap B_L$. This is possible by NVB with $N = 2^n$.

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With $N = \tau^{-1/p}$, this solves the problem, since we then have

$$\|\nabla(u - \widehat{I}_\ell u)\|_{\mathbb{R}^3} \lesssim \tau$$

and $|\widehat{\mathcal{T}}_\ell \setminus \mathcal{T}_0| \lesssim |\log \tau| \tau^{-3/p} \lesssim \tau^{-3/p-\nu}$ for any $\nu > 0$.

Numerical examples

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A quick discussion about 2D

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I will show 2D numerics though. I will only consider RHS with angular modes. For instance, for the first angular mode, the “fundamental” solution is

$$u(\mathbf{x}) = \frac{\sin(\theta)}{r},$$

which has similar decay properties that the 3D fundamental solution.

Numerical examples

Smooth solutions on structured meshes

We first consider $\Omega = \mathbb{R}^2 \setminus [-1, 1]^2$, $\mathbf{A} = \mathbf{I}$, and

$$u(\mathbf{x}) = \chi(r) \frac{\sin(m\theta)}{r^m}$$

for $m = 1$ and 2 , where χ is a smooth cutoff such that $u|_{\Gamma} = 0$.

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We employ the Dörfler parameter 0.3 , meaning that we mark the element so that

$$\sum_{K \in \mathcal{M}_\ell} \eta_\ell^2(u_\ell, K) \geq 0.3^2 \sum_{K \in \mathcal{T}_\ell} \eta_\ell^2(u_\ell, K).$$

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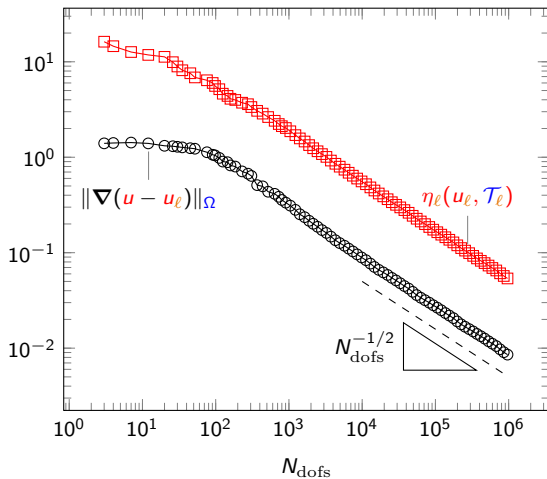
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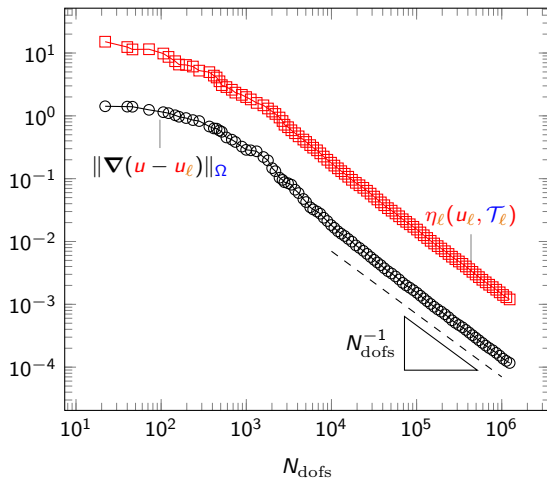
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We consider $p = 1, 2$ and 3 .

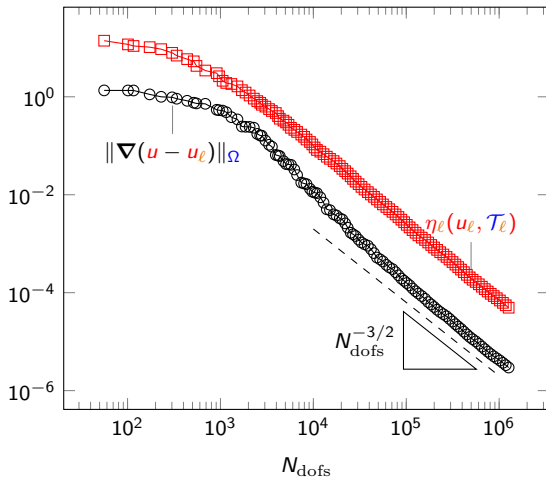
$m = 1$ and $p = 1$



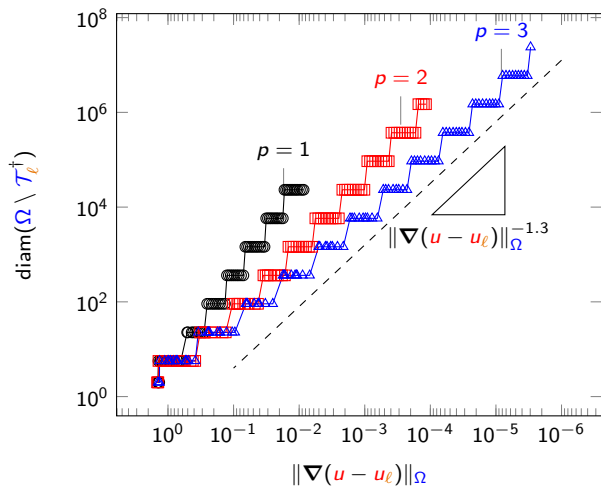
$m = 1$ and $p = 2$



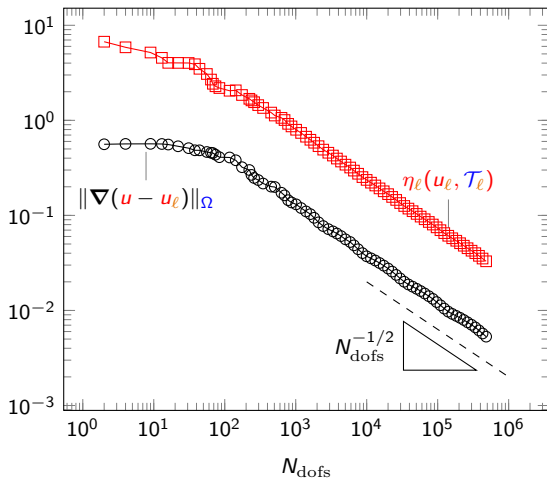
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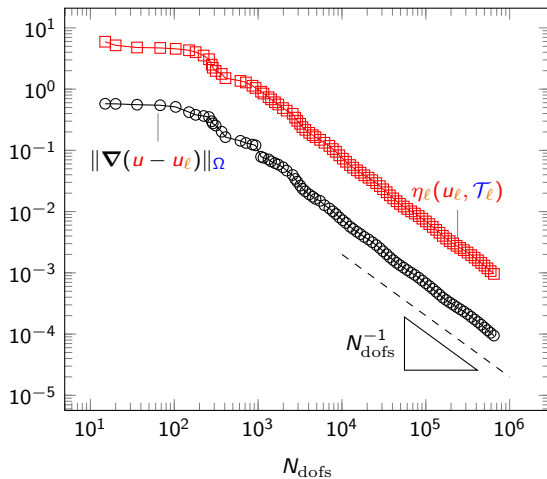
Increase in diameter for $m = 1$



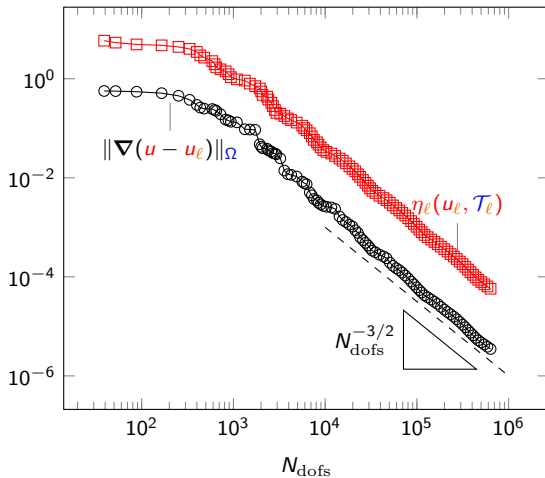
$m = 2$ and $p = 1$



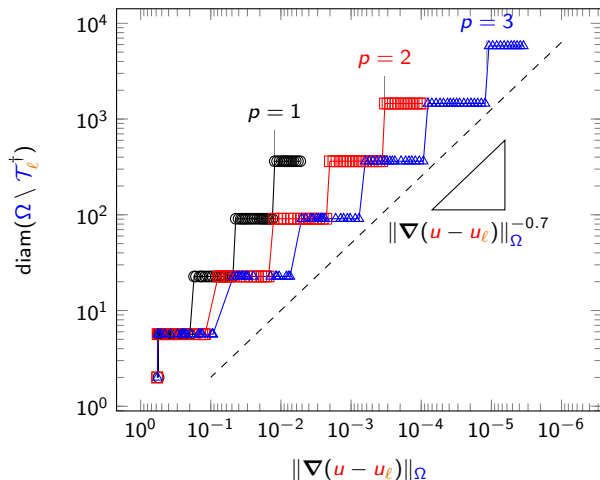
$m = 2$ and $p = 2$



$m = 2$ and $p = 3$



Increase in diameter for $m = 2$



Numerical examples

Corner singularities on structured meshes

We keep the setting with $\Omega = \mathbb{R}^2 \setminus [-1, 1]^2$ and $\mathbf{A} = \mathbf{I}$, but we let

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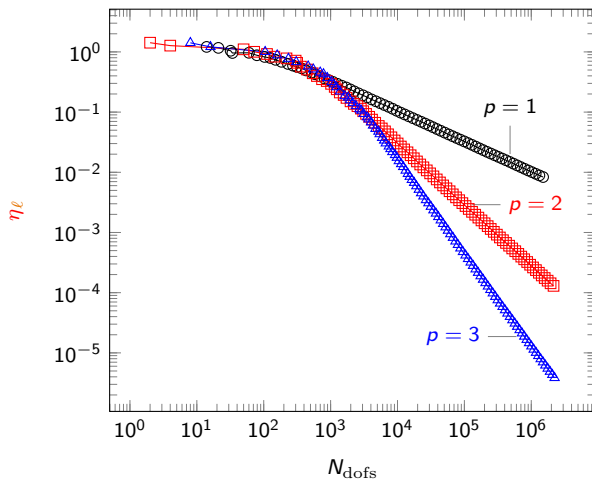
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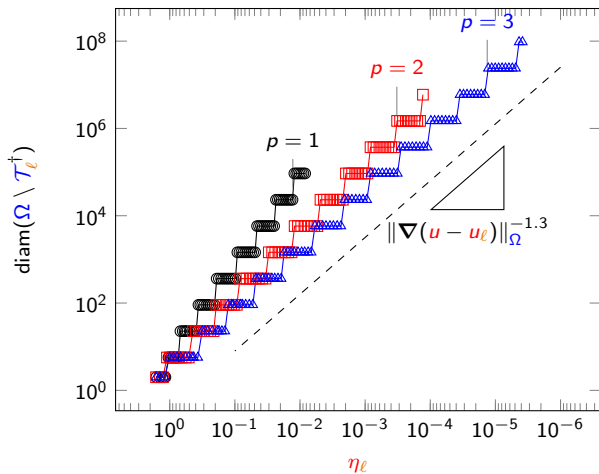
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We consider the Dörfler parameter 0.3, and $p = 1, 2$ and 3.

Convergence of the estimator





Numerical examples

Infinite L-shape with unstructured meshes

We consider $\Omega = \mathbb{R}^2 \setminus [-\infty, 0]^2$, $\mathbf{A} = \mathbf{I}$, and

$$f = \frac{\sin(\theta/3)}{(1+r)^5}.$$

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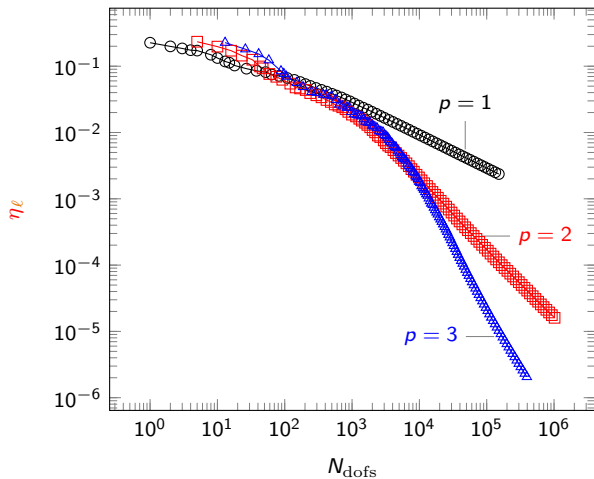
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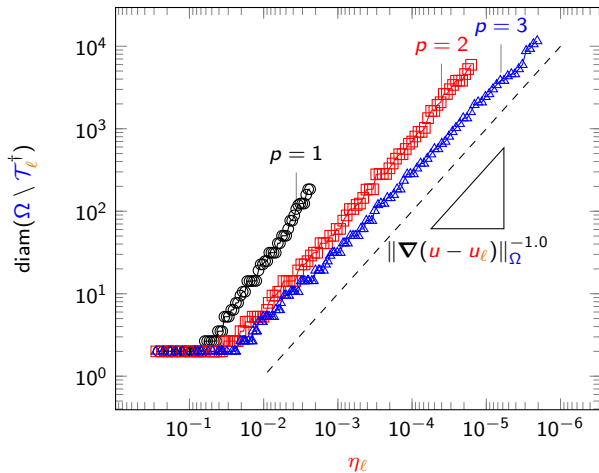
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Concluding remarks

The theory and implementation for AFEM naturally extend to unbounded domains.

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Related results in the simpler setting of reaction–diffusion equations are available:



T. Chaumont-Frelet, *arXiv*, 2024.



T. Chaumont-Frelet and G. Gantner, *arXiv*, 2025.