

Numerical methods, a priori and a posteriori error estimates, and *hp* finite elements

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Inria Paris & Ecole des Ponts

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Inria



Outline

- 1 Numerical approximations of PDEs
- 2 A priori and a posteriori error analysis
- 3 A posteriori error estimates
- 4 Mesh adaptivity
- 5 hp finite elements: a priori error estimates
- 6 hp finite elements: mesh & polynomial degree adaptivity

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Numerical approximations of PDEs

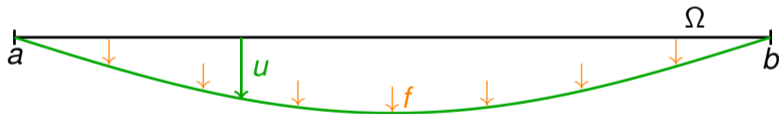
Numerical methods

- mathematically-based algorithms evaluated by **computers**
- deliver **approximate solutions**
- conception: more effort \Rightarrow closer to the unknown solution
- example: elastic string

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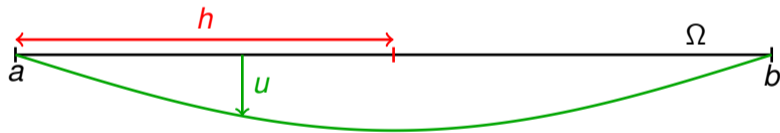


Numerical approximation u_h and its convergence to u

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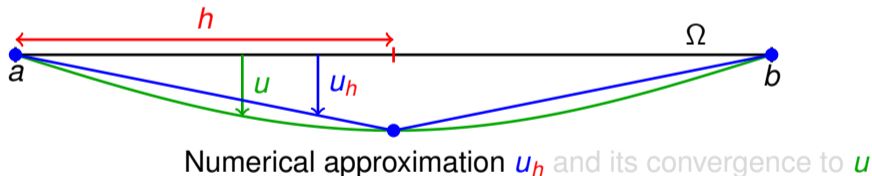


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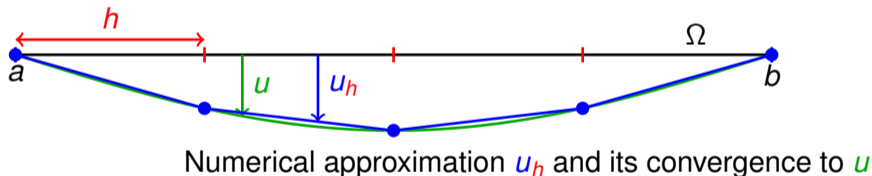
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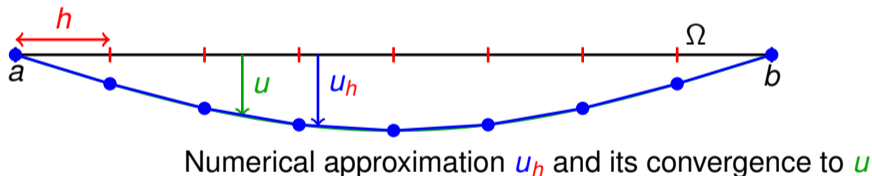
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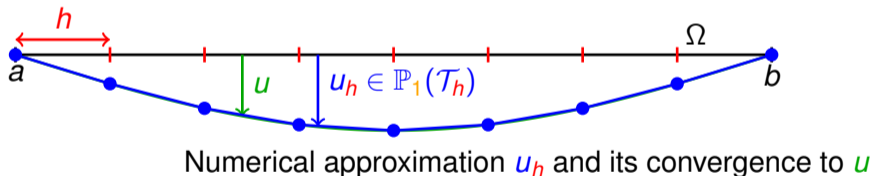
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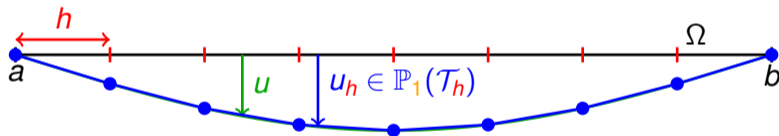
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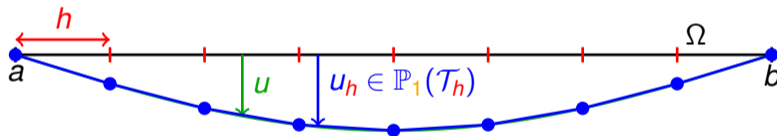
Error

$$\|\nabla(u - u_h)\| = \left\{ \int_a^b |(u - u_h)'|^2 \right\}^{\frac{1}{2}}$$

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Numerical approximation u_h and its convergence to u

Error

$$\|\nabla(u - u_h)\| = \left\{ \int_a^b |(u - u_h)'|^2 \right\}^{\frac{1}{2}}$$

Polynomial degree p

$$u_h \in \mathbb{P}_p(\mathcal{T}_h)$$

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Crucial questions

- 1 Does the method **converge**?
 $\|\nabla(u - u_h)\| \rightarrow 0$? For $h \searrow 0$? For $p \nearrow \infty$?
- 2 At which **speed**? $\|\nabla(u - u_h)\| \leq ?$
- 3 Is the **analysis optimal**? Is **uniform refinement** $h \searrow 0$ or $p \nearrow \infty$ **optimal**?

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A priori

error estimates

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A priori & a posteriori error estimates

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Answers

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- 2 $C(u, p)h^p$ in **h -analysis**, $C(u, p)\left(\frac{h}{p}\right)^p$ in **hp -analysis**.

A priori & a posteriori error estimates

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Answers

- 1 **A posteriori** error **estimates. Justification** of the **result**.
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- 3 **Adaptivity**, focusing, h & p refined **non uniformly**.

CDG Terminal 2E collapse in 2004 (opened in 2003)



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Case Studies in Engineering Failure Analysis 3 (2015) 88–95



Contents lists available at ScienceDirect

Case Studies in Engineering Failure Analysis

journal homepage: www.elsevier.com/locate/caeafa



Reliability study and simulation of the progressive collapse of
Roissy Charles de Gaulle Airport



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^aÉcole Supérieure d'Ingénieurs de Bryonville (ESIB), Université Saint-Joseph, CSF Mar Roubaix, PO Box 11-514, Blvd El-Sabb Belour 11072050, Lebanon

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probably **numerical simulations done with insufficient precision**,
I believe **without error certification** by a posteriori error estimates

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A posteriori error estimates: **certify** the error

Poisson equation

$$\begin{aligned} -\Delta u &= f & \text{in } \Omega, \\ u &= 0 & \text{on } \partial\Omega \end{aligned}$$

Guaranteed error upper bound (reliability)

$$\underbrace{\|\nabla(u - u_h)\|}_{\text{unknown error}} \leq \underbrace{\eta(u_h)}_{\text{computable estimator}}$$

Global error lower bound (global efficiency; mathematical equivalence of the error and estimator)

$$\eta(u_h) \leq C \|\nabla(u - u_h)\|$$

Local error lower bound (local efficiency; if the estimator predicts error in an element K , then it is in K and around)

$$\eta_K(u_h) \leq C \|\nabla(u - u_h)\|_K$$

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A posteriori error estimates: reconstructions

Theorem (Error characterization)

Let $u \in H_0^1(\Omega)$ be the weak solution and let $u_h \in H^1(\mathcal{T}_h)$ be arbitrary. Then

$$\|\nabla(u - u_h)\|^2 = \min_{\substack{\mathbf{v} \in \mathbf{H}(\text{div}, \Omega) \\ \nabla \cdot \mathbf{v} = f}} \|\nabla u_h + \mathbf{v}\|^2 + \min_{v \in H_0^1(\Omega)} \|\nabla(u_h - v)\|^2.$$

Comments

- It is now enough to choose suitable $\sigma_h \in \mathbf{H}(\text{div}, \Omega)$ and $s_h \in H_0^1(\Omega)$.
- A simple choice for nonconforming finite elements given in the lecture notes.

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How large is the overall error? (model pb, known smooth solution)

h	p	$\eta(u_h)$	rel. error estimate	$\frac{\eta(u_h)}{\ \nabla u_h\ }$	$\ \nabla(u - u_h)\ $	rel. error	$\frac{\ \nabla(u - u_h)\ }{\ \nabla u_h\ }$	$\frac{\eta(u_h)}{\ \nabla(u - u_h)\ }$
h_0	1	1.25	28%		1.07	24%		1.17
$\approx h_0/2$	2	0.67	18%		0.57	15%		1.17
$\approx h_0/4$	3	0.30	12%		0.26	10%		1.17
$\approx h_0/8$	4	0.15	8%		0.13	7%		1.17
$\approx h_0/2$	2	0.67	18%		0.57	15%		1.17
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$\approx h_0/8$	4	0.15	8%		0.13	7%		1.17

A. Ern, M. Vohralik, SIAM Journal on Numerical Analysis (2015)
 V. Dougalj, A. Ern, M. Vohralik, SIAM Journal on Scientific Computing (2018)

How large is the overall error? (model pb, known smooth solution)

h	p	$\eta(u_h)$	rel. error estimate $\frac{\eta(u_h)}{\ \nabla u_h\ }$	$\ \nabla(u - u_h)\ $	rel. error $\frac{\ \nabla(u - u_h)\ }{\ \nabla u_h\ }$	$\rho_{hp} = \frac{\eta(u_h)}{\ \nabla(u - u_h)\ }$
h_0	1	1.25	28%	1.07	24%	1.17
$\approx h_0/2$		6.07×10^{-1}	28%	5.55×10^{-1}	24%	1.17
$\approx h_0/4$		3.10×10^{-1}	28%	2.83×10^{-1}	24%	1.17
$\approx h_0/8$		1.45×10^{-1}	28%	1.33×10^{-1}	24%	1.17
$\approx h_0/2$	2	4.23×10^{-1}	28%	3.87×10^{-1}	24%	1.17
$\approx h_0/4$	3	2.52×10^{-1}	28%	2.30×10^{-1}	24%	1.17
$\approx h_0/8$	4	2.50×10^{-1}	28%	2.28×10^{-1}	24%	1.17

A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)
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$\approx h_0/8$		1.45×10^{-1}	3.3%		1.30×10^{-1}	3.2%		
$\approx h_0/2$	2	4.23×10^{-1}	$9.2 \times 10^{-2}\%$		3.92×10^{-1}	$8.8 \times 10^{-2}\%$		
$\approx h_0/4$	3	2.52×10^{-1}	$5.9 \times 10^{-2}\%$		2.30×10^{-1}	$5.6 \times 10^{-2}\%$		
$\approx h_0/8$	4	2.50×10^{-1}	$5.9 \times 10^{-2}\%$		2.30×10^{-1}	$5.6 \times 10^{-2}\%$		

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$\approx h_0/8$	4	2.50×10^{-1}	5.9×10^{-1} %		2.58×10^{-1}		

A. Ern, M. WOHRE, SIAM Journal on Numerical Analysis (2015)

V. DOLBE, A. Ern, M. WOHRE, SIAM Journal on Scientific Computing (2018)

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$\approx h_0/2$		6.07×10^{-1}	14%		5.56×10^{-1}	13%		1.09
$\approx h_0/4$		3.10×10^{-1}	7.0%		2.92×10^{-1}	6.6%		1.05
$\approx h_0/8$		1.45×10^{-1}	3.3%		1.39×10^{-1}	3.1%		1.02
$\approx h_0/2$	2	4.23×10^{-1}	$9.2 \times 10^{-2}\%$		4.07×10^{-1}	$9.2 \times 10^{-2}\%$		1.01
$\approx h_0/4$	3	2.52×10^{-1}	$5.9 \times 10^{-2}\%$		2.60×10^{-1}	$5.9 \times 10^{-2}\%$		1.01
$\approx h_0/8$	4	2.50×10^{-1}	$5.8 \times 10^{-2}\%$		2.58×10^{-1}	$5.8 \times 10^{-2}\%$		1.01

A. Ern, M. WOHRE, SIAM Journal on Numerical Analysis (2015)

V. DOLBE, A. Ern, M. WOHRE, SIAM Journal on Scientific Computing (2018)

How large is the overall error? (model pb, known smooth solution)

h	p	$\eta(u_h)$	rel. error estimate	$\frac{\eta(u_h)}{\ \nabla u_h\ }$	$\ \nabla(u - u_h)\ $	rel. error $\frac{\ \nabla(u - u_h)\ }{\ \nabla u_h\ }$	$f^{eff} = \frac{\eta(u_h)}{\ \nabla(u - u_h)\ }$
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$\approx h_0/4$		3.10×10^{-1}	7.0%		2.92×10^{-1}	6.6%	1.06
$\approx h_0/8$		1.45×10^{-1}	3.3%		1.39×10^{-1}	3.1%	1.04
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A. Ern, M. WOHRAK, SIAM Journal on Numerical Analysis (2015)

V. DOLBEK, A. ERN, M. WOHRAK, SIAM Journal on Scientific Computing (2016)

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A. Ern, M. Wheeler, SIAM Journal on Numerical Analysis (2015)

Y. Dauge, A. Ern, M. Wheeler, SIAM Journal on Scientific Computing (2016)

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A. Ern, M. Vohralik, SIAM Journal on Numerical Analysis (2015)

V. Dolejší, A. Ern, M. Vohralik, SIAM Journal on Scientific Computing (2016)

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A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

V. Dolejší, A. Ern, M. Vohralík, SIAM Journal on Scientific Computing (2016)

Outline

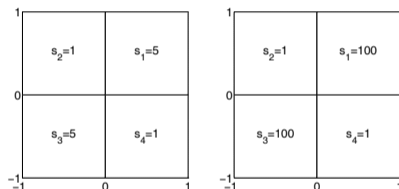
- 1 Numerical approximations of PDEs
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- 3 A posteriori error estimates
- 4 Mesh adaptivity**
- 5 *hp* finite elements: a priori error estimates
- 6 *hp* finite elements: mesh & polynomial degree adaptivity

Problem with singular solution

- consider the pure diffusion equation

$$-\nabla \cdot \mathbf{S} \nabla u = 0 \quad \text{in} \quad \Omega = (-1, 1) \times (-1, 1)$$

- discontinuous and inhomogeneous \mathbf{S} , two cases:

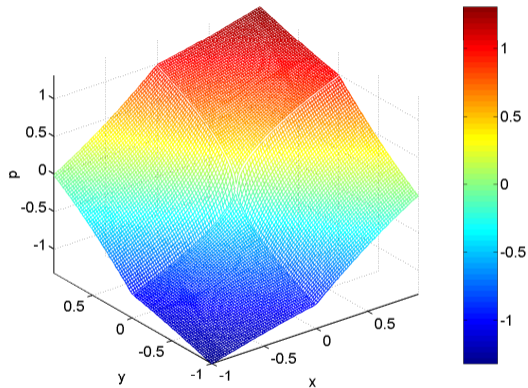


- analytical solution: **singularity** at the origin

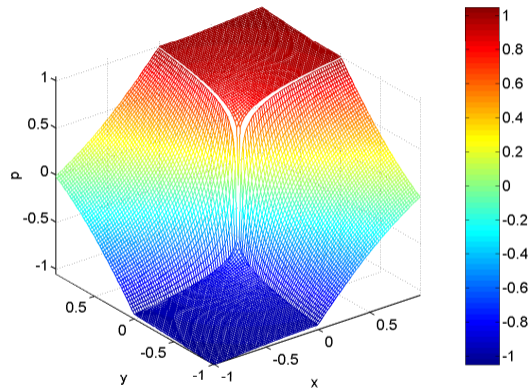
$$u(r, \theta) = r^\alpha (a_j \sin(\alpha\theta) + b_j \cos(\alpha\theta))$$

- (r, θ) polar coordinates in Ω
- a_j, b_j constants depending on Ω_j
- α regularity of the solution, $u \in H^{1+\alpha}(\Omega)$

Analytical solutions

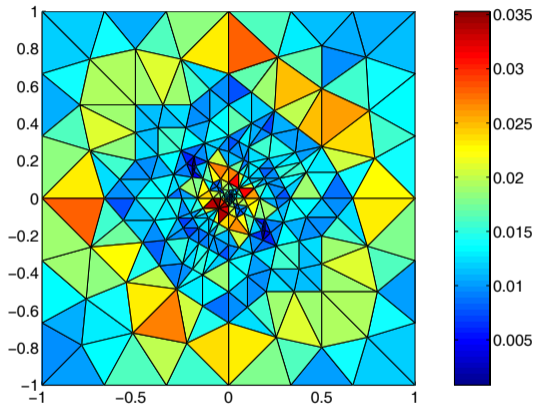


Case 1 ($\alpha \approx 0.54$)

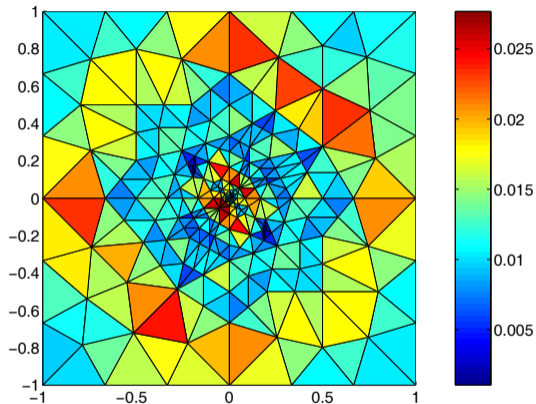


Case 2 ($\alpha \approx 0.13$)

Where is the error localized?



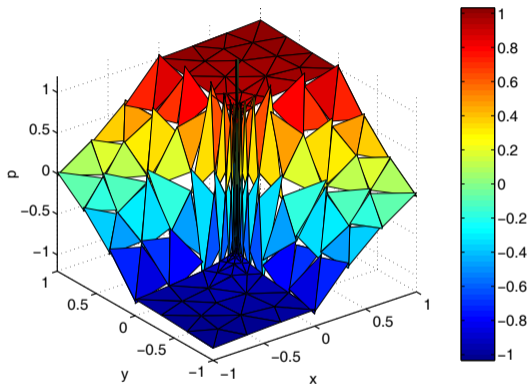
Estimated error distribution $\eta_K(u_h)$



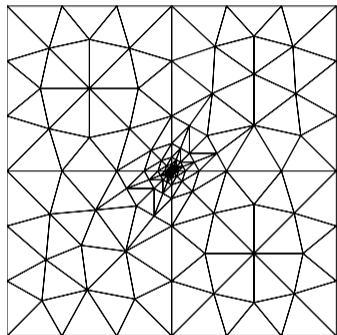
Exact error distribution $\|\nabla(u - u_h)\|_K$

M. Vohralik, SIAM Journal on Numerical Analysis (2007)

Can we adapt the mesh to better approximate the solution?



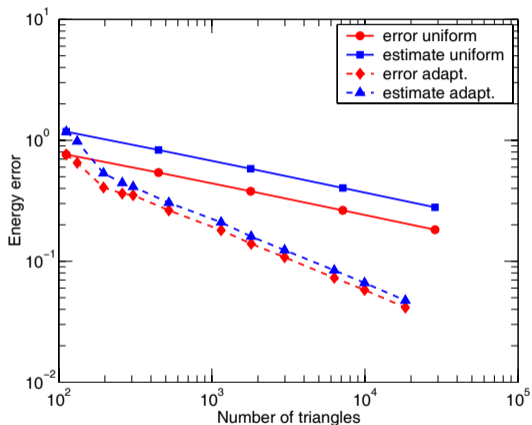
Nonconforming finite elements



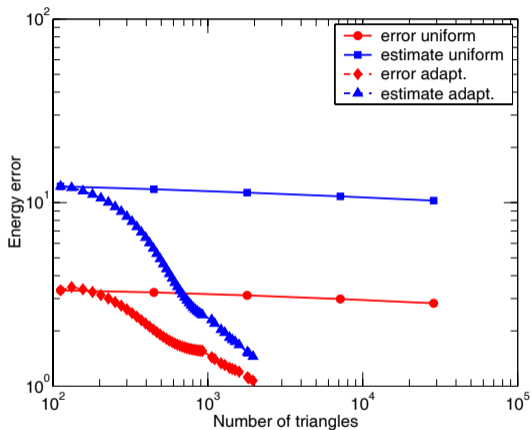
Adaptively refined mesh

M. Vohralik, SIAM Journal on Numerical Analysis (2007)

Does this lead to a better error decrease?



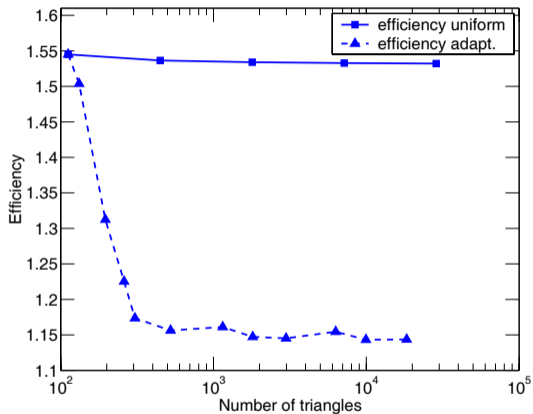
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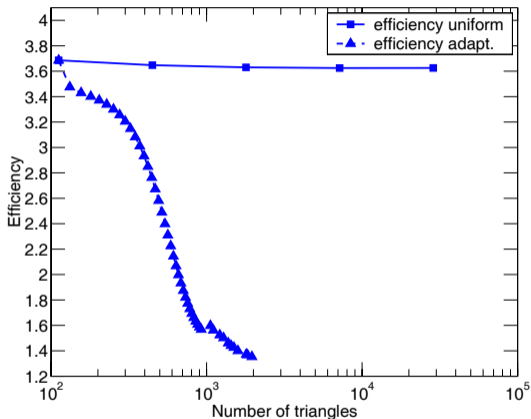
Case 2 ($\alpha \approx 0.13$)

M. Vohralik, SIAM Journal on Numerical Analysis (2007)

Quality of the estimates for a singular solution



Case 1 ($\alpha \approx 0.54$)



Case 2 ($\alpha \approx 0.13$)

M. Vohralik, SIAM Journal on Numerical Analysis (2007)

Adaptive mesh refinement

Mesh adaptivity

- Dörfler marking: subset \mathcal{M}_ℓ containing θ -fraction of the estimates

$$\sum_{K \in \mathcal{M}_\ell} \eta_K(u_\ell)^2 \geq \theta^2 \sum_{K \in \mathcal{T}_\ell} \eta_K(u_\ell)^2$$

- refine the elements in \mathcal{M}_ℓ

Convergence on a sequence of adaptively refined meshes \mathcal{T}_ℓ

- $\|\nabla(u - u_\ell)\| \rightarrow 0$ for $\ell \rightarrow \infty$
- some mesh elements may not be refined at all: $h \not\rightarrow 0$ **uniformly**

Optimal error decay rate wrt degrees of freedom

- $\|\nabla(u - u_\ell)\| \leq C|\text{DoF}_\ell|^{-p/d}$ (replaces h^p)
- same for smooth & singular solutions: ~~higher-order only pays-off for sm. sol.~~
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h vs. hp a priori analysis

Theorem (Deny–Lions/Bramble–Hilbert)

For all $K \in \mathcal{T}_h$ and $v \in H^{p+1}(K)$,

$$\min_{v_h \in \mathcal{P}_p(K)} \|\nabla(v - v_h)\|_K \leq \sqrt{(p+1)!} \left(\frac{h_K}{\pi}\right)^p |v|_{H^{p+1}(K)}.$$

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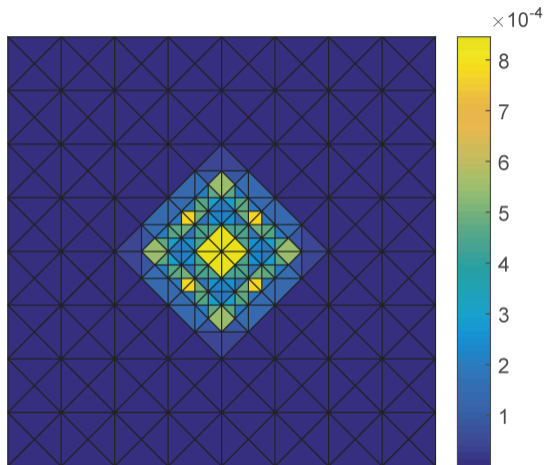
Comments

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- convergence for both $h \searrow 0$ & $p \nearrow \infty$

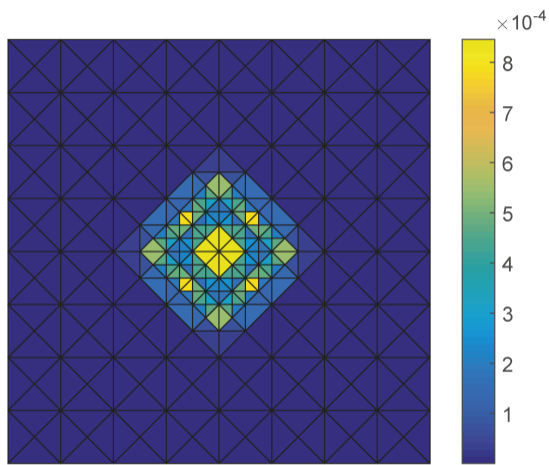
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- 6 hp finite elements: mesh & polynomial degree adaptivity

Where is the error localized?



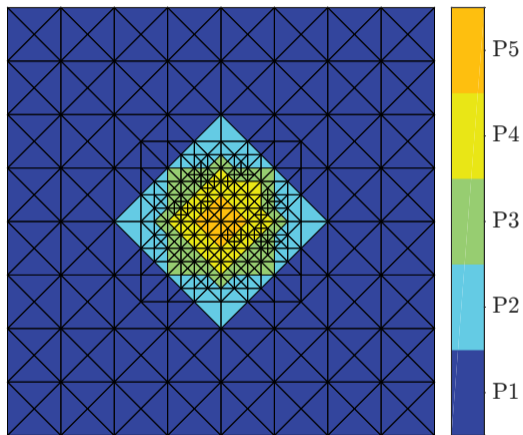
Estimated error distribution $\eta_K(u_h)$



Exact error distribution $\|\nabla(u - u_h)\|_K$

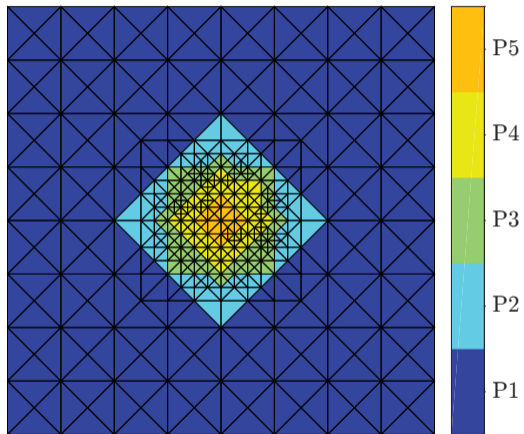
P. Daniel, A. Ern, I. Smears, M. Vohralík, *Computers & Mathematics with Applications* (2018)

Can we decrease the error efficiently? hp adaptivity, (smooth solution)

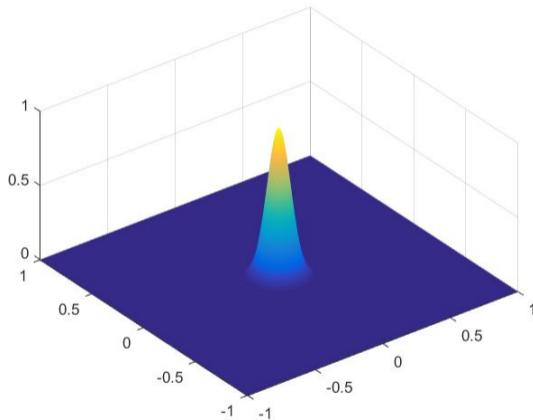


Mesh \mathcal{T}_ℓ and pol. degrees p_K

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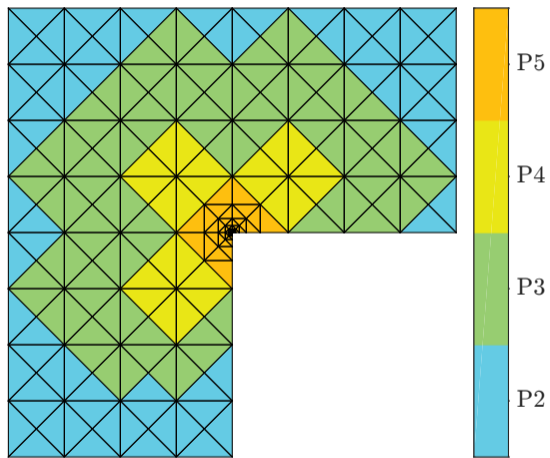
Mesh \mathcal{T}_ℓ and pol. degrees p_K



Exact solution

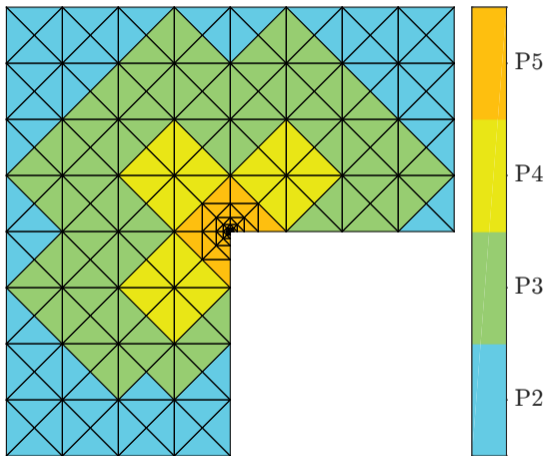
P. Daniel, A. Ern, I. Smears, M. Vohralík, Computers & Mathematics with Applications (2018)

Can we decrease the error efficiently? *hp* adaptivity, (**singular** solution)

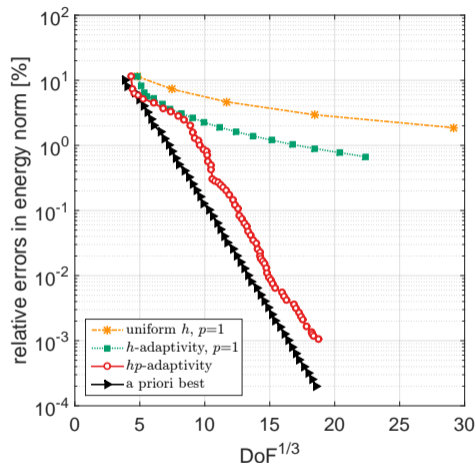


Mesh \mathcal{T}_ℓ and polynomial degrees p_K

Can we decrease the error efficiently? hp adaptivity, (singular solution)



Mesh \mathcal{T}_ℓ and polynomial degrees p_K



Relative error as a function of DoF

P. Daniel, A. Ern, I. Smears, M. Vohralík, Computers & Mathematics with Applications (2018)

Adaptive mesh & polynomial degree refinement

Mesh & polynomial degree adaptivity

- **decision** between h or p refinement needs to be done
- much harder than just h -adaptivity

Convergence on a sequence of adaptively refined hp spaces V_ℓ

- $\|\nabla(u - u_\ell)\| \rightarrow 0$ for $\ell \rightarrow \infty$
- ~~$h \searrow 0$ uniformly, $p \nearrow 0$ uniformly~~

Optimal error decay rate wrt degrees of freedom

- for $d = 2$, hp refinement gives

$$\|\nabla(u - u_\ell)\| \leq C_1 \frac{1}{e^{C_2 \text{DoF}^{1/3}}}$$

- **exponential** convergence **rate**
- for $d = 2$ and $p = 1$ fixed, adaptive mesh h refinement only gives

$$\|\nabla(u - u_\ell)\| \leq C \frac{1}{\text{DoF}^{1/2}}$$

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