

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

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ENSTA, March 8, 2024

*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

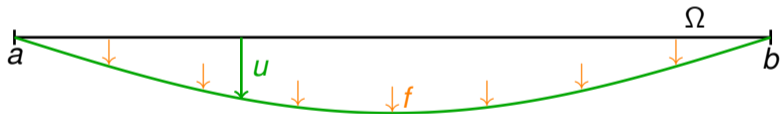
## Numerical methods

- 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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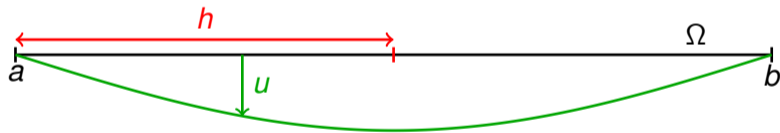


Numerical approximation  $u_h$  and its convergence to  $u$

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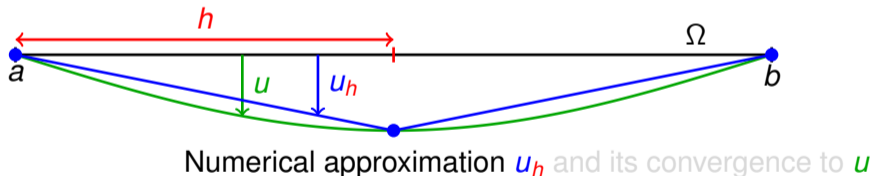


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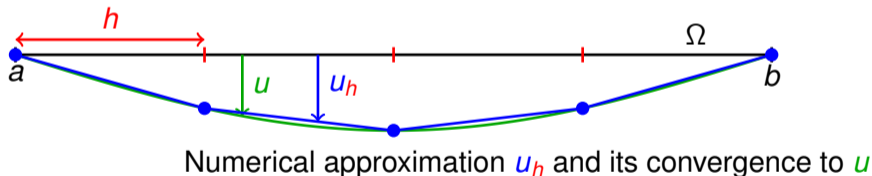
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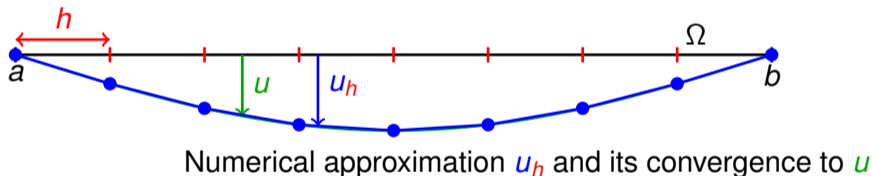




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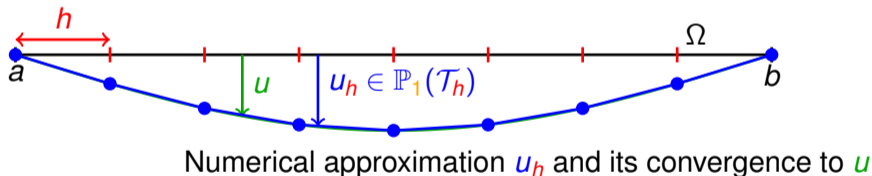
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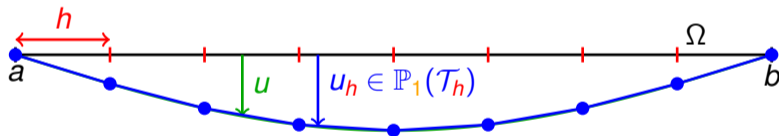
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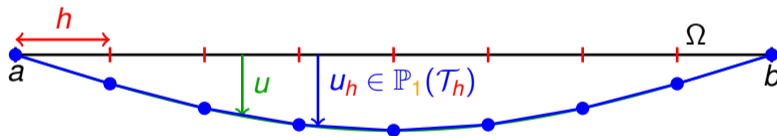
### Error

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

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### 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**?  $\|\nabla(u - u_h)\| \leq ?$
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## Answers

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## Answers

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- 2 Elementwise **estimators**.
- 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



Reliability study and simulation of the progressive collapse of  
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<sup>a</sup>École 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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# 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)

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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

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

## Comments

- It is now enough to choose suitable  $\sigma_h \in \mathbf{H}(\operatorname{div}, \Omega)$  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	14%		0.53	12%		1.17
$\approx h_0/4$	3	0.30	7%		0.26	6%		1.17
$\approx h_0/8$	4	0.15	4%		0.13	3%		1.17
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A. Ern, M. WOHRE, SIAM Journal on Numerical Analysis (2015)  
 V. Doležal, A. Ern, M. WOHRE, SIAM Journal on Scientific Computing (2018)

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$h_0$	1	1.25	28%	1.07	24%	1.17
$\approx h_0/2$	1	$6.07 \times 10^{-1}$	28%	$5.65 \times 10^{-1}$	24%	1.17
$\approx h_0/4$	1	$3.10 \times 10^{-1}$	28%	$2.83 \times 10^{-1}$	24%	1.17
$\approx h_0/8$	1	$1.45 \times 10^{-1}$	28%	$1.35 \times 10^{-1}$	24%	1.17
$\approx h_0/2$	2	$4.23 \times 10^{-1}$	28%	$3.95 \times 10^{-1}$	24%	1.17
$\approx h_0/4$	2	$2.52 \times 10^{-1}$	28%	$2.35 \times 10^{-1}$	24%	1.17
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$\approx h_0/2$	2	$4.23 \times 10^{-1}$	$9.5 \times 10^{-1}$ %		$4.07 \times 10^{-1}$	9.5%	1.17
$\approx h_0/4$	3	$2.52 \times 10^{-1}$	$5.9 \times 10^{-1}$ %		$2.60 \times 10^{-1}$	5.9%	1.17
$\approx h_0/8$	4	$2.50 \times 10^{-1}$	$5.9 \times 10^{-1}$ %		$2.58 \times 10^{-1}$	5.9%	1.17

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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$h$	$p$	$\eta(u_h)$	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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$\approx h_0/8$		$1.45 \times 10^{-1}$	3.3%		$1.39 \times 10^{-1}$	3.1%		1.03
$\approx h_0/2$	2	$4.23 \times 10^{-1}$	$9.2 \times 10^{-2}\%$		$4.07 \times 10^{-1}$	$9.2 \times 10^{-2}\%$		1.02
$\approx h_0/4$	3	$2.52 \times 10^{-1}$	$5.9 \times 10^{-2}\%$		$2.60 \times 10^{-1}$	$5.9 \times 10^{-2}\%$		1.01
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V. Dolejší, A. Ern, M. Vohralik, SIAM Journal on Scientific Computing (2016)



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$\approx h_0/4$	<b>3</b>	$2.62 \times 10^{-4}$	<b><math>5.9 \times 10^{-3}\%</math></b>	$2.60 \times 10^{-4}$	$5.9 \times 10^{-3}\%$	<b>1.01</b>
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V. Dolejší, A. Ern, M. Vohralík, SIAM Journal on Scientific Computing (2016)

# Outline

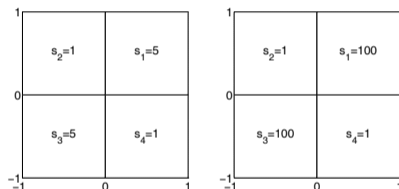
- 1 Numerical approximations of PDEs
- 2 A priori and a posteriori error analysis
- 3 A posteriori error estimates
- 4 Mesh adaptivity**
- 5 *hp* finite elements: a priori error estimates
- 6 *hp* finite elements: mesh & polynomial degree adaptivity

# 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:

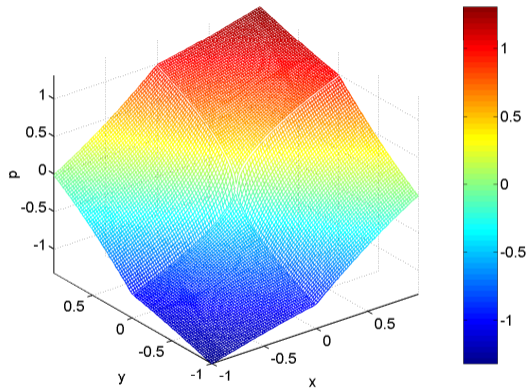


- analytical solution: **singularity** at the origin

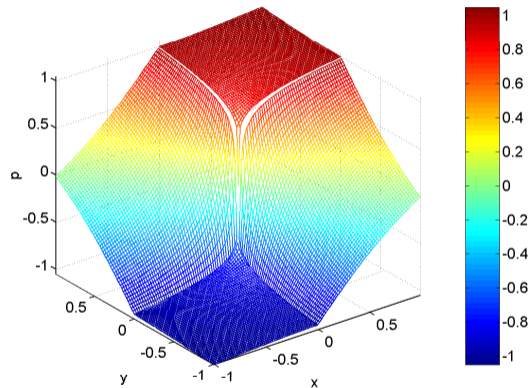
$$u(r, \theta) = r^\alpha (a_j \sin(\alpha\theta) + b_j \cos(\alpha\theta))$$

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

# Analytical solutions

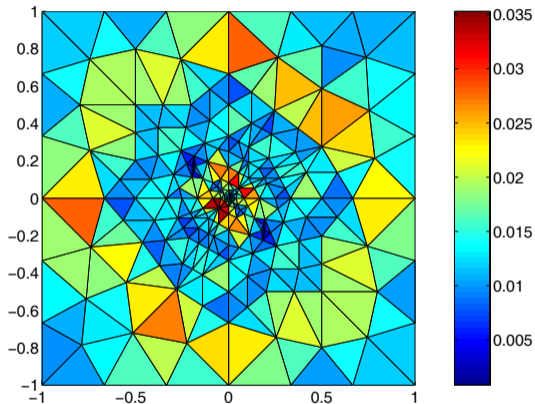


Case 1 ( $\alpha \approx 0.54$ )

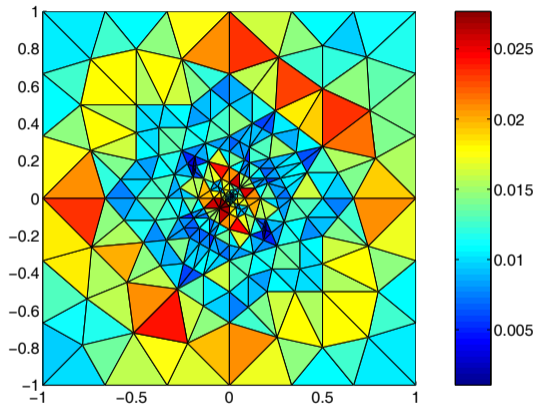


Case 2 ( $\alpha \approx 0.13$ )

# Where is the error localized?



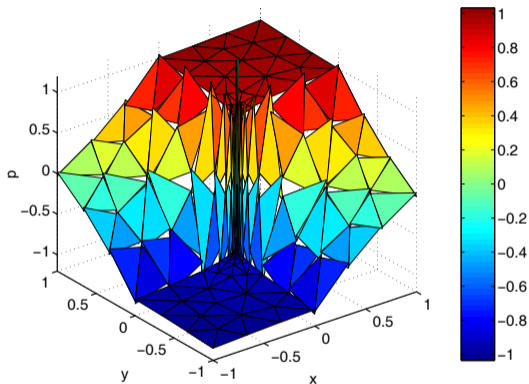
Estimated error distribution  $\eta_K(u_h)$



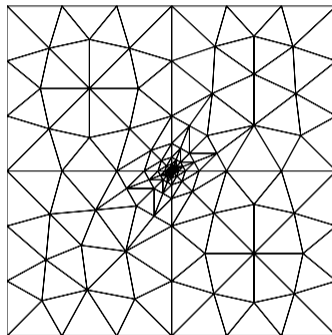
Exact error distribution  $\|\nabla(u - u_h)\|_K$

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

# Can we adapt the mesh to better approximate the solution?



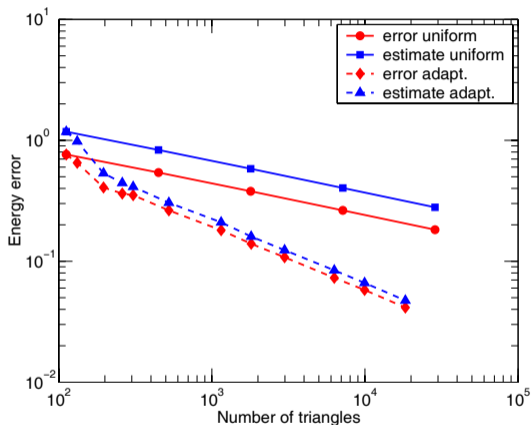
Nonconforming finite elements



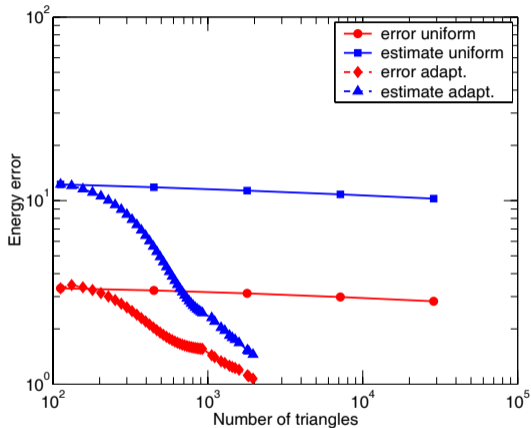
Adaptively refined mesh

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# Does this lead to a better error decrease?



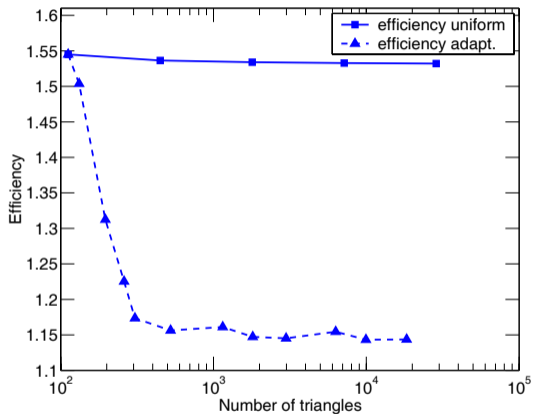
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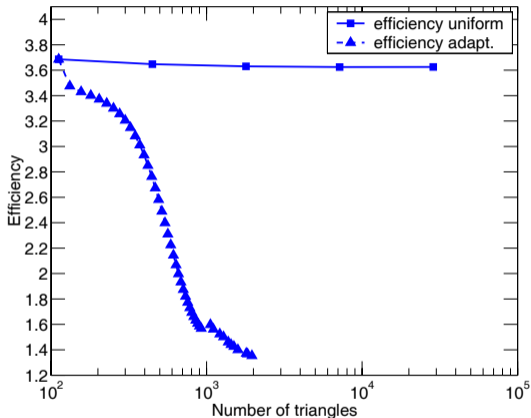
Case 2 ( $\alpha \approx 0.13$ )

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  $\mathcal{M}_\ell$  containing  $\theta$ -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  $\mathcal{M}_\ell$

## Convergence on a sequence of adaptively refined meshes $\mathcal{T}_\ell$

- $\|\nabla(u - u_\ell)\| \rightarrow 0$  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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- 5  $hp$  finite elements: a priori error estimates**
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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

- $C(p)$  depends unfavorably on  $p$
- for fixed  $p$ , convergence as  $h^p$  for  $h \searrow 0$

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### Theorem (Element $hp$ approximation)

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

## Theorem (Element $hp$ approximation)

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

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

## Theorem (A priori rate of convergence)

Let  $u|_K \in H^{p+1}(K)$  for all  $K \in \mathcal{T}_h$ . Then

$$\|\nabla(u - u_h)\| \leq C(p) h^p |u|_{H^{p+1}(\mathcal{T}_h)}.$$

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## Comments

- $C(p)$  depends unfavorably on  $p$
- for fixed  $p$ , convergence as  $h^p$  for  $h \searrow 0$

# $h$ vs. $hp$ a priori analysis

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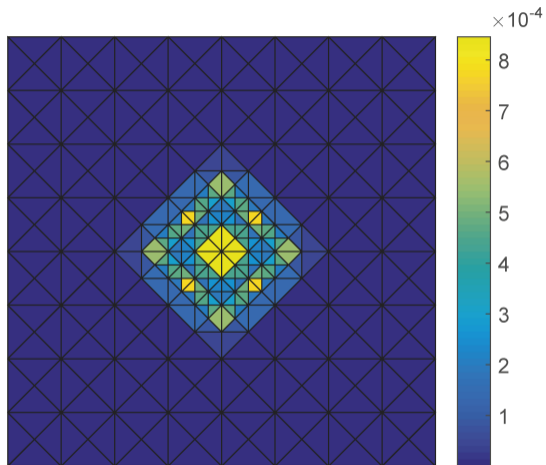
### Comments

- $C$  does not depend on  $p$
- convergence for both  $h \searrow 0$  &  $p \nearrow \infty$

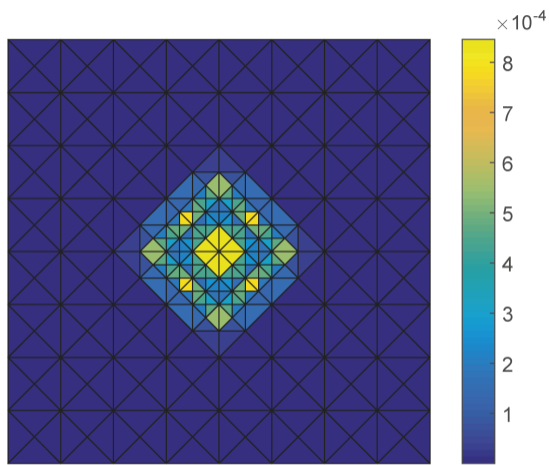
# Outline

- 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

# Where is the error localized?



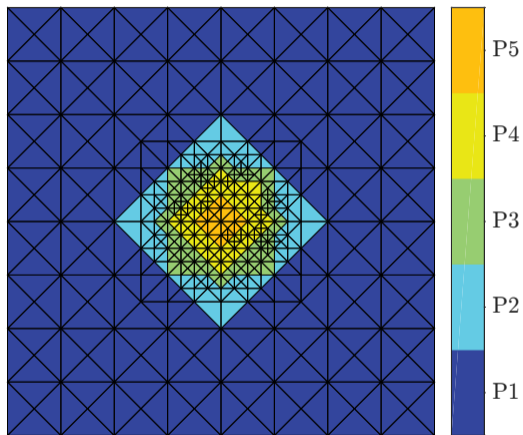
Estimated error distribution  $\eta_K(u_h)$



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

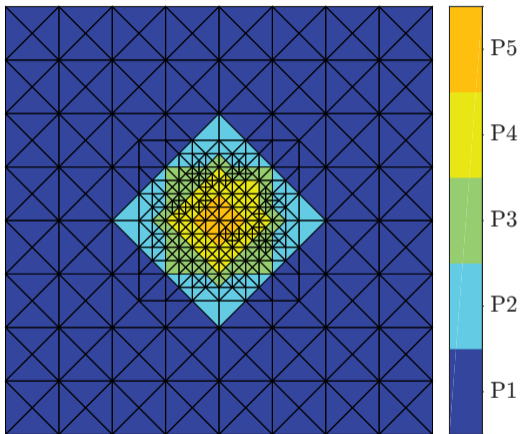
P. Daniel, A. Ern, I. Smears, M. Vohralík, Computers & Mathematics with Applications (2018)

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

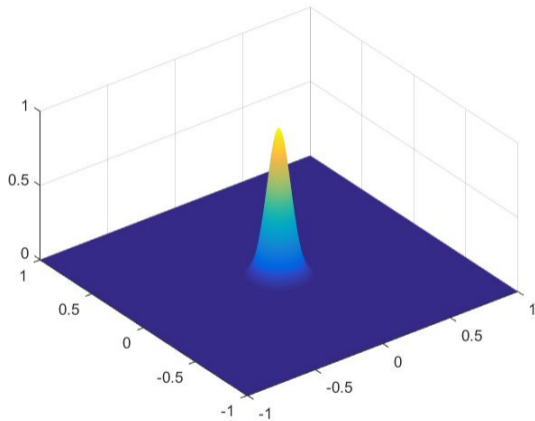


Mesh  $\mathcal{T}_\ell$  and pol. degrees  $p_K$

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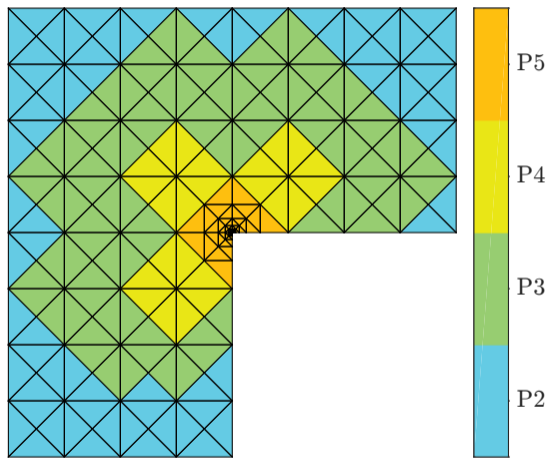
Mesh  $\mathcal{T}_\ell$  and pol. degrees  $p_K$



Exact solution

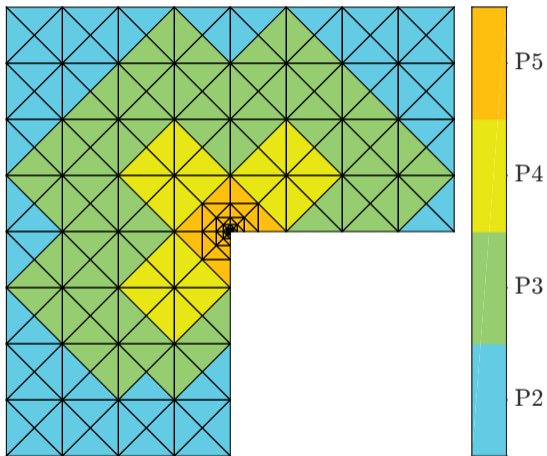
P. Daniel, A. Ern, I. Smears, M. Vohralík, Computers & Mathematics with Applications (2018)

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

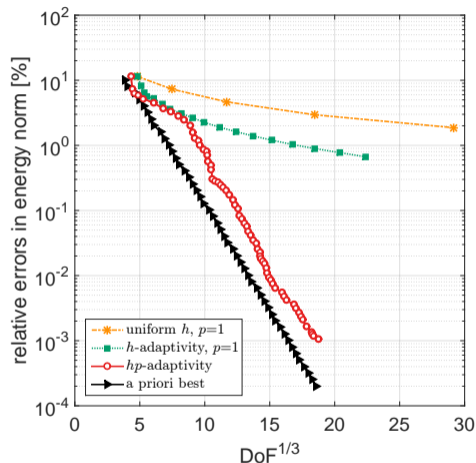


Mesh  $\mathcal{T}_\ell$  and polynomial degrees  $p_K$

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



Mesh  $\mathcal{T}_\ell$  and polynomial degrees  $p_K$



Relative error as a function of DoF



# 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  $V_\ell$

- $\|\nabla(u - u_\ell)\| \rightarrow 0$  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

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