

A posteriori error estimates robust with respect to the strength of nonlinearities

André Harnist, Koondanibha Mitra, Ari Rappaport, and **Martin Vohralík**

Inria Paris & Ecole des Ponts

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Inria



Outline

- 1 Introduction
 - Numerical approximation of partial differential equations
 - A posteriori error estimates
- 2 Setting and known results
 - Setting (gradient-dependent case)
 - Error measures
 - Known results
- 3 Iterative linearization
- 4 A posteriori error estimates for an augmented energy difference
- 5 Numerical experiments
- 6 Extensions
 - Setting (gradient-independent case)
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- 7 Conclusions

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Numerical approximation of partial differential equations

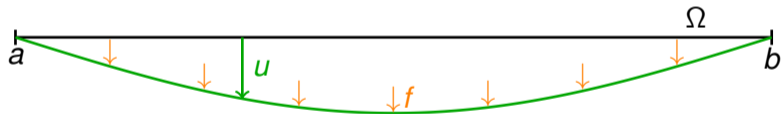
Numerical methods

- 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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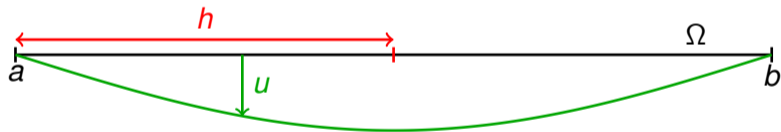


Numerical approximation u_h and its convergence to u

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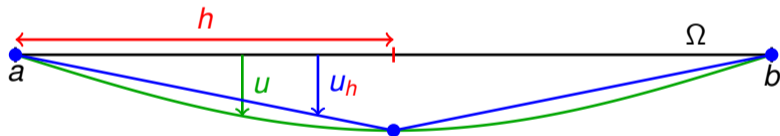


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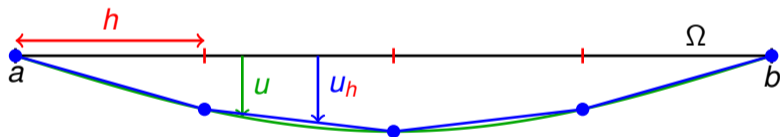


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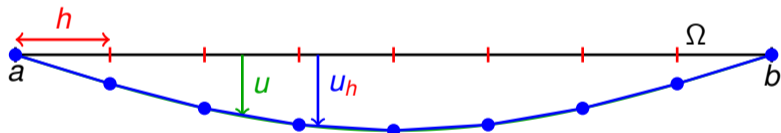


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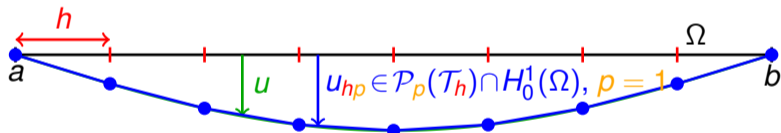


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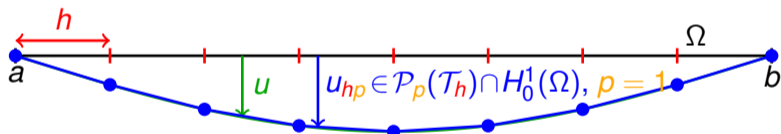


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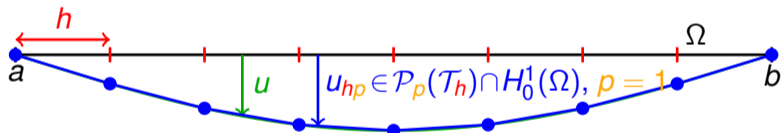
Error

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

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Numerical approximation u_{hp} and its convergence to u

Error

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

Need to solve a linear system

$$\mathbb{A}_{hp} \mathbf{U}_{hp} = \mathbf{F}_{hp}$$

3 crucial questions

Crucial questions

- 1 How **large** is the overall **error**?
- 2 **Where** (model/space/time/linearization/algebra) is it **localized**?
- 3 Can we **decrease** it **efficiently**?

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3 crucial questions & suggested answers

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Suggested answers

- 1 **A posteriori** error **estimates**.

3 crucial questions & suggested answers

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Suggested answers

- 1 **A posteriori** error **estimates**.
- 2 Identification of **error components**.

3 crucial questions & suggested answers

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- 1 How **large** is the overall **error**?
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Suggested answers

- 1 **A posteriori** error **estimates**.
- 2 Identification of **error components**.
- 3 **Balancing** error components, **adaptivity** (working where needed).

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



- no earthquake, flooding, tsunami, heavy rain, extreme temperature
- deterministic, steady problem, PDE known, data known, implementation OK

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



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



Y. El Kamari^a, W. Raphael^{a,*}, A. Chateaneuf^{b,c}

^aEcole Supérieure d'Ingénierie de Brest (ESIB), Université Saint-Joseph, CSF Mer Rivolin, PO Box 11-514, Brest CEDEX 11 29208, France

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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 in a FE discretization

Laplacian: find $u : \Omega \rightarrow \mathbb{R}$ such that

$$\begin{aligned} -\nabla \cdot (\nabla u) &= f && \text{in } \Omega, \\ u &= 0 && \text{on } \partial\Omega. \end{aligned}$$

Guaranteed error upper bound (reliability)

$$\underbrace{\|\nabla(u - u_\ell)\|}_{\text{unknown error}}$$

$$\underbrace{\eta(u_\ell)}_{\text{estimator computable from } u_\ell}$$

Error lower bound (efficiency)

$$\eta(u_\ell) \leq C_{\text{eff}} \|\nabla(u - u_\ell)\|$$

- C_{eff} a generic constant independent of Ω , u , u_ℓ and namely of the number of mesh elements $|\mathcal{T}_\ell|$ (h if \mathcal{T}_ℓ uniform) and of the polynomial degree p (for $d \leq 3$)

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

$h \approx 1/ \mathcal{T}_\ell ^{1/2}$	p	$\eta(u_\ell)$	rel. error estimate $\frac{\eta(u_\ell)}{\ \nabla u_\ell\ }$	$\ \nabla(u - u_\ell)\ $	rel. error $\frac{\ \nabla(u - u_\ell)\ }{\ \nabla u\ }$	$\gamma^{opt} = \frac{\eta(u_\ell)}{\ \nabla(u - u_\ell)\ }$
h_0	1	1.25	28%	1.07	24%	1.17
$\approx h_0/2$	1	0.625	14%	0.53	12%	1.17
$\approx h_0/4$	1	0.3125	7%	0.27	6%	1.17
$\approx h_0/8$	1	0.15625	4%	0.14	3%	1.17
$\approx h_0/2$	2	0.625	14%	0.53	12%	1.17
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A. Ern, M. Vohralik, SIAM Journal on Numerical Analysis (2015)
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h_0	1	1.25	28%	1.07	24%	1.17
$\approx h_0/2$		6.07×10^{-1}	25%	5.35×10^{-1}	24%	1.17
$\approx h_0/4$		3.10×10^{-1}	25%	2.82×10^{-1}	24%	1.17
$\approx h_0/8$		1.45×10^{-1}	25%	1.29×10^{-1}	24%	1.17
$\approx h_0/2$	2	4.23×10^{-2}	25%	3.71×10^{-2}	24%	1.17
$\approx h_0/4$	3	2.62×10^{-2}	25%	2.31×10^{-2}	24%	1.17
$\approx h_0/8$	4	2.60×10^{-2}	25%	2.30×10^{-2}	24%	1.17

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$\approx h_0/8$		1.45×10^{-1}	3.3%	1.39×10^{-1}	1.9%	
$\approx h_0/2$	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.07×10^{-2}		
$\approx h_0/4$	3	2.62×10^{-3}	$5.9 \times 10^{-2}\%$	2.60×10^{-3}		
$\approx h_0/8$	4	2.60×10^{-4}	$5.9 \times 10^{-3}\%$	2.58×10^{-4}		

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$\approx h_0/8$		1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	1.03
$\approx h_0/2$	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.07×10^{-2}	$9.2 \times 10^{-1}\%$	1.03
$\approx h_0/4$	3	2.62×10^{-2}	$5.9 \times 10^{-1}\%$	2.60×10^{-2}	$5.9 \times 10^{-1}\%$	1.02
$\approx h_0/8$	4	2.60×10^{-2}	$5.9 \times 10^{-1}\%$	2.58×10^{-2}	$5.8 \times 10^{-1}\%$	1.02

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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
$\approx h_0/2$	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.07×10^{-2}	$9.2 \times 10^{-1}\%$	1.04
$\approx h_0/4$	3	2.62×10^{-2}	$5.9 \times 10^{-1}\%$	2.60×10^{-2}	$5.9 \times 10^{-1}\%$	1.01
$\approx h_0/8$	4	2.60×10^{-2}	$5.9 \times 10^{-1}\%$	2.58×10^{-2}	$5.8 \times 10^{-1}\%$	1.01

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$\approx h_0/8$		1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	1.04
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$\approx h_0/4$	3	2.62×10^{-3}	$5.9 \times 10^{-3}\%$	2.60×10^{-3}	$5.9 \times 10^{-3}\%$	1.01
$\approx h_0/8$	4	2.60×10^{-4}	$5.9 \times 10^{-4}\%$	2.58×10^{-4}	$5.8 \times 10^{-4}\%$	1.01

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$h \approx 1/ \mathcal{T}_\ell ^{1/2}$	p	$\eta(u_\ell)$	rel. error estimate $\frac{\eta(u_\ell)}{\ \nabla u_\ell\ }$	$\ \nabla(u - u_\ell)\ $	rel. error $\frac{\ \nabla(u - u_\ell)\ }{\ \nabla u_\ell\ }$	$f^{\text{eff}} = \frac{\eta(u_\ell)}{\ \nabla(u - u_\ell)\ }$
h_0	1	1.25	28%	1.07	24%	1.17
$\approx h_0/2$		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%	1.06
$\approx h_0/8$		1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	1.04
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Linear problems

A posteriori error estimates **robust** with respect to the **discretization parameters** $|\mathcal{T}_\ell|$ (h if \mathcal{T}_ℓ uniform) and p ($d \leq 3$).

Nonlinear problems

A posteriori error estimates **robust** with respect to the **strength of nonlinearities**?

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A model nonlinear problem

Nonlinear elliptic problem

Find $u : \Omega \rightarrow \mathbb{R}$ such that

$$\begin{aligned} -\nabla \cdot (a(|\nabla u|) \nabla u) &= f \quad \text{in } \Omega, \\ u &= 0 \quad \text{on } \partial\Omega. \end{aligned}$$

- $\Omega \subset \mathbb{R}^d$, $d \geq 1$, open polytope with Lipschitz boundary $\partial\Omega$
- $f \in L^2(\Omega)$

Assumption (Nonlinear function a)

Function $a : [0, \infty) \rightarrow (0, \infty)$, for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^d$,

$$|a(|\mathbf{x}|)\mathbf{x} - a(|\mathbf{y}|)\mathbf{y}| \leq a_c |\mathbf{x} - \mathbf{y}| \quad (\text{Lipschitz continuity}),$$

$$(a(|\mathbf{x}|)\mathbf{x} - a(|\mathbf{y}|)\mathbf{y}) \cdot (\mathbf{x} - \mathbf{y}) \geq a_m |\mathbf{x} - \mathbf{y}|^2 \quad (\text{strong monotonicity}).$$

- $a_m \leq a(r) \leq a_c$, $a_m \leq (a(r)r)'(r) \leq a_c$

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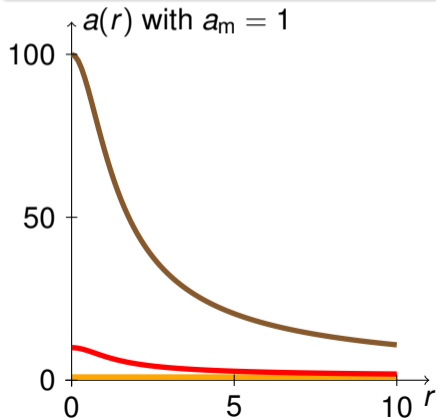
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Strength of the nonlinearity, $0 < a_m \leq a_c < \infty$ real parameters

Example (Mean curvature nonlinearity)

$$a(r) := a_m + \frac{a_c - a_m}{\sqrt{1 + r^2}}.$$

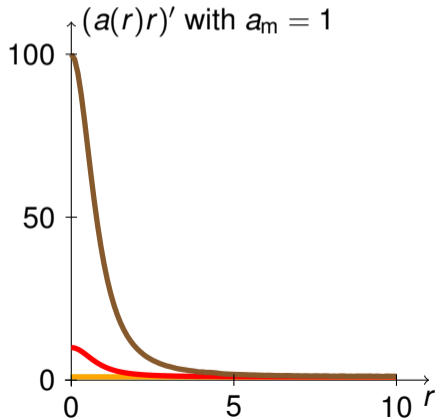
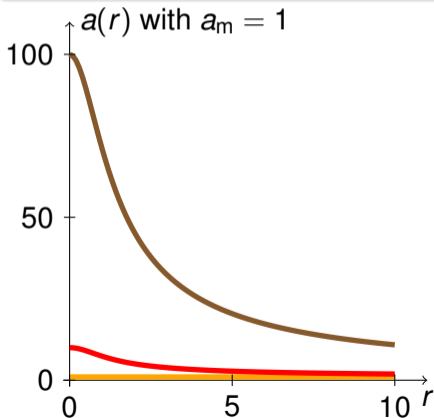


$$\begin{aligned} a_c &= 100 \\ a_c &= 10 \\ a_c &= 1 \end{aligned}$$

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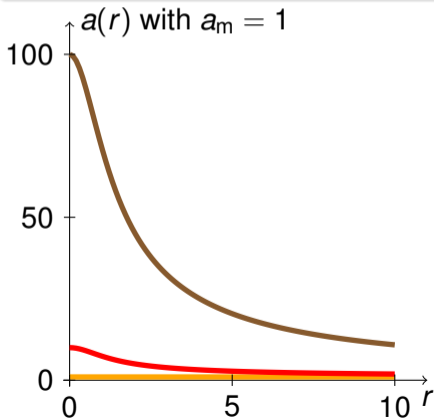


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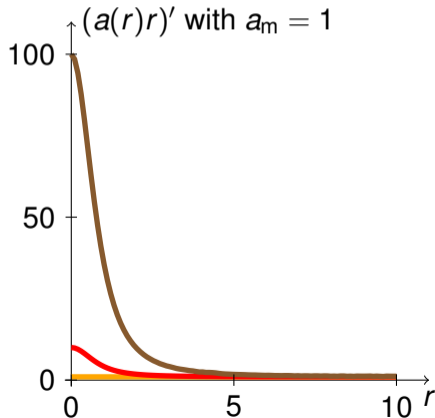
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$$\frac{a_c}{a_m} = \frac{\text{Lipschitz continuity}}{\text{strong monotonicity}}$$



Weak solution and its finite element approximation

Definition (Weak solution)

$u \in H_0^1(\Omega)$ such that

$$(a(|\nabla u|)\nabla u, \nabla v) = (f, v) \quad \forall v \in H_0^1(\Omega).$$

Definition (Finite element approximation)

$u_\ell \in V_\ell^p$ such that

$$(a(|\nabla u_\ell|)\nabla u_\ell, \nabla v_\ell) = (f, v_\ell) \quad \forall v_\ell \in V_\ell^p.$$

- \mathcal{T}_ℓ simplicial mesh of Ω
- $p \geq 1$ polynomial degree
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$$\mathcal{A}_\ell(U_\ell) = F_\ell$$

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Sobolev space and error

Sobolev space

$$H_0^1(\Omega)$$

Sobolev norm error

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

Energy and energy differences

Definition (Energy functional)

$$\mathcal{J} : H_0^1(\Omega) \rightarrow \mathbb{R}$$

$$\mathcal{J}(v) := \int_{\Omega} \phi(|\nabla v|) - (f, v), \quad v \in H_0^1(\Omega),$$

with function $\phi : [0, \infty) \rightarrow [0, \infty)$ such that, for all $r \in [0, \infty)$,

$$\phi(r) := \int_0^r a(s) s \, ds.$$

Equivalently

$$u = \arg \min_{v \in H_0^1(\Omega)} \mathcal{J}(v), \quad u_{\ell} = \arg \min_{v_{\ell} \in V_{\ell}^p} \mathcal{J}(v_{\ell}).$$

Energy difference

$$\mathcal{J}(u_{\ell}) - \mathcal{J}(u)$$

- $\mathcal{J}(u_{\ell}) - \mathcal{J}(u) \geq 0$, $\mathcal{J}(u_{\ell}) - \mathcal{J}(u) = 0$ if and only if $u_{\ell} = u$
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Definition (Residual)

$\mathcal{R} : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$; for $w \in H_0^1(\Omega)$, $\mathcal{R}(w) \in H^{-1}(\Omega)$ is given by

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Definition (Dual norm of the finite element residual)

$$||| \mathcal{R}(u_\ell) - \mathcal{R}(u) |||_{-1} = \boxed{||| \mathcal{R}(u_\ell) |||_{-1}} := \sup_{v \in H_0^1(\Omega)} \frac{\langle \mathcal{R}(u_\ell), v \rangle}{||| v |||}.$$

- $||| \mathcal{R}(u_\ell) |||_{-1} \geq 0$, $||| \mathcal{R}(u_\ell) |||_{-1} = 0$ if and only if $u_\ell = u$
- subordinate to the choice of the norm $||| \cdot |||$ on the Sobolev space $H_0^1(\Omega)$
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Known results

Sobolev norm

$$a_m \|\nabla(u_\ell - u)\| \leq \eta(u_\ell) \leq C_{\text{eff}} a_c \|\nabla(u_\ell - u)\|$$

- Pousin & Rappaz (1994), Verfürth (1994), Kim (2007), Houston, Süli, & Wihler (2008), Garau, Morin, & Zuppa (2011), Gantner, Haberl, Praetorius, & Stiftner (2018), Heid & Wihler (2020), Botti & Riedlbeck (2020), ...

Energy difference

$$\mathcal{J}(u_\ell) - \mathcal{J}(u) \leq \frac{1}{2} \eta(u_\ell)^2 \leq C_{\text{eff}}^2 \frac{a_c^2}{a_m^2} (\mathcal{J}(u_\ell) - \mathcal{J}(u))$$

Strength of the nonlinearity

Not robust with respect to $\frac{a_c}{a_m}$.

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$$\mathcal{J}(u_\ell) - \mathcal{J}(u) \leq \frac{1}{2} \eta(u_\ell)^2 \leq C_{\text{eff}}^2 \frac{a_c^2}{a_m^2} (\mathcal{J}(u_\ell) - \mathcal{J}(u))$$

- Zangl (2018), Zangl & Wollmann (2020)
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- Strength of the nonlinearity
- Not robust with respect to $\frac{a_c}{a_m}$

Known results

Sobolev norm

$$a_m \|\nabla(u_\ell - u)\| \leq \eta(u_\ell) \leq C_{\text{eff}} a_c \|\nabla(u_\ell - u)\|$$

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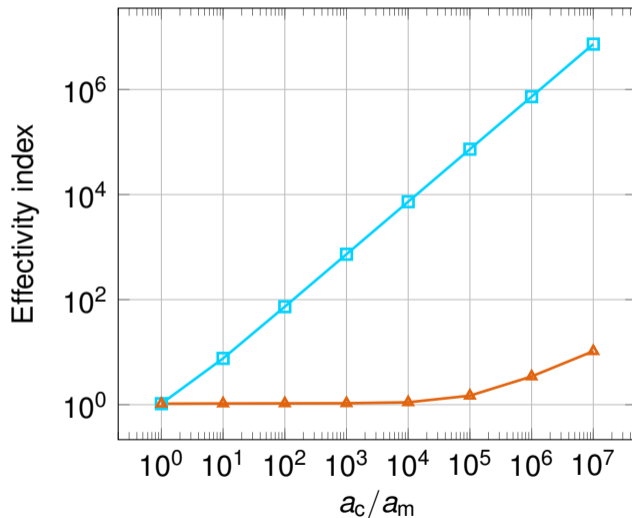
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Non-robustness



—□— Sobolev norm $\|\nabla(u_\ell - u)\|$
—▲— Energy difference $\mathcal{J}(u_\ell) - \mathcal{J}(u)$

Known results

Dual norm of the residual

$$|||\mathcal{R}(u_\ell)|||_{-1} \leq \eta(u_\ell) \leq C_{\text{eff}} |||\mathcal{R}(u_\ell)|||_{-1}$$

- Chaillou & Suri (2006), El Alaoui, Ern, & Vohralík (2011), Blechta, Málek, & Vohralík (2020), ...

Strength of the nonlinearity

- **Robust** with respect to $\frac{a_c}{a_m}$ if $|||v||| = \|\nabla v\|$.
- $|||\mathcal{R}(u_\ell)|||_{-1}$ localizes over patches of elements.
- $|||\mathcal{R}(u_\ell)|||_{-1}$ is a $H^{-1}(\Omega)$ residual norm (essentially estimates the flux error $\|a(|\nabla u_\ell|)\nabla u_\ell - a(|\nabla u|)\nabla u\|$).

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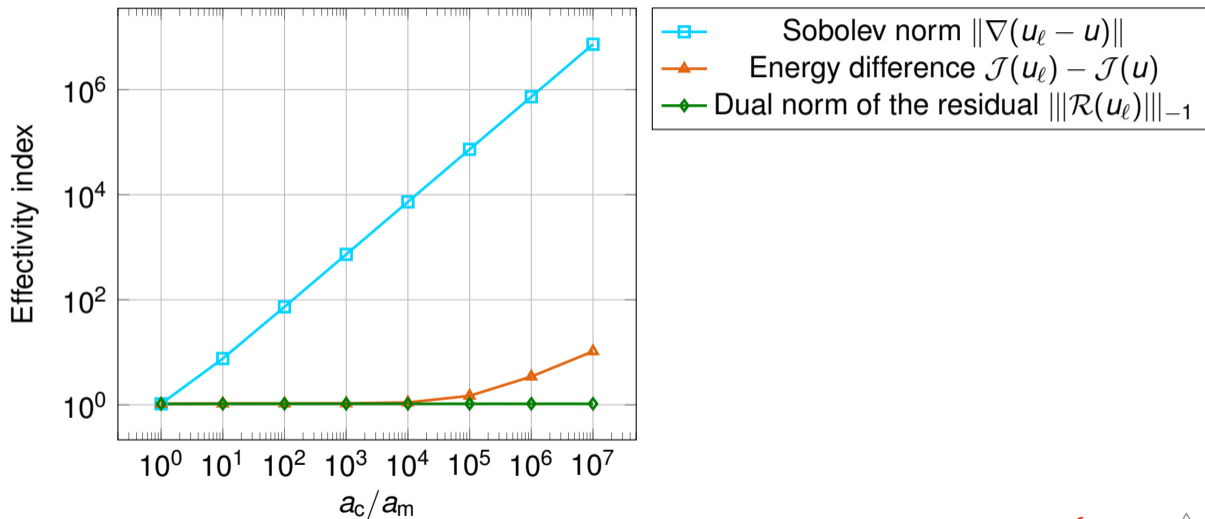
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Non-robustness



Nonlinear problems

A posteriori error estimates **robust** with respect to the **strength of nonlinearities** in more **physically-based error measures**?

Iterative linearization

Addressing iterative linearization

- Chaillou & Suri (2006), Ern & Vohralík (2013), Bernardi, Dakroub, Mansour, & Sayah (2015), Gantner, Haberl, Praetorius, & Stiftner (2018), Heid & Wihler (2020), Botti & Riedlbeck (2020), ...

Observation

None of the above approaches employ **in the analysis**, to define norms, the **iterative linearization**, i.e., **how** do we solve the nonlinear system $\mathcal{A}_\ell(\mathbf{U}_\ell) = \mathbf{F}_\ell$.

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Iterative linearization

Definition (Linearized finite element approximation)

$u_\ell^k \in V_\ell^p$ such that

$$(\mathbf{A}_\ell^{k-1} \nabla u_\ell^k, \nabla v_\ell) = (f, v_\ell) + (\mathbf{b}_\ell^{k-1}, \nabla v_\ell) \quad \forall v_\ell \in V_\ell^p.$$

- $u_\ell^0 \in V_\ell^p$ a given initial guess
- iterative linearization index $k \geq 1$
- **linearization**: $\mathbf{A}_\ell^{k-1}: \Omega \rightarrow \mathbb{R}^{d \times d}$ diffusion matrix, $\mathbf{b}_\ell^{k-1}: \Omega \rightarrow \mathbb{R}^d$ RHS vector

Definition (Linearized energy functional)

$$\mathcal{J}_\ell^{k-1}: H_0^1(\Omega) \rightarrow \mathbb{R}$$

$$\mathcal{J}_\ell^{k-1}(v) := \frac{1}{2} \left\| (\mathbf{A}_\ell^{k-1})^{\frac{1}{2}} \nabla v \right\|^2 - (f, v) - (\mathbf{b}_\ell^{k-1}, \nabla v), \quad v \in H_0^1(\Omega).$$

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$$u_\ell^k := \arg \min_{v_\ell \in V_\ell^p} \mathcal{J}_\ell^{k-1}(v_\ell)$$

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$$\mathbf{A}_\ell^{k-1} = a(|\nabla u_\ell^{k-1}|) \mathbf{I}_d, \quad \mathbf{b}_\ell^{k-1} = \mathbf{0}.$$

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$$\mathbf{A}_\ell^{k-1} = \gamma \mathbf{I}_d, \quad \mathbf{b}_\ell^{k-1} = (\gamma - a(|\nabla u_\ell^{k-1}|)) \nabla u_\ell^{k-1},$$

with $\gamma \geq \frac{a_c^2}{a_m}$ a constant parameter.

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$$\mathbf{A}_\ell^{k-1} = a(|\nabla u_\ell^{k-1}|) \mathbf{I}_d + \theta \frac{a'(|\nabla u_\ell^{k-1}|)}{|\nabla u_\ell^{k-1}|} \nabla u_\ell^{k-1} \otimes \nabla u_\ell^{k-1},$$

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A posteriori error estimates for an augmented energy difference

Theorem (A posteriori estimate of energy)

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- C_ℓ^k **computable**: we can affirm **robustness a posteriori**, for the given case

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Moreover, for k satisfying a stopping criterion, there holds

$$\eta_\ell^k \leq C_{\text{eff}}(d, \kappa_{\mathcal{T}}, p \text{ if } d \geq 4) C_\ell^k \varepsilon_\ell^k,$$

where

$$C_\ell^k := \max_{\mathbf{a} \in \mathcal{V}_\ell} \left(\frac{\sup_{\omega_\ell^{\mathbf{a}}} A_{\mathbf{c},\ell}^{k-1}}{\inf_{\omega_\ell^{\mathbf{a}}} A_{\mathbf{m},\ell}^{k-1}} \right) \begin{cases} = 1 & \text{Zarantonello} \\ \leq \frac{A_{\mathbf{c}}}{A_{\mathbf{m}}} \leq \frac{a_{\mathbf{c}}}{a_{\mathbf{m}}} & \text{in general.} \end{cases}$$

- $C_\ell^k = 1$ for Zarantonello \implies **robustness** wrt the **strength of nonlinearities**
- C_ℓ^k given by **local** (patch) **properties**: typically **much better** than $a_{\mathbf{c}}/a_{\mathbf{m}}$
- C_ℓ^k **computable**: we can affirm **robustness a posteriori**, for the given case

A posteriori error estimates for an augmented energy difference

Theorem (A posteriori estimate of augmented energy)

Let $f \in \mathcal{P}_{p-1}(\mathcal{T}_\ell)$ for simplicity. For all linearization steps $k \geq 1$,

$$\underbrace{\mathcal{J}(u_\ell^k) - \mathcal{J}(u)}_{\text{energy difference}} + \underbrace{\lambda_\ell^k \left(\mathcal{J}_\ell^{k-1}(u_\ell^k) - \mathcal{J}_\ell^{k-1}(u_{\langle \ell \rangle}^k) \right)}_{\text{en. diff. linearization}} \leq \underbrace{\mathcal{J}(u_\ell^k) - \mathcal{J}^*(\sigma_\ell^k)}_{\text{en. diff. estimate}} + \underbrace{\lambda_\ell^k \left(\mathcal{J}_\ell^{k-1}(u_\ell^k) - \mathcal{J}_\ell^{*,k-1}(\sigma_\ell^k) \right)}_{\text{en. diff. linearization estimate}}.$$

Moreover, for k satisfying a stopping criterion, there holds

$$\eta_\ell^k \leq C_{\text{eff}}(d, \kappa_{\mathcal{T}}, p \text{ if } d \geq 4) C_\ell^k \mathcal{E}_\ell^k,$$

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- λ_ℓ^k computable weight to make the two components comparable
- $C_\ell^k = 1$ for Zarantonello \implies **robustness** wrt the **strength of nonlinearities**
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Smooth solution

Setting

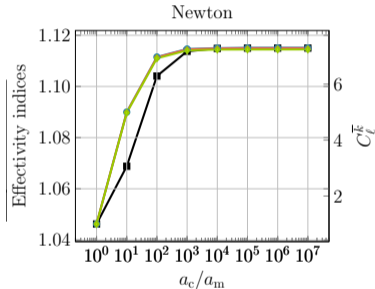
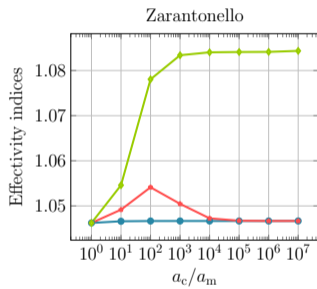
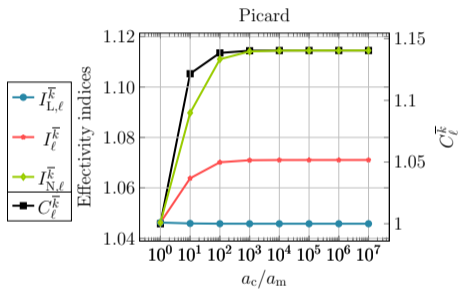
- unit square $\Omega = (0, 1)^2$
- known smooth solution $u(x, y) := 10 x(x - 1)y(y - 1)$
- mean curvature or exponential nonlinearity

$$a(r) = a_m + \frac{a_c - a_m}{\sqrt{1 + r^2}} \quad \text{or} \quad a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}}$$

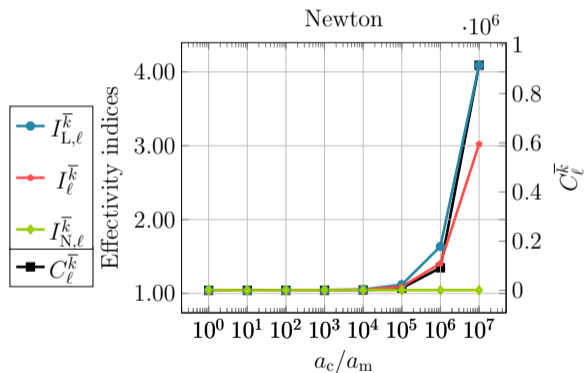
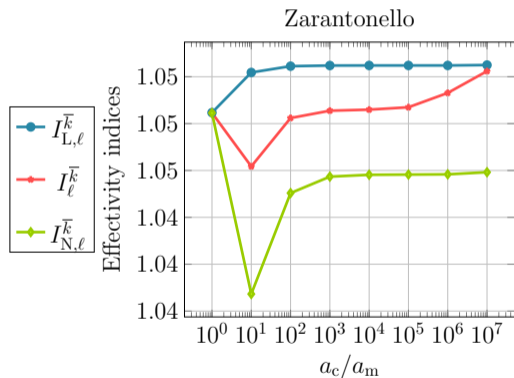
- $p = 1$, 3969 DOFs
- stopping iteration \bar{k} such that $\|\nabla(u_{\ell}^{\bar{k}-1} - u_{\ell}^{\bar{k}})\| < 10^{-6}$
- effectivity indices

$$\underbrace{I_{\ell}^k := \left(\frac{\eta_{\ell}^k}{\varepsilon_{\ell}^k} \right)^{\frac{1}{2}}}_{\text{total}}, \quad \underbrace{I_{N,\ell}^k := \left(\frac{\mathcal{J}(u_{\ell}^k) - \mathcal{J}^*(\sigma_{\ell}^k)}{\mathcal{J}(u_{\ell}^k) - \mathcal{J}(u)} \right)^{\frac{1}{2}}}_{\text{energy difference}}, \quad \underbrace{I_{L,\ell}^k := \left(\frac{\mathcal{J}_{\ell}^{k-1}(u_{\ell}^k) - \mathcal{J}_{\ell}^{*,k-1}(\sigma_{\ell}^k)}{\mathcal{J}_{\ell}^{k-1}(u_{\ell}^k) - \mathcal{J}_{\ell}^{k-1}(u_{\langle \ell \rangle}^k)} \right)^{\frac{1}{2}}}_{\text{energy difference linearization}}$$

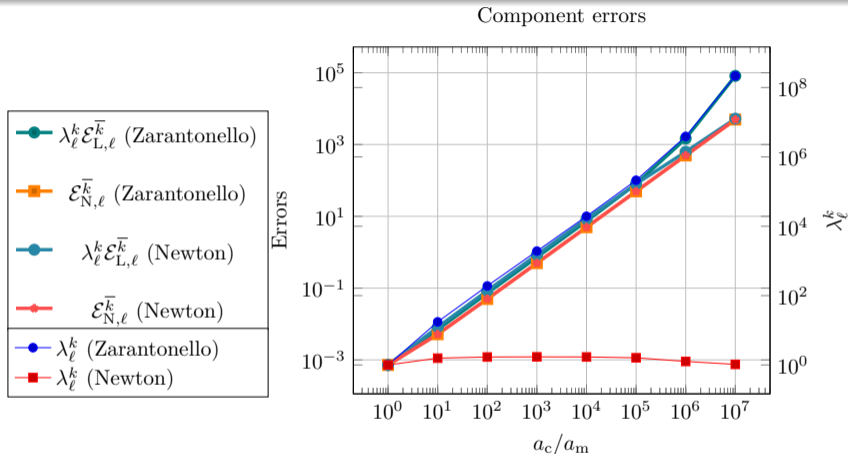
Error certification robust wrt the nonlinearities (mean curvature)



Error certification robust wrt the nonlinearities (exponential, robustness only for Zarantonello)



Error certification robust wrt the nonlinearities (exponential)



Components of $\mathcal{E}_\ell^{\bar{k}}$: the energy difference $\mathcal{E}_{N,\ell}^{\bar{k}} = \mathcal{J}(u_\ell^{\bar{k}}) - \mathcal{J}(u)$, the energy difference of the linearized problem $\mathcal{E}_{L,\ell}^{\bar{k}} = \mathcal{J}_\ell^{\bar{k}-1}(u_\ell^{\bar{k}}) - \mathcal{J}_\ell^{\bar{k}-1}(u_{(\ell)}^{\bar{k}})$, & the weight λ_ℓ^k

Singular solution

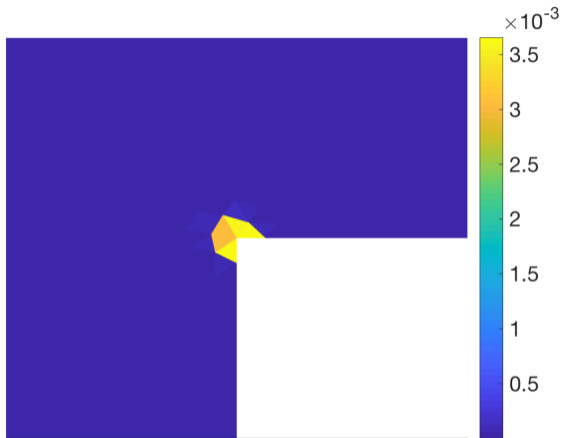
Setting

- L-shaped domain $\Omega = (-1, 1)^2 \setminus ([0, 1] \times (-1, 0])$
- known singular solution $u(\rho, \theta) = \rho^{\frac{2}{3}} \sin(\frac{2}{3}\theta)$
- exponential nonlinearity

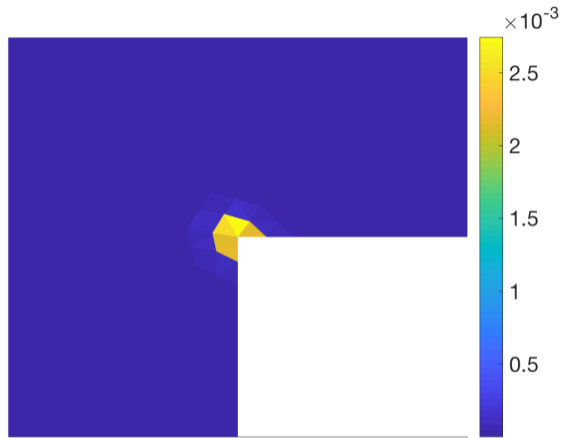
$$a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}}$$

- $p = 1$, uniform or adaptive mesh refinement

Where is the error localized?



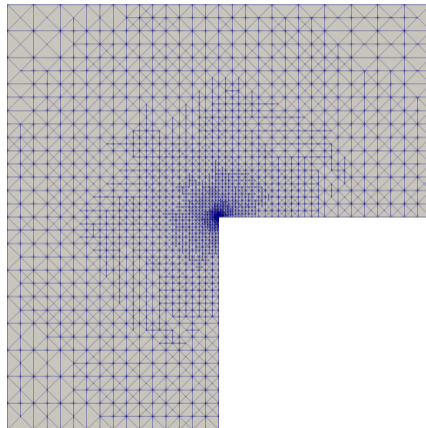
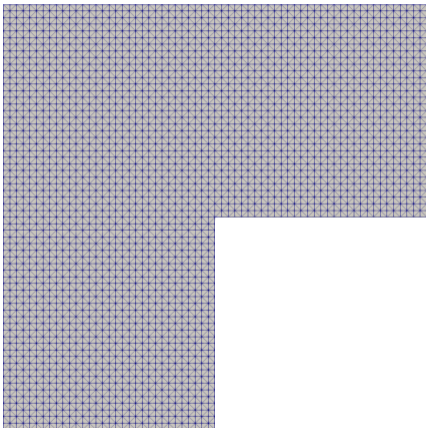
Estimated total errors $\eta_K(u_\ell^i)$



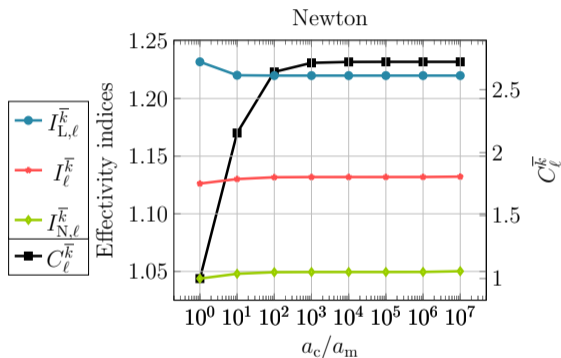
Exact total errors $\|\nabla(u - u_\ell^i)\|_K$

J. Papež, U. Rūde, M. Vohralík, B. Wohlmuth, Computer Methods in Applied Mechanics and Engineering (2020)

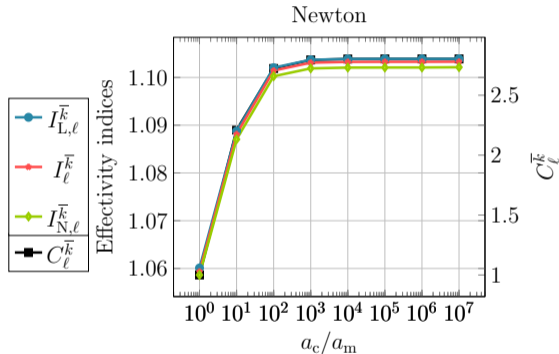
Adaptive mesh refinement



Error certification robust wrt the nonlinearities

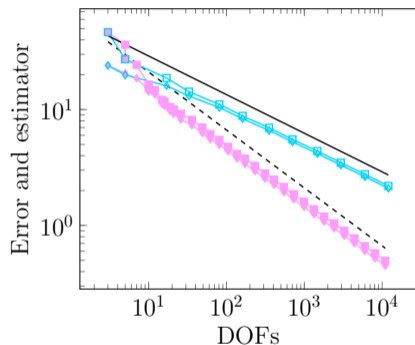


Uniform mesh refinement

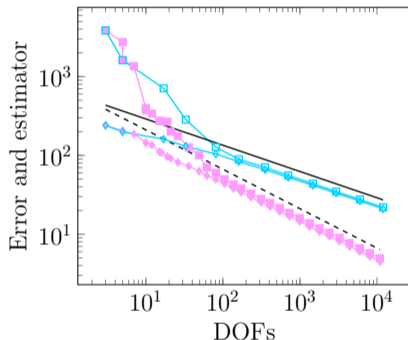
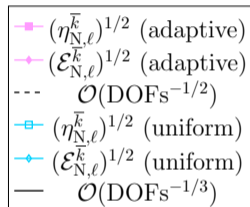


Adaptive mesh refinement

Decreasing the error efficiently: optimal decay rate wrt DoFs



$$\frac{a_c}{a_m} = 10^3$$



$$\frac{a_c}{a_m} = 10^6$$

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A model nonlinear problem

Nonlinear elliptic problem

Find $u : \Omega \rightarrow \mathbb{R}$ such that

$$-\nabla \cdot (\underbrace{\tau \mathbf{K}(\mathbf{x})}_{\text{diffusion}} \underbrace{(\mathcal{D}(\mathbf{x}, u) \nabla u + \mathbf{q}(\mathbf{x}, u))}_{\text{advection}}) + \underbrace{f(\mathbf{x}, u)}_{\text{reaction}} = 0 \quad \text{in } \Omega,$$

$$u = 0 \quad \text{on } \partial\Omega.$$

- $\tau > 0$ a parameter (time step size in transient problems)

Assumption (Nonlinear functions \mathcal{D} , \mathbf{q} , and f)

$$|\mathcal{D}(\mathbf{x}_1, \xi_1) - \mathcal{D}(\mathbf{x}_2, \xi_2)| \leq \mathcal{D}_M (|\mathbf{x}_1 - \mathbf{x}_2| + |\xi_1 - \xi_2|) \quad \forall \mathbf{x}_1, \mathbf{x}_2 \in \Omega \text{ and } \xi_1, \xi_2 \in \mathcal{R},$$

$$0 \leq f(\mathbf{x}, \xi_2) - f(\mathbf{x}, \xi_1) \leq f_M (\xi_2 - \xi_1) \quad \forall \mathbf{x} \in \Omega \text{ and } \xi_2 \geq \xi_1,$$

\mathbf{q} is "small" wrt $\mathbf{K}\mathcal{D}$.

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Strength of the nonlinearity

ratio a_c/a_m

Iterative linearization

Definition (Linearized finite element approximation)

$u_\ell^k \in V_\ell^p$ such that

$$\left((u_\ell^k - u_\ell^{k-1}, v_\ell) \right)_{u_\ell^{k-1}} = - \underbrace{\langle \mathcal{R}(u_\ell^{k-1}), v_\ell \rangle}_{\text{residual}} \quad \forall v_\ell \in V_\ell^p.$$

- linearization: **reaction–diffusion scalar product**

$$\left((w, v) \right)_{u_\ell^{k-1}} := \underbrace{\left(L(\mathbf{x}, u_\ell^{k-1}) w, v \right)}_{\text{reaction coef.}} + \underbrace{\left(\alpha(\mathbf{x}, u_\ell^{k-1}) \nabla w, \nabla v \right)}_{\text{diffusion coef.}}, \quad w, v \in H_0^1(\Omega)$$

- covers many linearization schemes: Picard (fixed-point), L & M-schemes, ...

Iteration-dependent norm

$$\| \| v \| \|_{1, u_\ell^{k-1}} := \left((v, v) \right)_{u_\ell^{k-1}}^{\frac{1}{2}}, \quad v \in H_0^1(\Omega)$$

- induced by the linearization scalar product

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An orthogonal decomposition of the total residual/error

Theorem (Orthogonal decomposition of the total error into linearization and discretization components)

For all linearization steps $k \geq 1$, there holds

$$\underbrace{\|\mathcal{R}(u_\ell^{k-1})\|_{-1, u_\ell^{k-1}}^2}_{\substack{\text{total residual/error} \\ \|\mathbf{u}_\ell^{k-1} - \mathbf{u}_{\langle \ell \rangle}^k\|_{1, u_\ell^{k-1}}}} = \underbrace{\|\mathbf{u}_\ell^{k-1} - \mathbf{u}_\ell^k\|_{1, u_\ell^{k-1}}^2}_{\substack{\text{linearization} \\ \text{error}}} + \underbrace{\|\mathcal{R}_{\text{disc}}^{u_\ell^{k-1}}(\mathbf{u}_\ell^k)\|_{-1, u_\ell^{k-1}}^2}_{\substack{\text{discretization residual/error} \\ \|\mathbf{u}_\ell^k - \mathbf{u}_{\langle \ell \rangle}^k\|_{1, u_\ell^{k-1}}}}.$$

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$$C_\ell^k \left\{ \begin{array}{l} = 1 \end{array} \right. \quad \text{Zarantonello}$$

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where

$$C_\ell^k := \max_{\mathbf{a} \in \mathcal{V}_\ell} \left(\frac{h_{\omega_a}^2 \sup_{\omega_a} L_M^{k-1} + \pi^2 \sup_{\omega_a} \mathbf{a}_M^{k-1}}{h_{\omega_a}^2 \inf_{\omega_a} L_m^{k-1} + \pi^2 \inf_{\omega_a} \mathbf{a}_m^{k-1}}, \frac{\sup_{\omega_a} \mathbf{a}_M^{k-1}}{\inf_{\omega_a} \mathbf{a}_m^{k-1}} \right) \left\{ \begin{array}{l} = 1 \\ \text{Zarantonello} \end{array} \right.$$

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- $C_\ell^k = 1$ for Zarantonello \implies **robustness** wrt the **strength of nonlinearities**
- C_ℓ^k given by **local** (patch) **properties**:

A posteriori error estimates for an iteration-dependent norm

Theorem (A posteriori estimate of iteration-dependent norm)

For all linearization steps $k \geq 1$,

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- also **local efficiency**

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The Richards equation

Setting

- one time step of the Richards equation
- unit square $\Omega = (0, 1)^2$
- realistic data

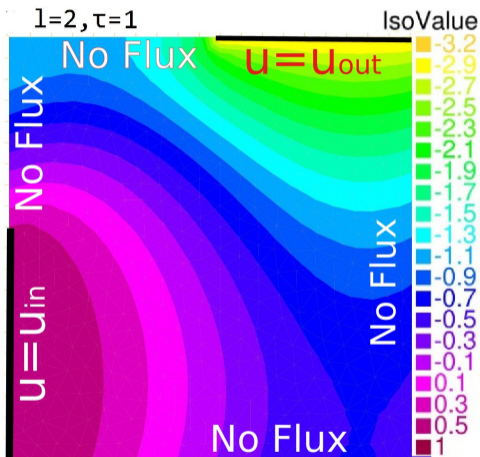
$$f(\mathbf{x}, \xi) = S(\xi) - S(u_\ell^{n-1}(\mathbf{x})), \quad \mathcal{D}(\mathbf{x}, \xi) = \kappa(S(\xi)), \quad \mathbf{q}(\mathbf{x}, \xi) = -\kappa(S(\xi)) \mathbf{g},$$

$$\underline{\mathbf{K}} = \begin{bmatrix} 1 & 0.2 \\ 0.2 & 1 \end{bmatrix}, \quad \mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

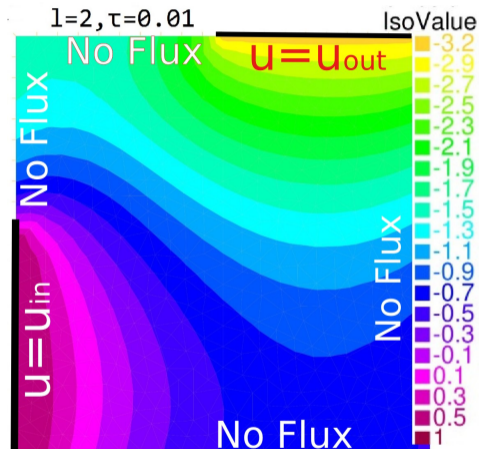
- time step length $\tau \in [10^{-3}, 1]$
- van Genuchten saturation and permeability laws

$$S(\xi) := \left(1 + (2 - \xi)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \quad \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^{\lambda}\right)^2, \quad \lambda = 0.5$$

One time step of the Richards equation with realistic data

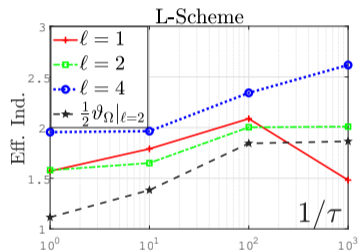
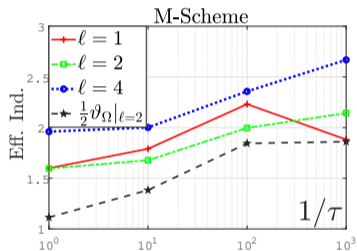
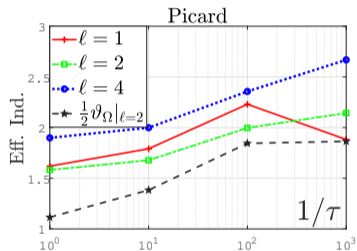


Time step length $\tau = 1$

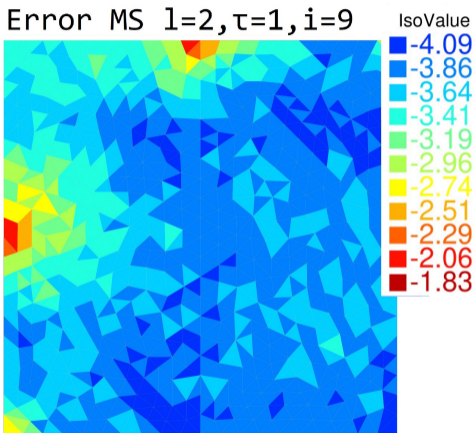


Time step length $\tau = 0.01$

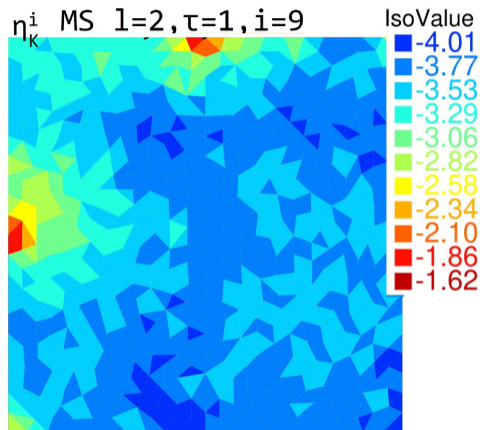
Error certification robust wrt the nonlinearities



Where is the error localized?



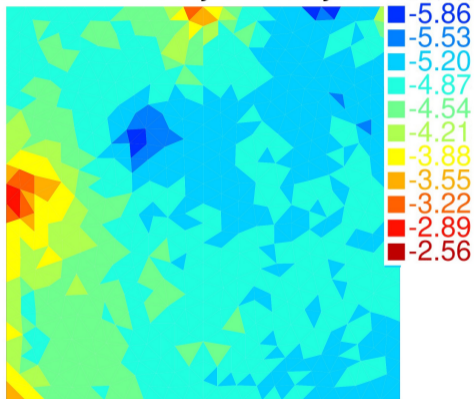
Error, $\tau = 1$



Estimate, $\tau = 1$

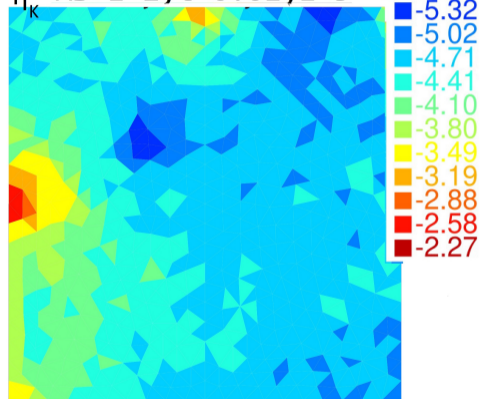
Where is the error localized?

Error MS $l=2, \tau=0.01, i=5$



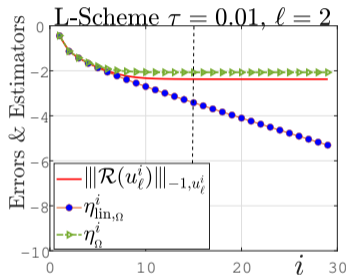
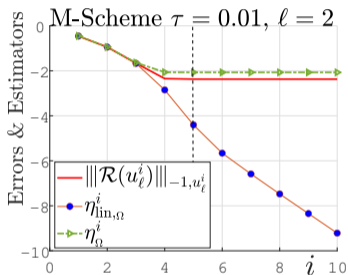
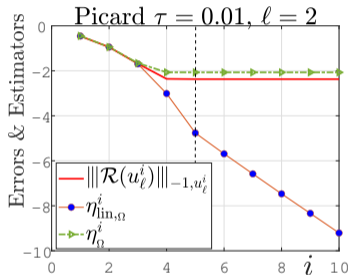
Error, $\tau = 0.01$

η_k^i MS $l=2, \tau=0.01, i=5$



Estimate, $\tau = 0.01$

Error components and adaptivity via stopping criteria



Time step length $\tau = 0.01$

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Thank you for your attention!