

A posteriori error estimates robust with respect to nonlinearities and orthogonal decomposition based on iterative linearization

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- 1 Introduction
- 2 Iteration-dependent norms
 - Setting
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 - Discretization and iterative linearization
 - Iteration-dependent norm and orthogonal decomposition
 - A posteriori error estimates
 - Numerical experiments
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 - Fenchel conjugate, dual energy, estimator, flux equilibration
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- 4 Conclusions

Goals

Error control

a posteriori error estimates

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

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efficient

$$\underbrace{|||u - u_\ell|||}_{\text{unknown error}} \leq \underbrace{\eta(u_\ell)}_{\text{computable estimator}} \leq C_{\text{eff}} |||u - u_\ell|||,$$

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Guaranteed a posteriori error estimates **efficient** and **robust** with respect to the **strength of nonlinearities**:

$$\underbrace{|||u - u_\ell|||}_{\text{unknown error}} \leq \underbrace{\eta(u_\ell)}_{\text{computable estimator}} \leq C_{\text{eff}} |||u - u_\ell|||, \quad C_{\text{eff}} \text{ independent of } \textit{data}$$

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$$\underbrace{|||u - u_\ell|||}_{\text{unknown error}} \leq \left\{ \sum_{K \in \mathcal{T}_\ell} \underbrace{\eta_K(u_\ell)^2}_{\text{element estimator}} \right\}^{1/2} \leq C_{\text{eff}} |||u - u_\ell|||,$$

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$$\eta_K(u_\ell) \leq C_{\text{eff}} |||u - u_\ell|||_{\omega_K}, \quad \text{for all } K \in \mathcal{T}_\ell.$$

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Question

- what to choose for $||| \cdot |||$?

Previous results

Model nonlinear elliptic problem

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

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

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- $\Omega \subset \mathbb{R}^d$, $1 \leq d \leq 3$, open bounded polytope with Lipschitz boundary $\partial\Omega$
- a strongly monotone (a_m) and Lipschitz continuous (a_c)
- f piecewise polynomial for simplicity
- numerical approximation u_ℓ

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- f piecewise polynomial for simplicity
- numerical approximation u_ℓ
- strength of the nonlinearity (“nonlinear condition number”): a_c/a_m

Previous 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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Sobolev norm (**not robust** wrt $\frac{a_c}{a_m}$)

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Energy difference

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

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- El Alaoui, Ern, & Vohralík (2011), Blechta, Málek, & Vohralík (2020), ... *Inria*

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A model nonlinear problem I

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

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Assumption (Gradient-dependent diffusivity)

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

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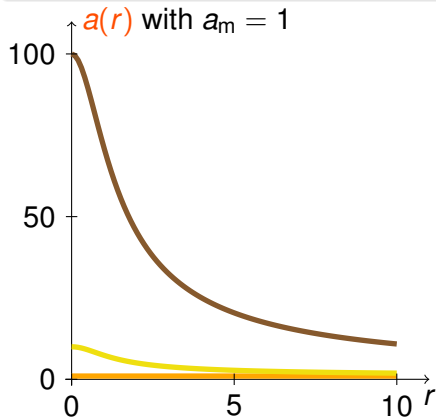
Example (Mean curvature nonlinearity)

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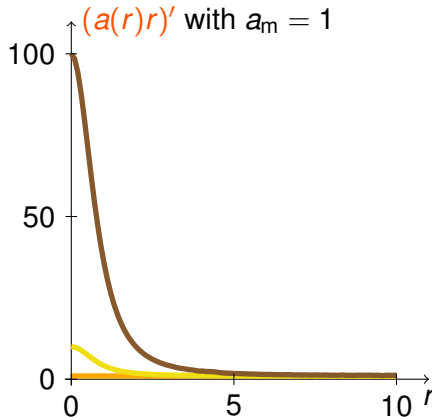
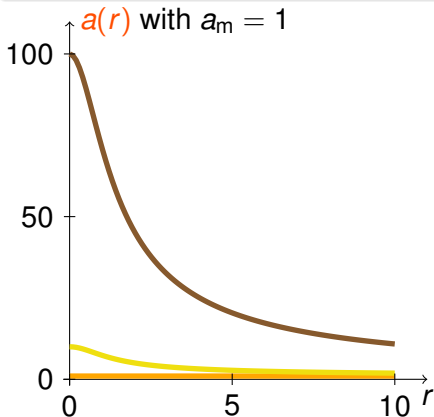


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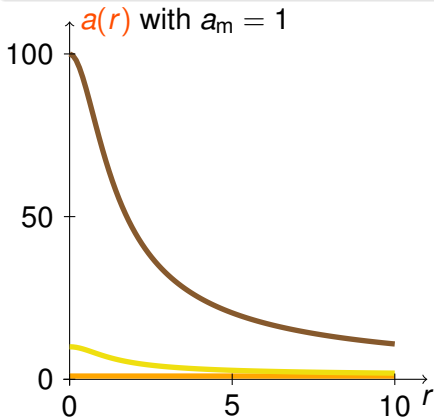


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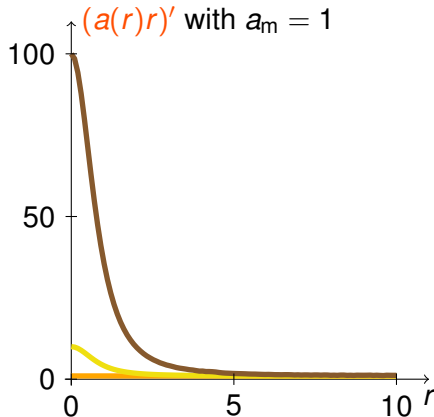
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Strength of the nonlinearity

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



A model nonlinear problem II

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

$$\langle \mathcal{R}(u), v \rangle := \left(\tau \underbrace{\mathbf{K}(\mathbf{a}(u))}_{\text{diffusion}} \nabla u + \underbrace{\mathbf{q}(u)}_{\text{advection}}, \nabla v \right) + \left(\underbrace{f(u)}_{\text{reaction}}, v \right) = 0 \quad \forall v \in H_0^1(\Omega).$$

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- semilinear equations with $a = 1$, $\mathbf{K} = \mathbf{I}_d$, $\tau = 1$, $\mathbf{q} = \mathbf{0}$: ignition of gases, gravitational influences on stars, quantum field theory ...
- implicit time-discretization of nonlinear (degenerate) parabolic equations ($\tau > 0$ a time step size): Fischer–KPP, porous medium, Richards, biofilms ...

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Assumption (Gradient-independent diffusivity)

- $|a(\mathbf{x}_1, u_1) - a(\mathbf{x}_2, u_2)| \leq L_a(|\mathbf{x}_1 - \mathbf{x}_2| + |u_1 - u_2|) \quad \forall \mathbf{x}_1, \mathbf{x}_2 \in \Omega \text{ and } u_1, u_2 \in \mathbb{R}$
- $0 \leq f(\mathbf{x}, u_2) - f(\mathbf{x}, u_1) \leq L_f(u_2 - u_1) \quad \forall \mathbf{x} \in \Omega \text{ and } u_1, u_2 \in \mathbb{R}, u_2 \geq u_1$
- $\mathbf{K}$ is uniformly symmetric positive definite and bounded with eigenvalues a_m, a_c
- $\mathbf{q}$ is “small” wrt a

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Discretization and fixed-point iterative linearization

- **discretization**: finite element subspace $V_\ell \subset H_0^1(\Omega)$, find $u_\ell \in V_\ell$ s.t.

$$(\mathbf{K} \mathbf{a}(u_\ell) \nabla u_\ell, \nabla v_\ell) + (f, v_\ell) = 0 \quad \forall v_\ell \in V_\ell.$$

Trivial nonlinear case

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

$$\langle \mathcal{R}(u), v \rangle := (\mathbf{K} \mathbf{a}(u) \nabla u, \nabla v) + (f, v) = 0 \quad \forall v \in H_0^1(\Omega).$$

Ideal but impossible choice

$$||| \cdot ||| := \| \mathbf{K}^{1/2} \mathbf{a}^{1/2}(u) \nabla(\cdot) \|$$

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- fixed-point **iterative linearization**: from $u_\ell^{k-1} \in V_\ell$, find $u_\ell^k \in V_\ell$ such that

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Trivial nonlinear case: main idea

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Iteration-dependent discrete energy norm

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Iteration-dependent discrete energy and dual norms

$$||| \cdot |||_{u_\ell^{k-1}} := \| \mathbf{K}^{1/2} \mathbf{a}^{1/2}(u_\ell^{k-1}) \nabla(\cdot) \|, \quad ||| \mathcal{R}(u_\ell) |||_{-1, u_\ell^{k-1}} := \sup_{v \in H_0^1(\Omega)} \frac{\langle \mathcal{R}(u_\ell), v \rangle}{||| v |||_{u_\ell^{k-1}}}$$

Main idea

Main idea

Apply in the **a posteriori analysis** and in **adaptivity**, to define norms, the **iterative linearization** on the **discrete level**.

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Weak solution

Definition (Weak solution)

Find $u \in H_0^1(\Omega)$ s.t.

$$\langle \mathcal{R}(u), v \rangle = 0 \quad \forall v \in H_0^1(\Omega).$$

Finite element discretization

Definition (Finite element discretization)

Find $u_\ell \in V_\ell$ s.t.

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Find $u_\ell \in V_\ell$ s.t.

$$\langle \mathcal{R}(u_\ell), v_\ell \rangle = 0 \quad \forall v_\ell \in V_\ell.$$

- $\mathcal{T}_\ell$ simplicial mesh of Ω
- $p \geq 1$ polynomial degree
- $V_\ell := \mathcal{P}_p(\mathcal{T}_\ell) \cap H_0^1(\Omega)$
- conforming finite elements

Iterative linearization

Definition (Iterative linearization)

Find $u_\ell^k \in V_\ell$ s.t.

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

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- iterative linearization index $k \geq 1$

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- $u_\ell^0 \in V_\ell$ a given initial guess
- **iteration-dependent** reaction–diffusion **scalar product**

$$((w, v))_{u_\ell^{k-1}} := (L_\ell^{k-1} w, v) + (A_\ell^{k-1} \nabla w, \nabla v), \quad w, v \in H_0^1(\Omega)$$

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- $A_\ell^{k-1} : \Omega \rightarrow \mathbb{R}^{d \times d}$ **matrix**-valued function constructed from u_ℓ^{k-1} , $L_\ell^{k-1} : \Omega \rightarrow \mathbb{R}$ **scalar**-valued function constructed from u_ℓ^{k-1}

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- $L_\ell^{k-1} = 0$ if $f = f(\mathbf{x})$ (linear, source term)

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- examples for the gradient-dependent diffusivity:
 - $\mathbf{A}_\ell^{k-1} = a(|\nabla u_\ell^{k-1}|) \mathbf{I}_d$, $L_\ell^{k-1} = \partial_u f(u_\ell^{k-1})$ (Kačanov)
 - $\mathbf{A}_\ell^{k-1} = a(|\nabla u_\ell^{k-1}|) \mathbf{I}_d + \frac{a'(|\nabla u_\ell^{k-1}|)}{|\nabla u_\ell^{k-1}|} \nabla u_\ell^{k-1} \otimes \nabla u_\ell^{k-1}$, $L_\ell^{k-1} = \partial_u f(u_\ell^{k-1})$ (Newton)
 - $\mathbf{A}_\ell^{k-1} = \gamma \mathbf{I}_d$ with $\gamma \geq \frac{a_c^2}{a_m}$ a **constant parameter**, $L_\ell^{k-1} = 0$ (Zarantonello)

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- examples for the gradient-independent diffusivity:
 - $\mathbf{A}_\ell^{k-1} = \tau \mathbf{K} a(u_\ell^{k-1})$ (fixed point)

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- examples for the gradient-independent diffusivity:

- $\mathbf{A}_\ell^{k-1} = \tau \mathbf{K} a(u_\ell^{k-1})$ (fixed point)
- Picard: $L_\ell^{k-1} = \partial_u f(u_\ell^{k-1})$
- Jäger–Kačur: $L_\ell^{k-1} = \max_{u \in \mathbb{R}} \left(\frac{f(u) - f(u_\ell^{k-1})}{u - u_\ell^{k-1}} \right)$
- L -scheme: $L_\ell^{k-1} = \text{cnst} \geq \frac{1}{2} \sup \partial_u f$
- M -scheme: $L_\ell^{k-1} = \partial_u f(u_\ell^{k-1}) + \tau \times \text{cnst}$

Iterative linearization

Definition (Iterative linearization)

Find $u_\ell^k \in V_\ell$ s.t.

$$\langle \mathcal{R}_{\text{disc}}^{u_\ell^{k-1}}(u_\ell^k), v_\ell \rangle := ((u_\ell^k - u_\ell^{k-1}, v_\ell))_{u_\ell^{k-1}} + \langle \mathcal{R}(u_\ell^{k-1}), v_\ell \rangle = 0 \quad \forall v_\ell \in V_\ell.$$

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Iteration-dependent norm

Definition (Iteration-dependent norm)

$$|||v|||_{u_\ell^{k-1}}^2 := ((v, v))_{u_\ell^{k-1}} = \|(\mathbf{L}_\ell^{k-1})^{1/2} v\|^2 + \|(\mathbf{A}_\ell^{k-1})^{1/2} \nabla v\|^2, \quad v \in H_0^1(\Omega)$$

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- induced by the linearization scalar product

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

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

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- **orthogonal decomposition** into **error components**
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- $u_{\langle \ell \rangle}^k \in H_0^1(\Omega)$ such that $\langle \mathcal{R}_{\text{disc}}^{u_\ell^{k-1}}(u_{\langle \ell \rangle}^k), v \rangle = 0 \forall v \in H_0^1(\Omega)$
- **discretization error** is given by a **linear** reaction–diffusion problem

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A posteriori error estimates in an iteration-dependent norm

Theorem (A posteriori estimate in iteration-dependent norm)

For all linearization steps $k \geq 1$,

$$|||\mathcal{R}(u_\ell^{k-1})|||_{-1, u_\ell^{k-1}} \leq \eta(u_\ell^k).$$

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Moreover, for all linearization steps $k \geq 1$, there holds

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✓ $C_\ell^k = 1$ for Zarantonello $\implies$ **robustness** wrt the **strength of nonlinearities**

A posteriori error estimates in an iteration-dependent norm

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- ✓ $C_\ell^k = 1$ for Zarantonello $\implies$ **robustness** wrt the **strength of nonlinearities**
- ✓ C_K^k given by **local conditioning** of the linearization matrix $\mathbf{A}_\ell^{k-1}$ (and scalar L_ℓ^{k-1}):

A posteriori error estimates in an iteration-dependent norm

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- ✓ $C_\ell^k = 1$ for Zarantonello $\implies$ **robustness** wrt the **strength of nonlinearities**
- ✓ C_K^k given by **local conditioning** of the linearization matrix $\mathbf{A}_\ell^{k-1}$ (and scalar L_ℓ^{k-1}): typically **much better** than global conditioning (= worst-case scenario)

A posteriori error estimates in an iteration-dependent norm

Theorem (A posteriori estimate in iteration-dependent norm)

For all linearization steps $k \geq 1$,

$$|||\mathcal{R}(u_\ell^{k-1})|||_{-1, u_\ell^{k-1}} \leq \eta(u_\ell^k).$$

Moreover, for all linearization steps $k \geq 1$ and **for each element** $K \in \mathcal{T}_\ell$, there holds

$$\eta_K(u_\ell^k) \leq C_{\text{eff}}(d, \kappa_{\mathcal{T}}, p) C_K^k |||\mathcal{R}(u_\ell^{k-1})|||_{-1, u_\ell^{k-1}, \omega_K} + \text{quadrature error terms},$$

where

$$C_K^k := \left(\frac{\max. \text{ eig. } \mathbf{A}_\ell^{k-1}|_{\omega_K}}{\min. \text{ eig. } \mathbf{A}_\ell^{k-1}|_{\omega_K}} \right)^{1/2} + \left(\frac{\max. L_\ell^{k-1}|_{\omega_K}}{\min. L_\ell^{k-1}|_{\omega_K}} \right)^{1/2} \quad \text{if react. dom.}$$

- ✓ $C_\ell^k = 1$ for Zarantonello $\implies$ **robustness** wrt the **strength of nonlinearities**
- ✓ C_K^k given by **local conditioning** of the linearization matrix $\mathbf{A}_\ell^{k-1}$ (and scalar L_ℓ^{k-1}): typically **much better** than global conditioning (= worst-case scenario)
- ✓ C_K^k **computable**: we can affirm **robustness a posteriori**, for the given case

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- ✓ **local efficiency**

Outline

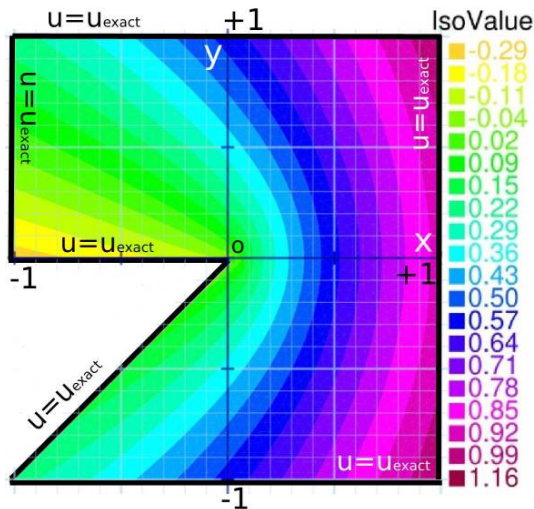
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Gradient-dependent diffusivity

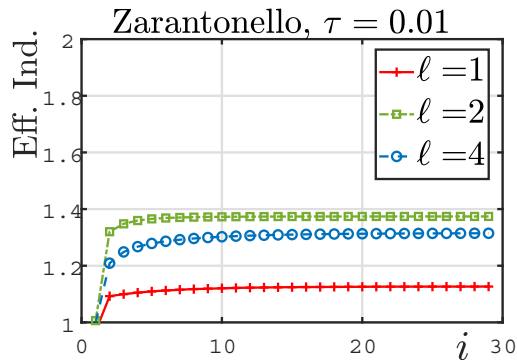
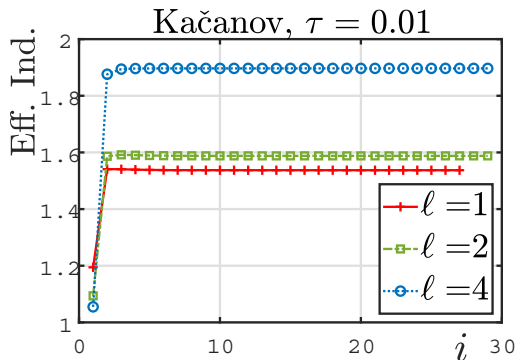
Setting

- mean curvature with $a(r) := a_m + \frac{a_c - a_m}{\sqrt{1+r^2}}$
- $a_c := a_m + 1$, $a_m := 2\tau$, $\frac{a_c}{a_m} = 1 + 0.5\frac{1}{\tau}$, $\frac{1}{\tau} \in [1, 10^3]$
- $f(\mathbf{x}, \xi) := \nu \xi - g(\mathbf{x})$, $\nu = 10^{-2}$
- $g(\mathbf{x}) := \left[r^{-1} \frac{(1-\lambda)(\lambda r^{\lambda-1})^3}{(1+(\lambda r^{\lambda-1})^2)^{\frac{3}{2}}} + \nu r^{\lambda} \right] \cos(\lambda\theta)$, $\lambda := 4/7$
- weak solution $u(\mathbf{x}) := r^{\lambda} \cos(\lambda\theta)$

Solution u

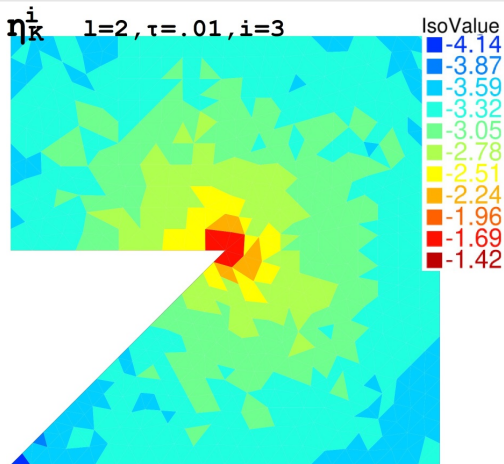


How large is the error?

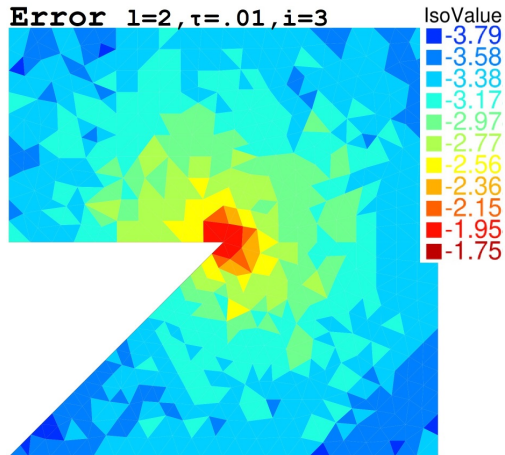


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Where is the error localized?

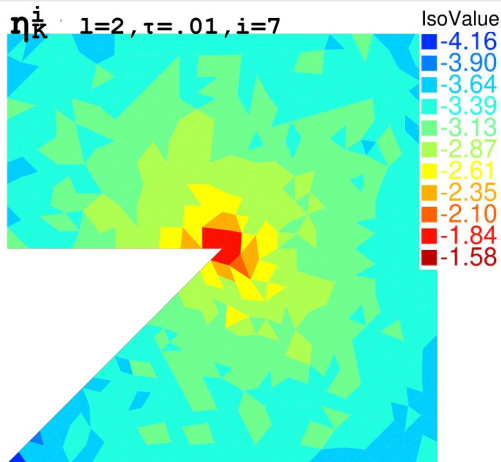


Estimated error, $\tau = 0.01$, Kačanov

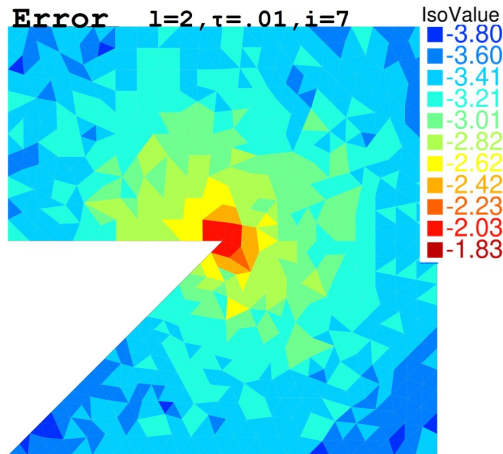


Exact error, $\tau = 1, \tau = 0.01$, Kačanov

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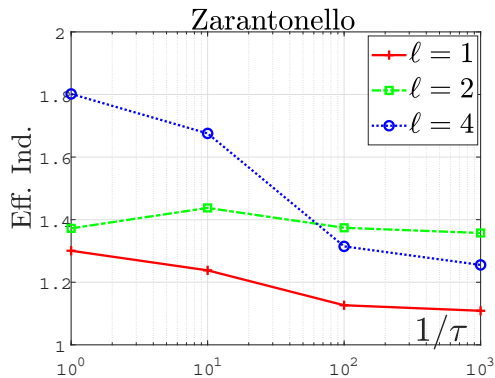
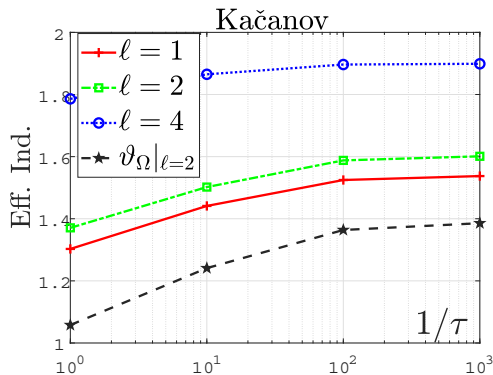


Estimated error, $\tau = 0.01$, Zarantonello



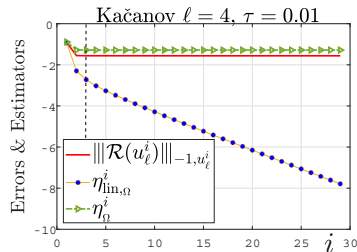
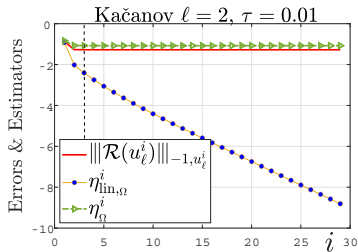
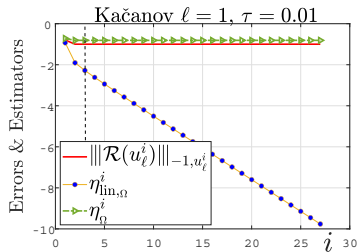
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Robustness wrt the nonlinearities



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Error components and adaptivity via stopping criteria



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Gradient-independent diffusivity

Setting

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

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

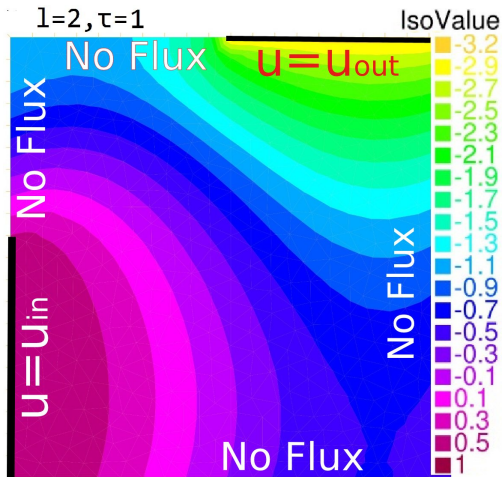
$$\mathbf{K} = \begin{bmatrix} 1 & 0.2 \\ 0.2 & 1 \end{bmatrix}, \quad \mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

- **van Genuchten saturation** and **permeability** laws

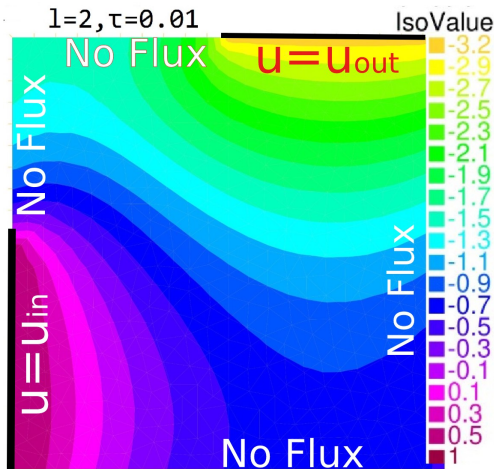
$$S(u) := \left(1 + (2 - u)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \quad \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^{\lambda}\right)^2, \quad \lambda = 0.5$$

- time step length $\tau \in [10^{-3}, 1]$

Solution u

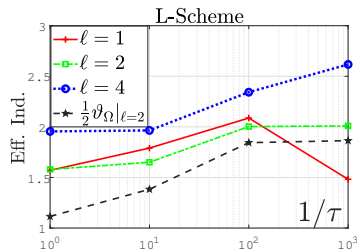
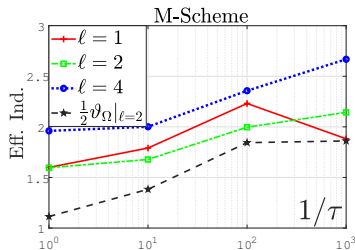
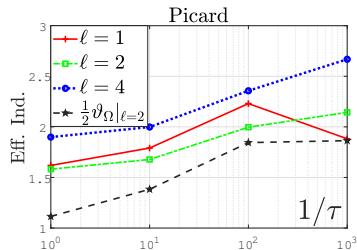


Time step length $\tau = 1$



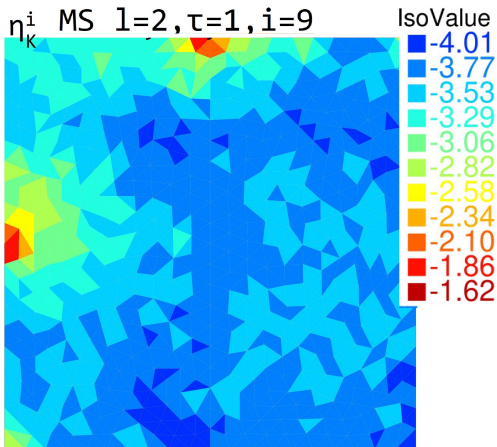
Time step length $\tau = 0.01$

How large is the error? Robustness wrt the nonlinearities

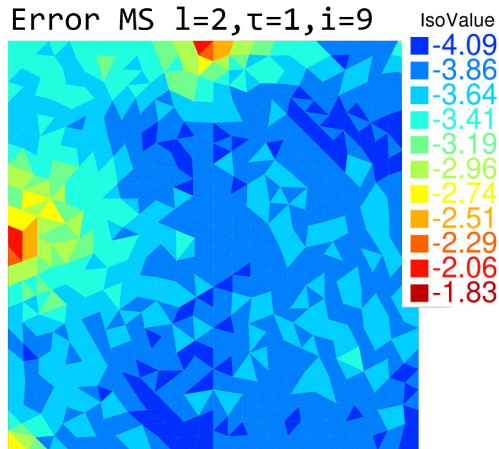


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Where is the error localized?

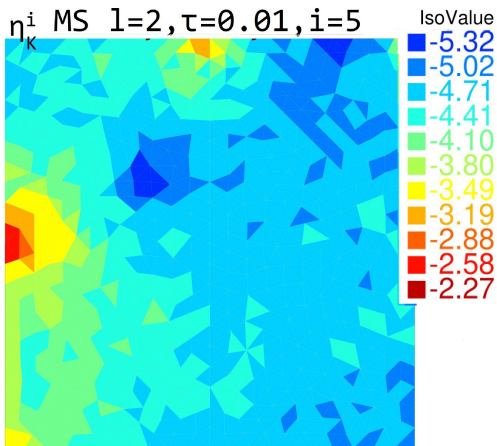


Estimated error, $\tau = 1$

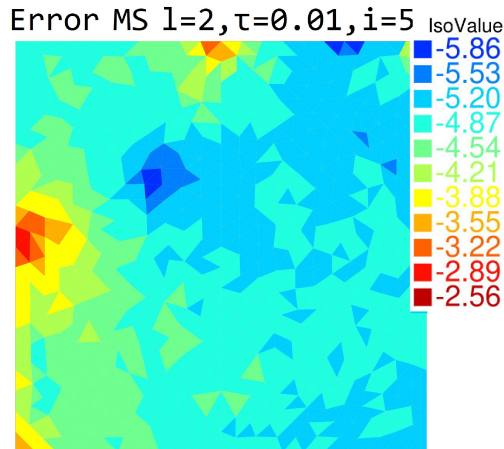


Exact error, $\tau = 1$

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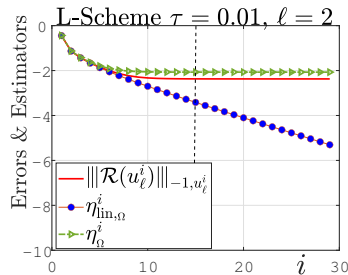
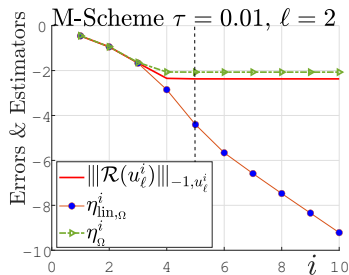
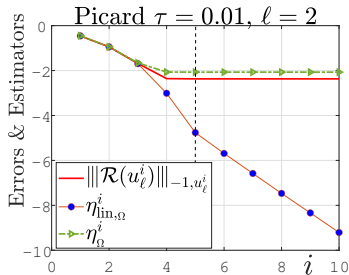


Estimated error, $\tau = 0.01$



Exact error, $\tau = 0.01$

Error components and adaptivity via stopping criteria



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A model nonlinear problem

Nonlinear elliptic problem

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

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

- $\Omega \subset \mathbb{R}^d$, $1 \leq d \leq 3$, open bounded polytope with Lipschitz boundary $\partial\Omega$
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Assumption (Gradient-dependent diffusivity)

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

$$\begin{aligned} |a(|\mathbf{x}|)\mathbf{x} - a(|\mathbf{y}|)\mathbf{y}| &\leq a_c |\mathbf{x} - \mathbf{y}| && (\text{Lipschitz continuity}), \\ (a(|\mathbf{x}|)\mathbf{x} - a(|\mathbf{y}|)\mathbf{y}) \cdot (\mathbf{x} - \mathbf{y}) &\geq a_m |\mathbf{x} - \mathbf{y}|^2 && (\text{strong monotonicity}). \end{aligned}$$

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

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Weak solution

Definition (Weak solution)

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

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

Energy and energy minimization

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 $r \in [0, \infty)$,

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

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Equivalently

$$u = \arg \min_{v \in H_0^1(\Omega)} \mathcal{J}(v)$$

Finite element approximation

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Energy difference

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

- $\mathcal{J}(u_\ell) - \mathcal{J}(u) \geq 0$, $\mathcal{J}(u_\ell) - \mathcal{J}(u) = 0$ if and only if $u_\ell = u$
- **physically-based** error measure

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Iterative linearization

Definition (Linearized finite element approximation)

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

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

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

Examples

Example (Picard (fixed-point))

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

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Example (Zarantonello)

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

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

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Example (Newton)

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

$$\mathbf{b}_\ell^{k-1} = a'(|\nabla u_\ell^{k-1}|) |\nabla u_\ell^{k-1}| \nabla u_\ell^{k-1}.$$

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Apply in the **a posteriori analysis** and in **adaptivity**, to define the way how we measure the error, the **iterative linearization** on the **discrete level**.

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Continuous minimizer of the linearized energy functional

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$$\mathcal{E}_\ell^k = \frac{1}{2} \text{energy difference} + \lambda_\ell^k \times \frac{1}{2} (\text{linearized energy difference})$$

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- $\lambda_\ell^k = \frac{\eta_{N,\ell}^k}{\eta_{L,\ell}^k + \widehat{\eta}_{\text{iter},\ell}^k}$ computable weight $\implies$ the two components comparable

Augmented energy difference

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- the associated a posteriori error **estimator** η_ℓ^k is **equivalent** to the usual energy difference estimator $\eta_{N,\ell}^k$ up to a factor $\frac{1}{2}$

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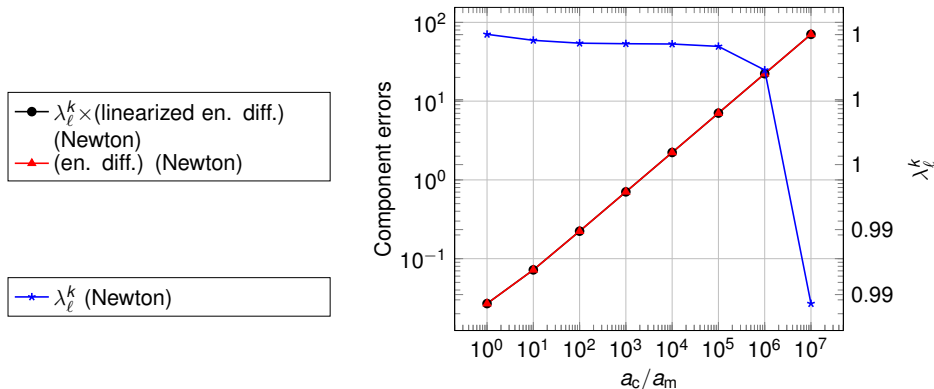
Practically

$$\mathcal{E}_\ell^k = \mathcal{J}(u_\ell^k) - \mathcal{J}(u) \text{ at convergence}$$

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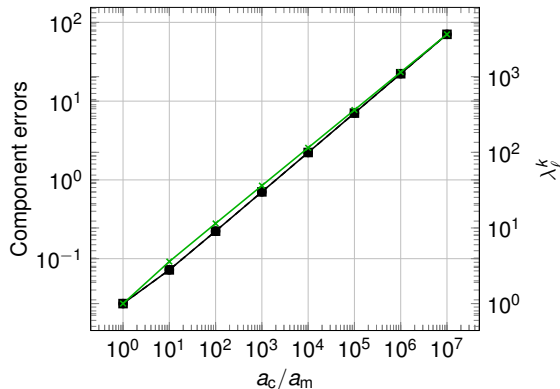
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—●— $\lambda_\ell^k \times (\text{linearized en. diff.})$
 (Zarantonello)
 -■- (en, dif.) (Zarantonello)

—x— λ_ℓ^k (Zarantonello)



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A posteriori error estimates for an augmented energy difference

Theorem (A posteriori estimate of augmented energy)

For all linearization steps $k \geq 1$,

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

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Fenchel conjugate, dual energy, estimator, flux equilibration

Definition (Fenchel conjugate)

$$\phi^*(\cdot, s) := \sup_{r \in [0, \infty)} (sr - \phi(\cdot, r)).$$

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$$\eta_{\ell}^k := \underbrace{\frac{1}{2}(\mathcal{J}(u_{\ell}^k) - \mathcal{J}^*(\sigma_{\ell}^k))}_{\text{en. diff. estimate}} + \lambda_{\ell}^k \underbrace{\frac{1}{2}(\mathcal{J}_{\ell}^{\text{red}, k-1}(u_{\ell}^k) - \mathcal{J}_{\ell}^{*, \text{red}, k-1}(\sigma_{\ell}^k))}_{\text{linearized en. diff. estimate}}$$

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Definition (Flux equilibration: $\sigma_{\ell}^k = \sum_{\mathbf{a} \in \mathcal{V}_{\ell}} \sigma_{\ell}^{\mathbf{a}, k}$)

$$\sigma_{\ell}^{\mathbf{a}, k} := \arg \min_{\mathbf{v}_{\ell} \in \mathcal{RT}_{p+1}(\mathcal{T}_{\mathbf{a}}) \cap \mathbf{H}_0(\text{div}, \omega_{\mathbf{a}})} \|(\mathbf{A}_{\ell}^{k-1})^{-\frac{1}{2}} (\psi^{\mathbf{a}} \Pi_{\ell, p-1}^{RTN}(\mathbf{A}_{\ell}^{k-1} \nabla u_{\ell}^k - \mathbf{b}_{\ell}^{k-1}) + \mathbf{v}_{\ell})\|_{\omega_{\mathbf{a}}}^2.$$

$$\nabla \cdot \mathbf{v}_{\ell} = \Pi_{\ell, p}(\psi^{\mathbf{a}} f - \nabla \psi^{\mathbf{a}} \cdot (\mathbf{A}_{\ell}^{k-1} \nabla u_{\ell}^k - \mathbf{b}_{\ell}^{k-1}))$$

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Smooth solution

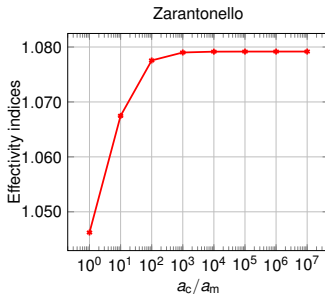
Setting

- unit square $\Omega = (0, 1)^2$
- known smooth solution $u(x, y) := 10 x(x - 1)y(y - 1)$
- $p = 1$
- effectivity indices

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

How large is the error? **Robustness** wrt the nonlinearities

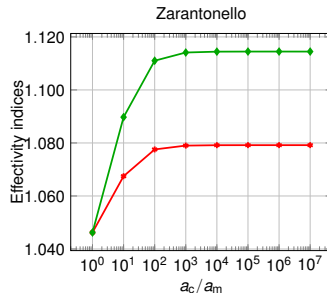
$$(a(r) = a_m + \frac{a_c - a_m}{\sqrt{1+r^2}})$$



$$I_\ell^k$$

How **large** is the error? **Robustness** wrt the nonlinearities

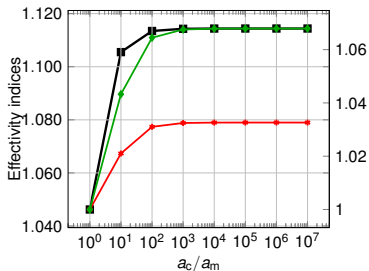
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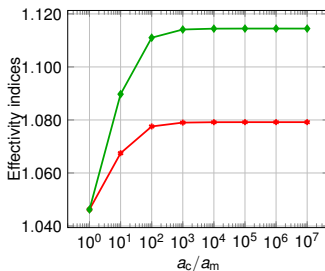
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$$(a(r) = a_m + \frac{a_c - a_m}{\sqrt{1+r^2}})$$

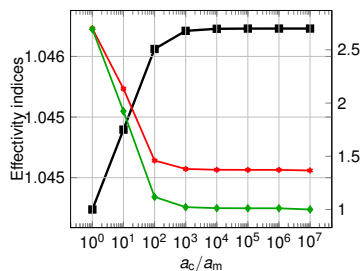
Picard



Zarantonello



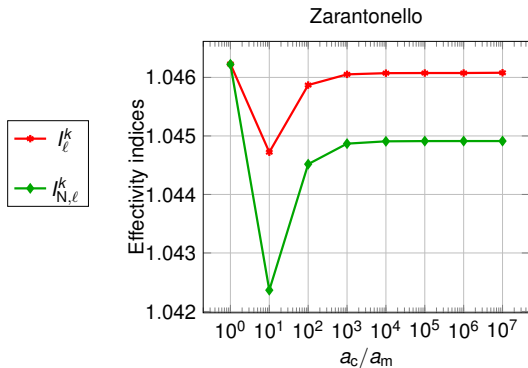
Newton



A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

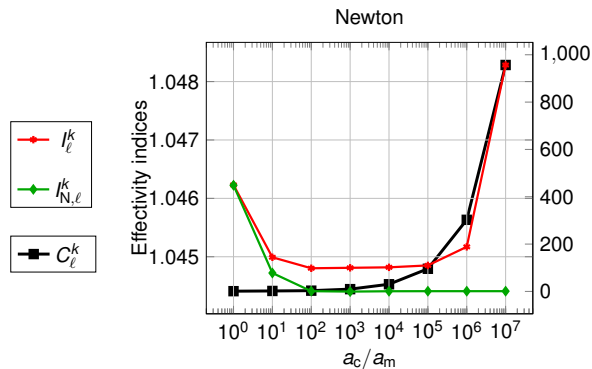
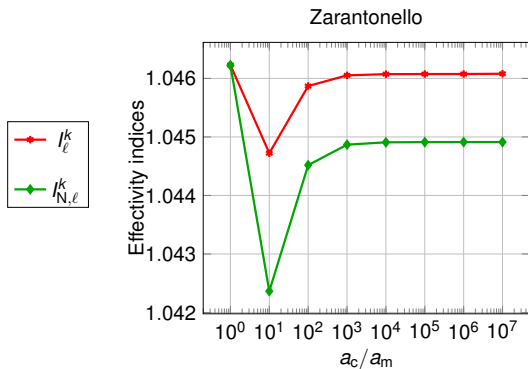
How **large** is the error? **Robustness** wrt the nonlinearities

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



How large is the error? Robustness wrt the nonlinearities

$$(a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}}, \text{ robustness only for Zarantonello})$$



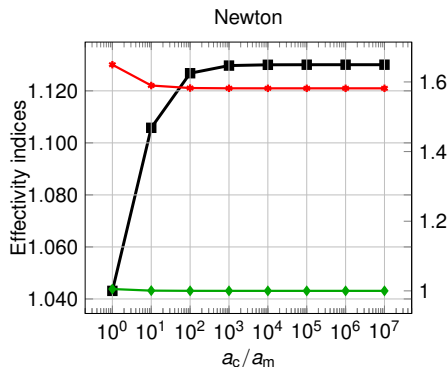
A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

Singular solution

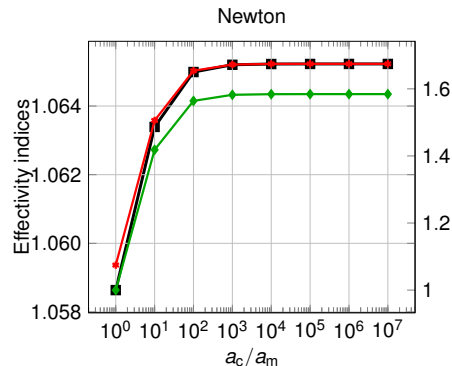
Setting

- L-shaped domain $\Omega = (-1, 1)^2 \setminus ([0, 1) \times (-1, 0])$
- known singular solution $u(\rho, \theta) = \rho^{\frac{2}{3}} \sin(\frac{2}{3}\theta)$
- $a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}}}$
- $p = 1$
- uniform or adaptive mesh refinement

How large is the error? Robustness wrt the nonlinearities



Uniform mesh refinement



Adaptive mesh refinement

A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

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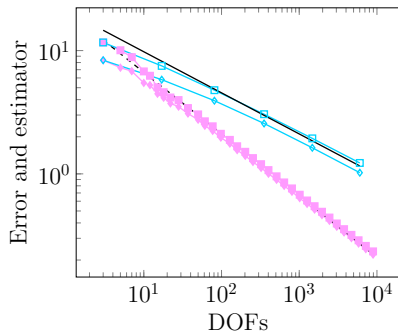
Thank you for your attention!

Outline

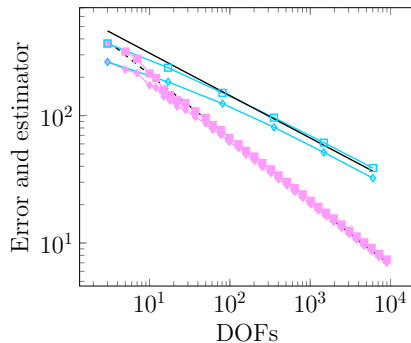
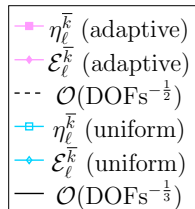
5 Adaptivity

6 Equilibrated flux reconstruction

Decreasing the error efficiently: optimal decay rate wrt DoFs



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



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

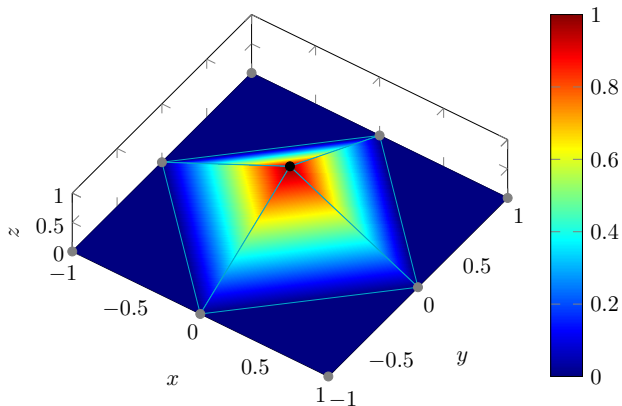
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Partition of unity

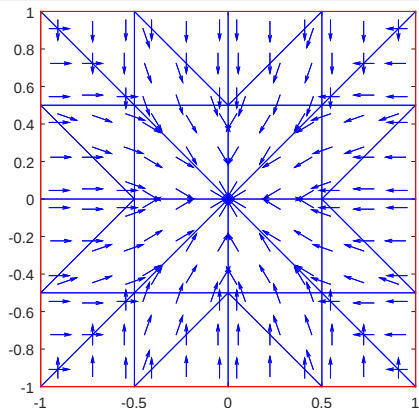
$$\sum_{\mathbf{a} \in \mathcal{V}_\ell} \psi^{\mathbf{a}} = 1$$



Hat basis function $\psi^{\mathbf{a}}$

Equilibrated flux reconstruction

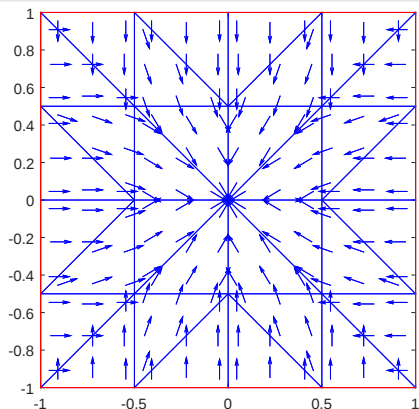
Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



Flux $\iota_\ell \notin \mathbf{H}(\text{div})$ (e.g. FE flux $-\nabla u_\ell$)

Equilibrated flux reconstruction

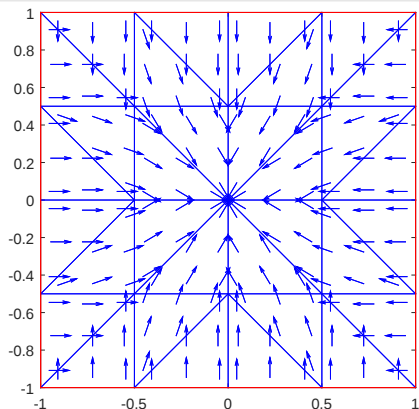
Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



Flux $\iota_\ell \notin H(\text{div}), \nabla \cdot \iota_\ell \neq f$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)

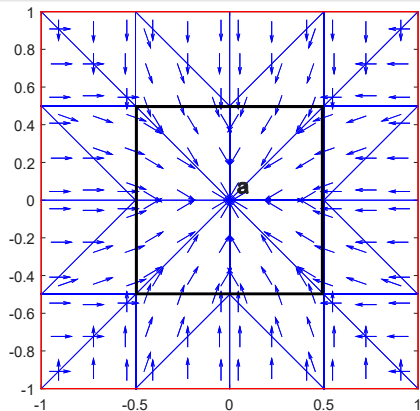


Flux $\boldsymbol{\iota}_\ell \notin \boldsymbol{H}(\text{div})$, $\nabla \cdot \boldsymbol{\iota}_\ell \neq f$

$$\underbrace{\boldsymbol{\iota}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}$$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)

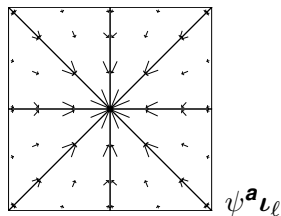
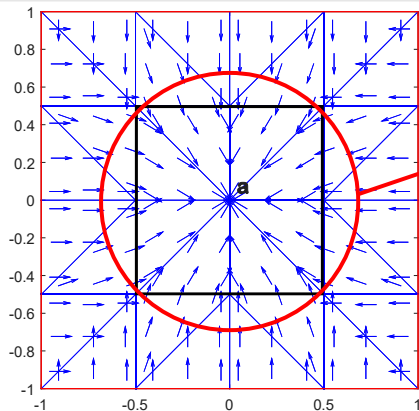


Flux $\boldsymbol{\iota}_\ell \notin \boldsymbol{H}(\text{div})$, $\nabla \cdot \boldsymbol{\iota}_\ell \neq f$

$$\underbrace{\boldsymbol{\iota}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}_{(f, \psi^a)_{\omega_a} + (\boldsymbol{\iota}_\ell, \nabla \psi^a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}}$$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)

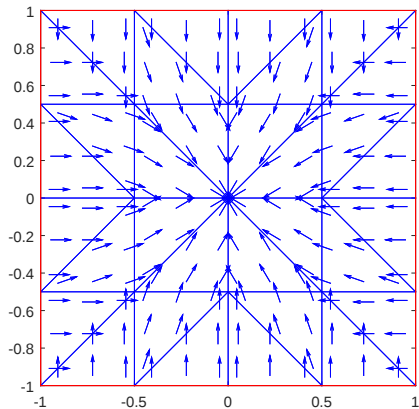


Flux $\boldsymbol{\iota}_\ell \notin \mathbf{H}(\text{div})$, $\nabla \cdot \boldsymbol{\iota}_\ell \neq f$

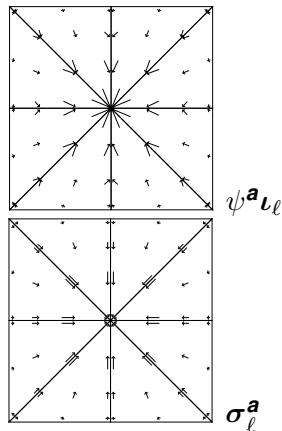
$$\underbrace{\boldsymbol{\iota}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}$$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



Flux $\boldsymbol{\iota}_\ell \notin \boldsymbol{H}(\text{div})$, $\nabla \cdot \boldsymbol{\iota}_\ell \neq f$

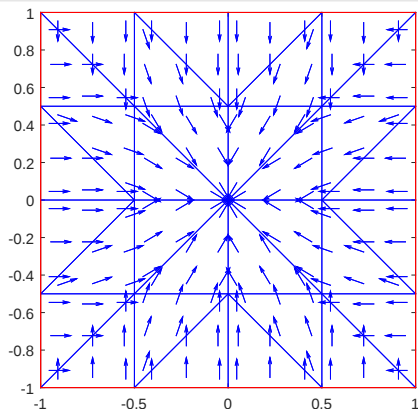


$$\underbrace{\boldsymbol{\iota}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}$$

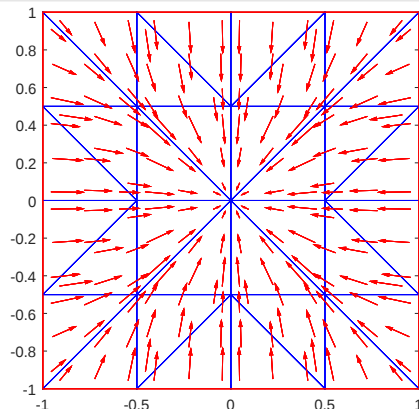
$$\sigma_\ell^{\boldsymbol{a}} := \arg \min_{\substack{\boldsymbol{v}_\ell \in \mathcal{RT}_{p+1}(\mathcal{T}_a) \cap \boldsymbol{H}_0(\text{div}, \omega_a) \\ \nabla \cdot \boldsymbol{v}_\ell = f \psi^{\boldsymbol{a}} + \boldsymbol{\iota}_\ell \cdot \nabla \psi^{\boldsymbol{a}}}} \|\psi^{\boldsymbol{a}} \boldsymbol{\iota}_\ell - \boldsymbol{v}_\ell\|_{\omega_a}^2$$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



Flux $\iota_\ell \notin \mathbf{H}(\text{div})$, $\nabla \cdot \iota_\ell \neq f$

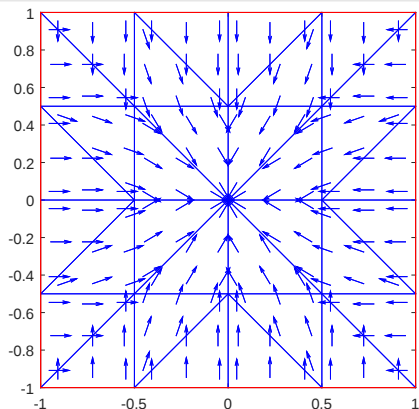


Equilibrated flux $\sigma_\ell \in \mathbf{H}(\text{div})$, $\nabla \cdot \sigma_\ell = f$

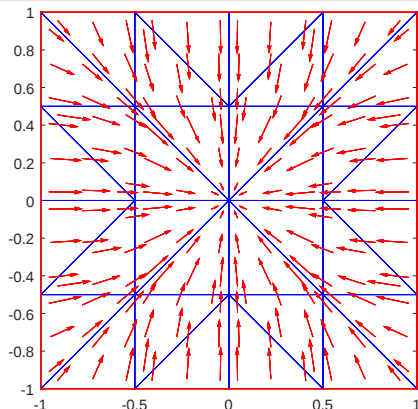
$$\underbrace{\iota_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)} \rightarrow \sigma_\ell := \sum_{a \in \mathcal{V}_\ell} \sigma_\ell^a \in \mathcal{RT}_{p+1}(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}), \nabla \cdot \sigma_\ell = f$$

Equilibrated flux reconstruction

Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



Flux $\iota_\ell \notin \mathbf{H}(\text{div})$, $\nabla \cdot \iota_\ell \neq f$



Equilibrated flux $\sigma_\ell \in \mathbf{H}(\text{div})$, $\nabla \cdot \sigma_\ell = f$