

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

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Inria Paris & Ecole des Ponts

Lisbon, September 5, 2023

Inria



Outline

- 1 Introduction
- 2 Gradient-dependent nonlinearities
 - Setting
 - Previous results
 - Iterative linearization
 - A posteriori error estimates for an augmented energy difference
 - Fenchel conjugate, dual energy, flux equilibration, estimator
 - Numerical experiments
- 3 Gradient-independent nonlinearities
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 - A posteriori error estimates for an iteration-dependent norm
 - Numerical experiments
- 4 Conclusions

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

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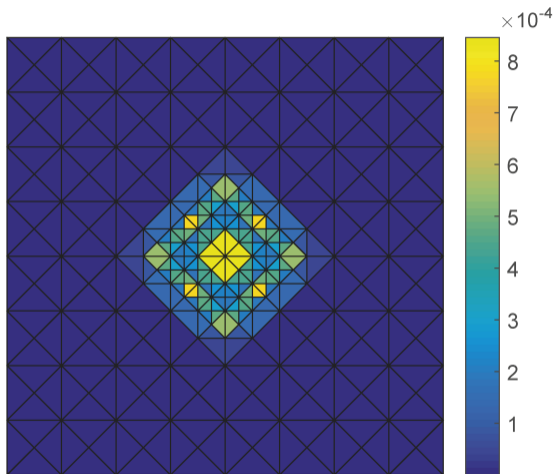
How large is the error? (steady linear Darcy, known solution)

$h \approx 1/ \mathcal{T}_\ell ^{1/2}$	p	relative error estimate $\frac{\eta(u_\ell)}{\ \nabla u_\ell\ }$	relative error $\frac{\ \nabla(u-u_\ell)\ }{\ \nabla u_\ell\ }$	effectivity index $\frac{\eta(u_\ell)}{\ \nabla(u-u_\ell)\ }$
h_0	1	28%	24%	1.17
$\approx h_0/2$		14%	13%	1.09
$\approx h_0/4$		7.0%	6.6%	1.06
$\approx h_0/8$		3.3%	3.1%	1.04
$\approx h_0/2$	2	$9.5 \times 10^{-1}\%$	$9.2 \times 10^{-1}\%$	1.04
$\approx h_0/4$	3	$5.9 \times 10^{-3}\%$	$5.9 \times 10^{-3}\%$	1.01
$\approx h_0/8$	4	$5.9 \times 10^{-6}\%$	$5.8 \times 10^{-6}\%$	1.01

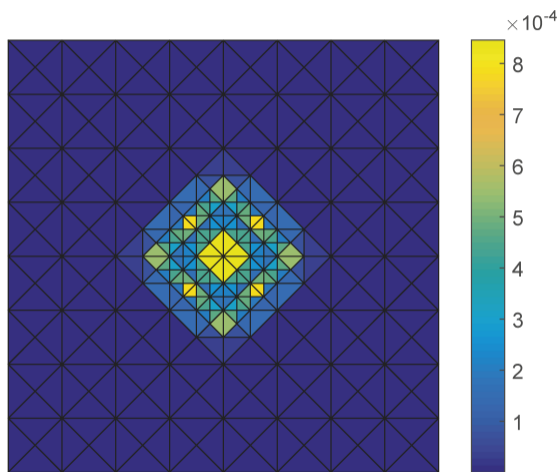
A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

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

Where (in space) is the error **localized**? (steady linear Darcy)



Estimated local error $\eta_K(u_\ell) = \|\nabla u_\ell + \sigma_\ell\|_K$



Exact local error $\|\nabla(u - u_\ell)\|_K$

Goals

Error control

a posteriori error estimates

$$\|u - u_\ell\| \leq \eta(u_\ell)$$

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efficient

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

Goals

Error control

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

$$\| \| u - u_\ell \| \| \leq \eta(u_\ell) \leq C_{\text{eff}} \| \| u - u_\ell \| \|, \quad C_{\text{eff}} \text{ independent of nonlinearities}$$

Goals

Error control

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

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

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

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

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Usage

- provide sharp **computable bounds** in **physically-based error measures**

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Usage

- provide sharp **computable bounds** in **physically-based error measures**
- predict the **error localization** (in discretization, in linearization, in linear solver)
- **adapt** the **nonlinear solver** and the **linear solver** (balancing error components), **adaptive mesh refinement** . . .

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A model steady 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 (Nonlinear function a)

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

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

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

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

Example of the nonlinear function a

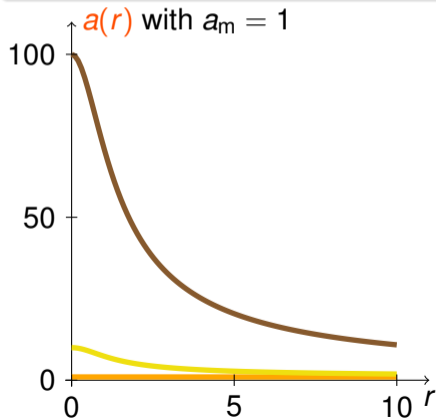
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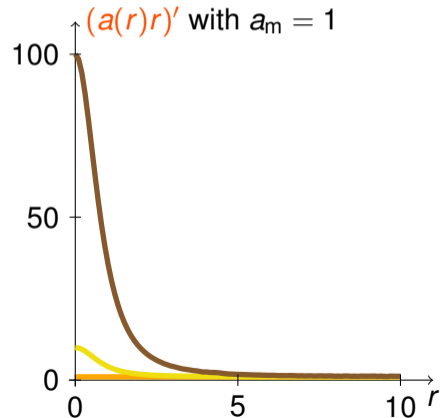
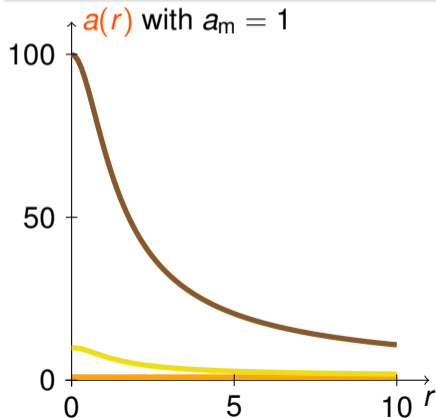


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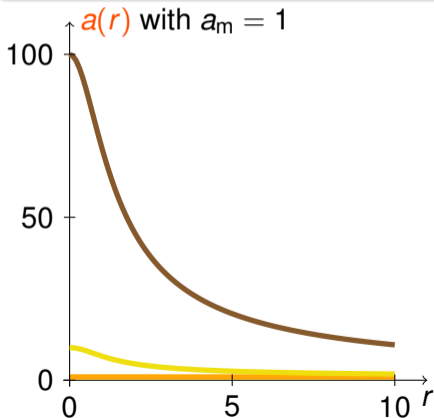


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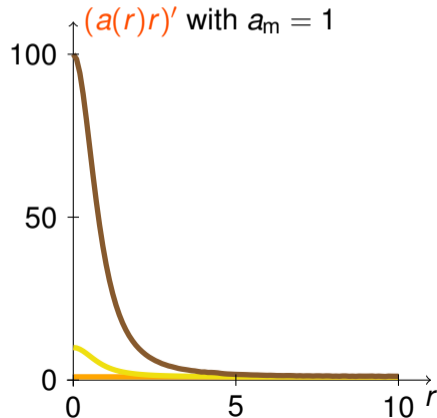
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Strength of the nonlinearity

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



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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

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

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

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Finite element approximation

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$u_\ell \in V_\ell^p$ 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
- $V_\ell^p := \mathcal{P}_p(\mathcal{T}_\ell) \cap H_0^1(\Omega)$
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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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Known results

Energy difference (not robust wrt $\frac{a_c}{a_m}$)

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

- Zeidler (1992), Han (1994), Repin (1997), Ladevèze & Moës (1997), Diening & Kreuzer (2008), Bartels & Milicevic (2020), ...

Sobolev norm

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

Dual norm of the residual

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

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• From: Hesthaven & Zang (2004), Voth & Zang (2005), Kim (2007), Houston, Hu, & Wathen (2007), Gopal, Mishra, & Zang (2011), Gopal, Haber, Prasad, & Wathen (2018), Hou & Wathen (2020).

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Dual norm of the residual (robust wrt $\frac{a_c}{a_m}$), "bypasses" the nonlinearity

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

Known results

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Iterative linearization

Need to **solve a nonlinear system**

$$\mathcal{A}_\ell(\mathbf{U}_\ell) = \mathbf{F}_\ell$$

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Definition (Linearized finite element approximation)

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

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

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

Examples

Example (Picard (fixed-point))

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

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

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Observation

None of the known approaches employs **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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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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Equivalently

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

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A posteriori error estimates for an augmented energy difference

Theorem (A posteriori estimate of augmented energy)

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

A posteriori error estimates for an augmented energy difference

Augmented energy difference

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

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Practically

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

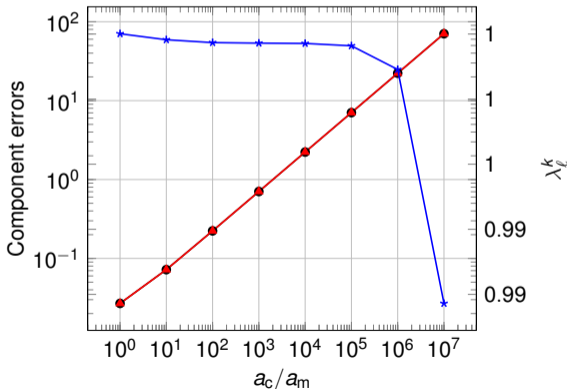
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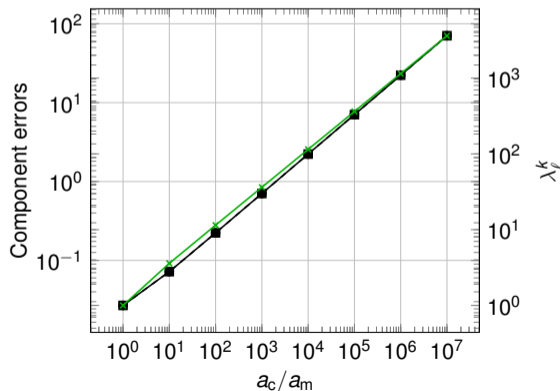
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Fenchel conjugate, dual energy, flux equilibration, estimator

Definition (Fenchel conjugate)

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

Definition (Dual energy)

$$\mathcal{J}^*(\mathbf{v}) := - \int_{\Omega} \phi^*(\cdot, |\mathbf{v}|), \quad \mathbf{v} \in \mathbf{H}(\text{div}, \Omega).$$

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

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

Definition (Estimator)

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

Fenchel conjugate, dual energy, flux equilibration, estimator

Definition (Fenchel conjugate)

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

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Smooth solution

Setting

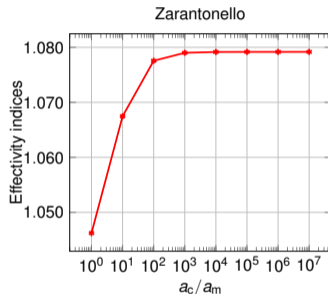
- unit square $\Omega = (0, 1)^2$
- known smooth solution $u(x, y) := 10 x(x - 1)y(y - 1)$
- $p = 1$
- effectivity indices

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

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

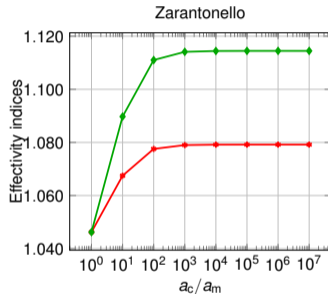
$$(a(r) = a_m + \frac{a_c - a_m}{\sqrt{1+r^2}})$$

J_ℓ^k



How large is the error? Robustness wrt the nonlinearities

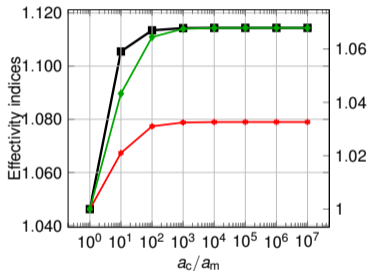
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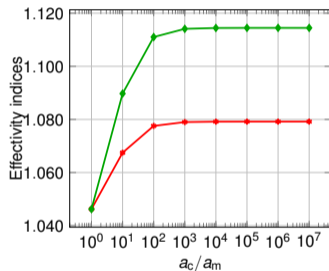
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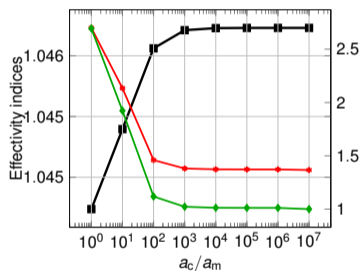
Picard



Zarantonello



Newton

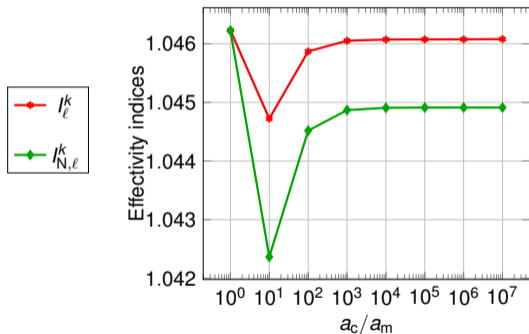


A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

How large is the error? Robustness wrt the nonlinearities

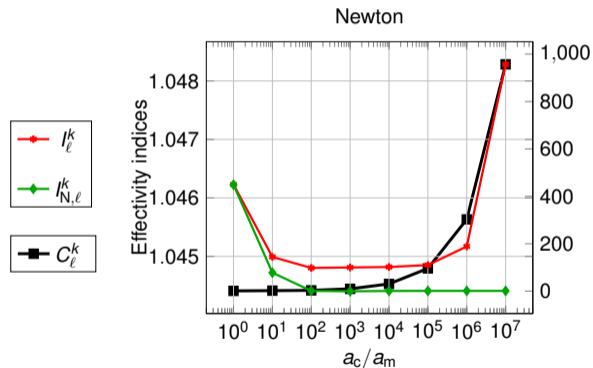
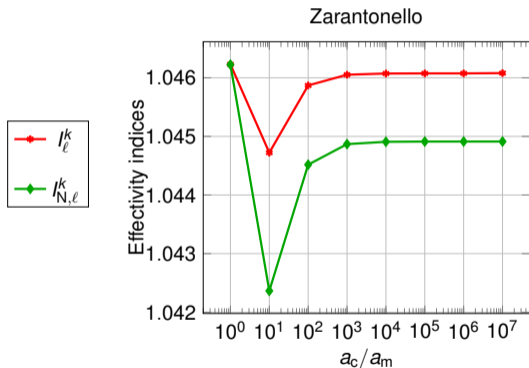
$$(a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}})$$

Zarantonello



How large is the error? Robustness wrt the nonlinearities

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



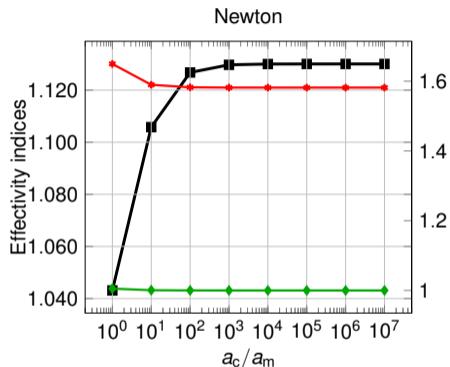
A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

Singular solution

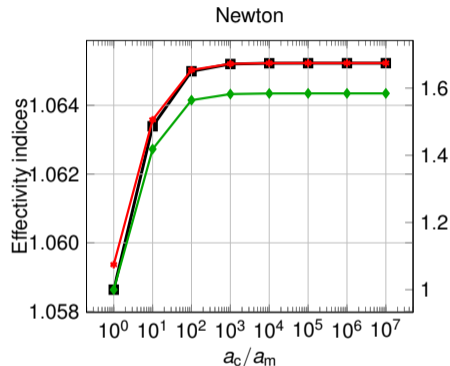
Setting

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

Observation

Not all nonlinear problems admit an energy minimization structure.

A model steady nonlinear problem

Nonlinear elliptic problem

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

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

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

- $\tau > 0$ a parameter (time step size in transient problems: applies to Richards on each time step)

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

$$|\mathcal{D}(\mathbf{x}_1, u_1) - \mathcal{D}(\mathbf{x}_2, u_2)| \leq \mathcal{D}_M (|\mathbf{x}_1 - \mathbf{x}_2| + |u_1 - u_2|) \quad \forall \mathbf{x}_1, \mathbf{x}_2 \in \Omega \text{ and } u_1, u_2 \in \mathbb{R},$$

$$0 \leq f(\mathbf{x}, u_2) - f(\mathbf{x}, u_1) \leq f_M (u_2 - u_1) \quad \forall \mathbf{x} \in \Omega \text{ and } u_1, u_2 \in \mathbb{R}, u_2 \geq u_1,$$

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

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Finite element discretization and iterative linearization

Definition (Linearized finite element approximation)

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

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

- covers most linearization schemes: Picard (fixed-point), L & M-schemes, ...
- linearization: reaction-diffusion scalar product

$$((w, v))_{u_\ell^{k-1}} = \underbrace{(\underbrace{L_\ell^{k-1}}_{\text{reaction coef.}} w, v)}_{\text{reaction coef.}} + \underbrace{(\underbrace{A_\ell^{k-1}}_{\text{diffusion coef.}} \nabla w, \nabla v)}_{\text{diffusion coef.}}, \quad w, v \in H_0^1(\Omega)$$

Iteration-dependent norm

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

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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

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

- orthogonal decomposition
- error components
- $u_{\langle \ell \rangle}^k \in H_0^1(\Omega)$ such that

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

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One time step of the Richards equation

Setting

- unit square $\Omega = (0, 1)^2$
- realistic data

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

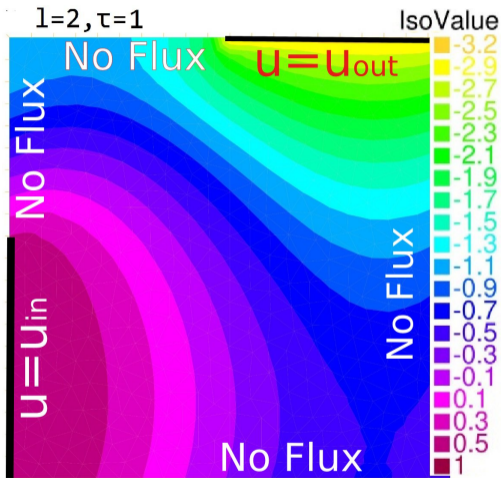
$$\mathbf{K} = \begin{bmatrix} 1 & 0.2 \\ 0.2 & 1 \end{bmatrix}, \quad \mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

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

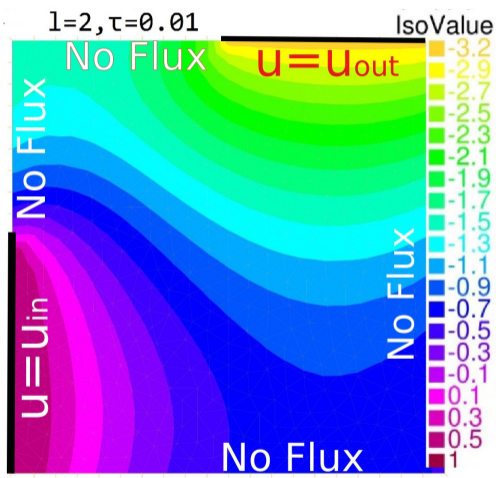
$$S(u) := \left(1 + (2 - u)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \quad \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^{\lambda}\right)^2, \quad \lambda = 0.5$$

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

One time step of the Richards equation: saturation u

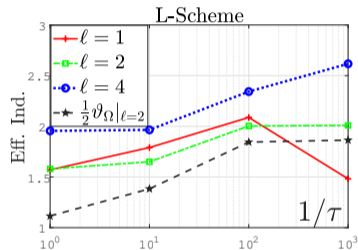
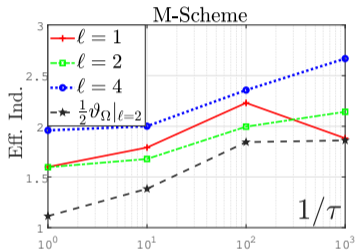
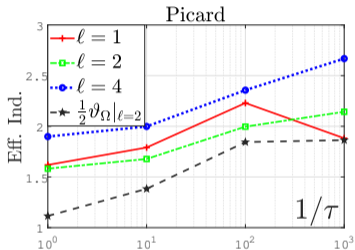


Time step length $\tau = 1$



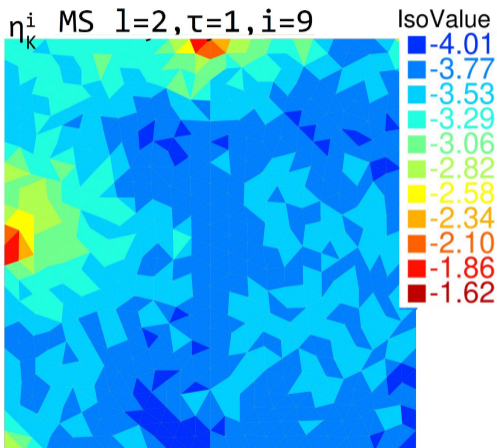
Time step length $\tau = 0.01$

How large is the error? Robustness wrt the nonlinearities

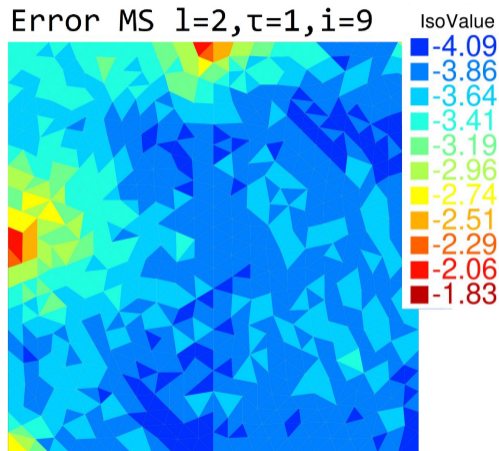


K. Mitra, M. Vohralik, to be submitted (2023)

Where is the error localized?

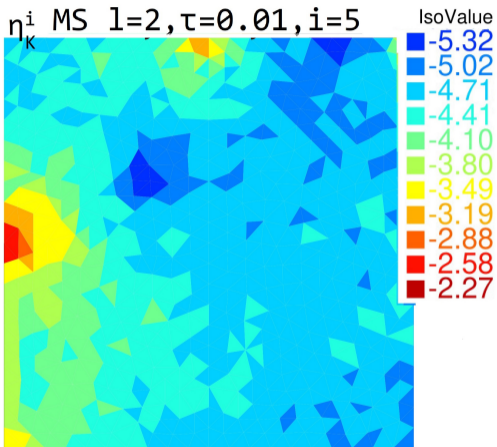


Estimated local error, $\tau = 1$

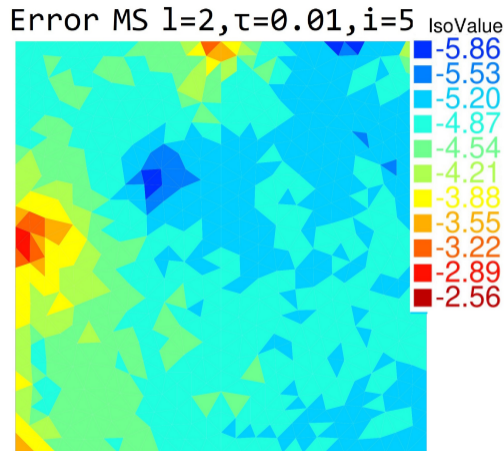


Exact local error, $\tau = 1$

Where is the error localized?

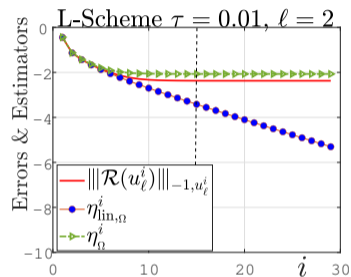
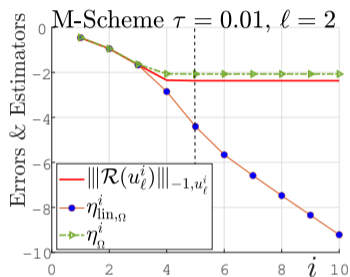
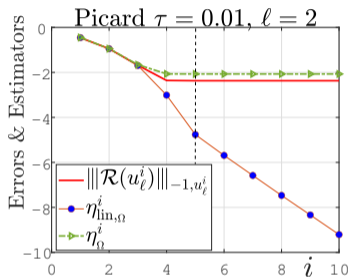


Estimated local error, $\tau = 0.01$



Exact local error, $\tau = 0.01$

Error components and adaptivity via stopping criteria



Time step length $\tau = 0.01$

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- **robustness** with respect to the **strength of nonlinearities** for model cases
- **augmenting** the **energy difference** by the (discretization) error on the given linearization step
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- 📄 HARNIST A., MITRA K., RAPPAPORT A., VOHRALÍK M. Robust energy a posteriori estimates for nonlinear elliptic problems. HAL Preprint 04033438, 2023.
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Thank you for your attention!

Outline

- 5 Other error measures
- 6 Adaptivity
- 7 Equilibrated flux reconstruction

Sobolev space and error

Sobolev space

$$H_0^1(\Omega)$$

Sobolev norm error

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

Residual and its dual norm

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

$$\langle \mathcal{R}(w), v \rangle := (a(|\nabla w|)\nabla w, \nabla v) - (f, v), \quad v \in H_0^1(\Omega).$$

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 $H_0^1(\Omega)$
- the most straightforward choice: $\|v\| := \|\nabla v\|$
- **mathematically-based** error measure

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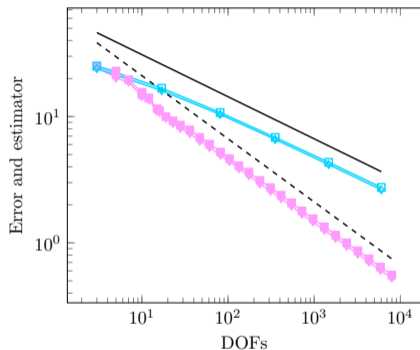
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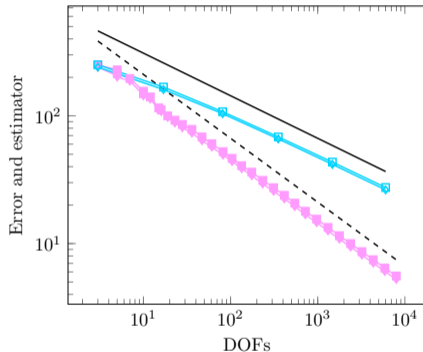
Outline

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- 6 Adaptivity**
- 7 Equilibrated flux reconstruction

Decreasing the error efficiently: optimal decay rate wrt DoFs



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



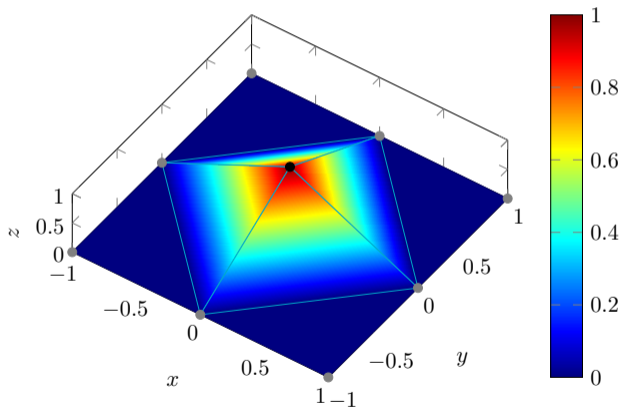
$$\frac{a_c}{a_m} = 10^6$$

Outline

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Partition of unity

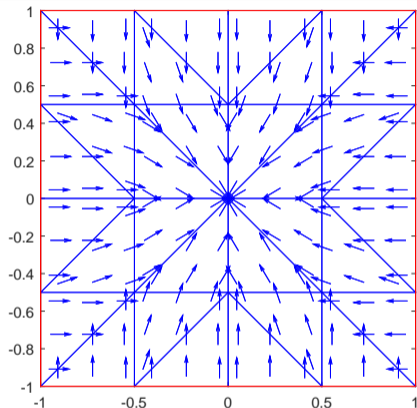
$$\sum_{\mathbf{a} \in \mathcal{V}_\ell} \psi^{\mathbf{a}} = 1$$



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

Equilibrated flux reconstruction

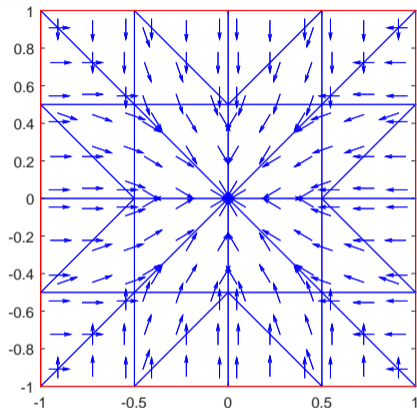
Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



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

Equilibrated flux reconstruction

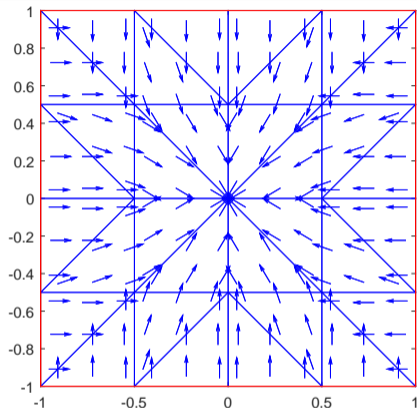
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Flux $\boldsymbol{u}_\ell \notin \boldsymbol{H}(\text{div}), \nabla \cdot \boldsymbol{u}_\ell \neq f$

Equilibrated flux reconstruction

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

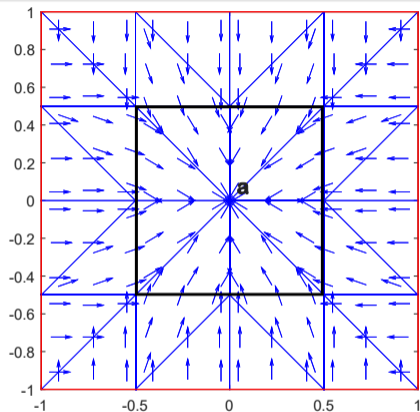


Flux $\boldsymbol{v}_\ell \notin \boldsymbol{H}(\text{div})$, $\nabla \cdot \boldsymbol{v}_\ell \neq f$

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

Equilibrated flux reconstruction

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



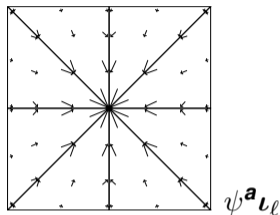
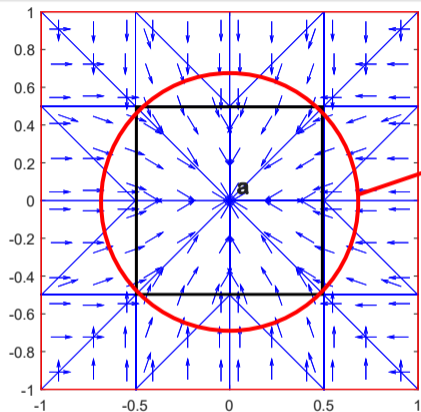
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$$(f, \psi^{\boldsymbol{a}})_{\omega_{\boldsymbol{a}}} + (\boldsymbol{u}_\ell, \nabla \psi^{\boldsymbol{a}})_{\omega_{\boldsymbol{a}}} = 0 \quad \forall \boldsymbol{a} \in \mathcal{V}_\ell^{\text{int}}$$

Equilibrated flux reconstruction

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

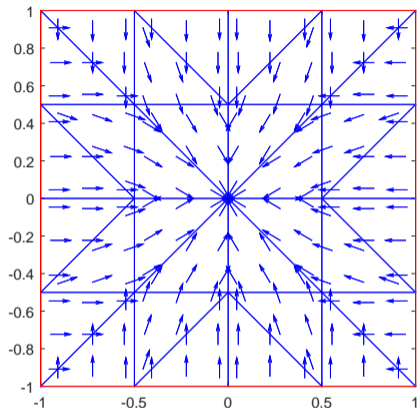


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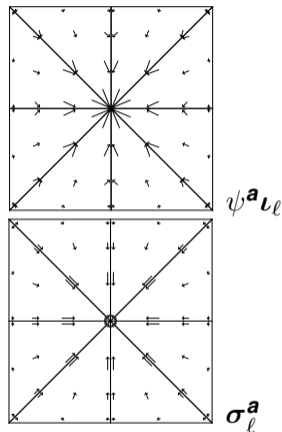
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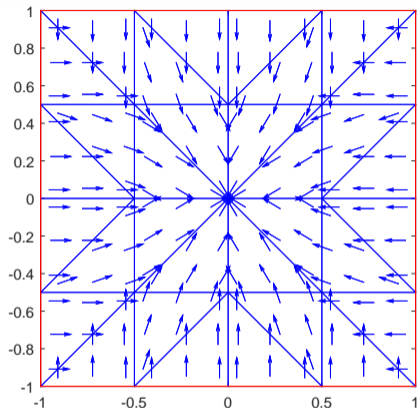


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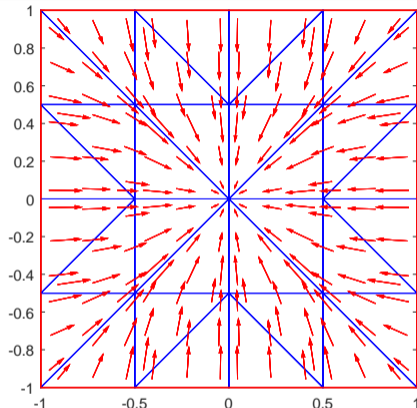
$$\sigma_\ell^a := \arg \min_{\substack{\boldsymbol{v}_\ell \in \mathcal{RT}_{p+1}(\mathcal{T}_a) \cap \boldsymbol{H}_0(\text{div}, \omega_a) \\ \nabla \cdot \boldsymbol{v}_\ell = f \psi^a + \boldsymbol{v}_\ell \cdot \nabla \psi^a}} \|\psi^a \boldsymbol{v}_\ell - \boldsymbol{v}_\ell\|_{\omega_a}^2$$

Equilibrated flux reconstruction

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



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

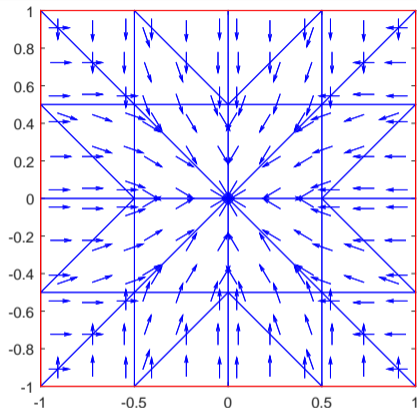


Equilibrated flux $\sigma_\ell \in \mathbf{H}(\text{div})$, $\nabla \cdot \sigma_\ell = f$

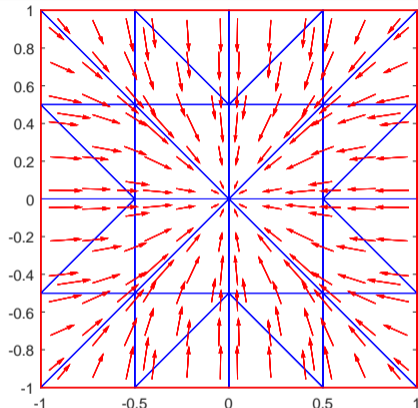
$$\underbrace{\iota_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)} \rightarrow \sigma_\ell := \sum_{a \in \mathcal{V}_\ell} \sigma_\ell^a \in \mathcal{RT}_{p+1}(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}), \nabla \cdot \sigma_\ell = f$$

Equilibrated flux reconstruction

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



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

Equilibrated flux reconstruction

Use

- **a posteriori error estimates**

- comparison of the original & reconstructed flux $\|\nabla u_\ell + \sigma_\ell\|$: discretization error
- error component fluxes: linearization and algebraic errors

- recovery of **mass conservative fluxes**

- local on patches of mesh elements from FE-type approximations
- local on elements from FV- & DG-type approximations
- inexact nonlinear solvers (still local)
- inexact linear solvers (price of one MG iteration)

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 - comparison of the original & reconstructed flux $\|\nabla u_\ell + \sigma_\ell\|$: **discretization error**
 - error component fluxes: **linearization** and **algebraic errors**
- recovery of **mass conservative fluxes**
 - local on patches of mesh elements from FE-type approximations
 - local on elements from FV- & DG-type approximations
 - **inexact nonlinear solvers** (still local)
 - **inexact linear solvers** (price of one MG iteration)