

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

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Outline

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

Setting

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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- a strongly monotone and Lipschitz continuous
- f piecewise polynomial for simplicity

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- numerical approximation u_ℓ

Goals

Error control

a posteriori error estimates

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

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

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

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Main question

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

Previous results

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

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

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

Energy difference

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

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

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Previous use of augmented error norms

- **advection-dominated problems**: augmenting the energy norm by the dual norm of the skew-symmetric part: robustness wrt advection (Verfürth 2005)
- **parabolic pbs**: augmenting by temporal jumps: space-time local efficiency

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Dual norm of the residual in an iteration-dependent energy norm

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

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

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- dual norm of the residual

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- iteration-dependent norm** induced by the **linearization scalar product**

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

New result #2b

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

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

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

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

$$(\|u_{\langle \ell \rangle}^k - u_\ell^{k-1}, v\|)_{u_\ell^{k-1}} = - \underbrace{\langle \mathcal{R}(u_\ell^{k-1}), v \rangle}_{\text{residual}} \quad \forall v \in H_0^1(\Omega)$$

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Assumption (Nonlinear function a)

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

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

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

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 $\partial\Omega$
- f piecewise polynomial for simplicity

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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

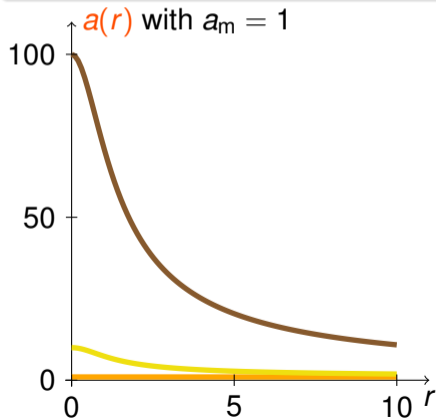
Example (Mean curvature nonlinearity)

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

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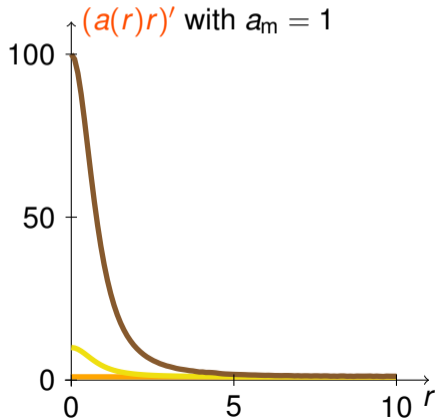
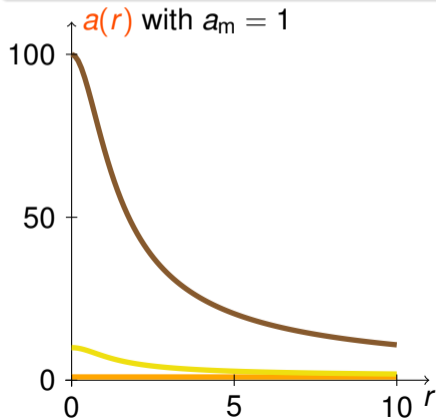


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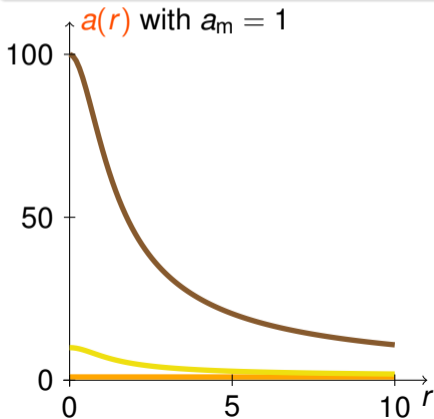


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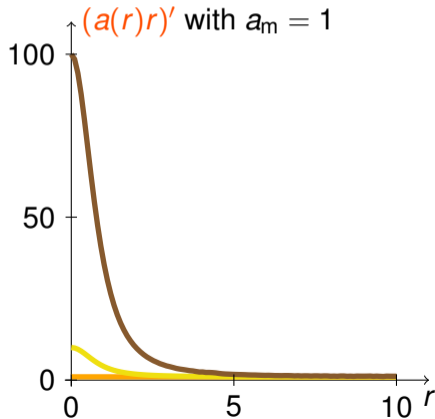
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Strength of the nonlinearity

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



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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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$u_\ell \in V_\ell^p$ 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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Energy difference

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 $u_\ell = u$
- **physically-based** error measure

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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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- $u_\ell^0 \in V_\ell^p$ 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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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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- ✓ 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

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

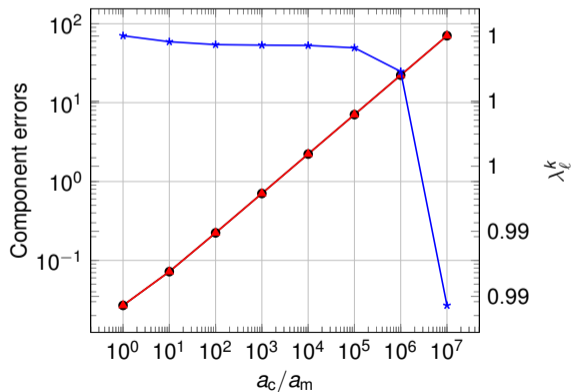
A posteriori error estimates for an augmented energy difference

Augmented energy difference

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

● $\lambda_\ell^k \times (\text{linearized en. diff.})$
 (Newton)
▲ (en. diff.) (Newton)

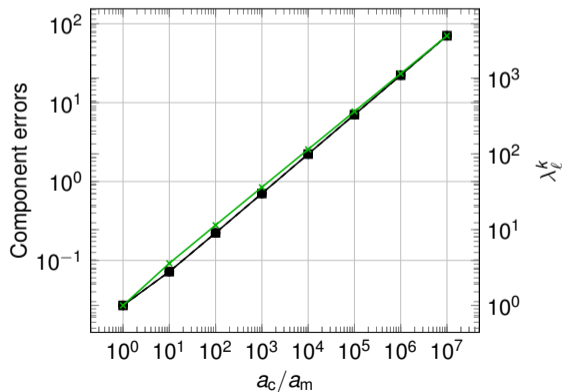
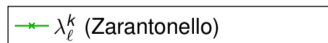
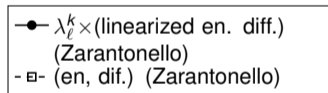
★ λ_ℓ^k (Newton)



A posteriori error estimates for an augmented energy difference

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- 2 Gradient-dependent nonlinearities
 - Setting
 - 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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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)

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



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Smooth solution

Setting

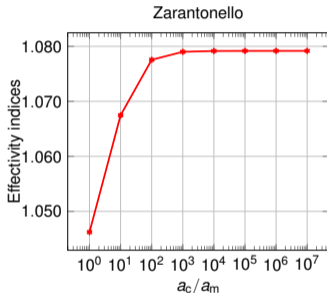
- unit square $\Omega = (0, 1)^2$
- known smooth solution $u(x, y) := 10 x(x - 1)y(y - 1)$
- $p = 1$
- effectivity indices

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

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

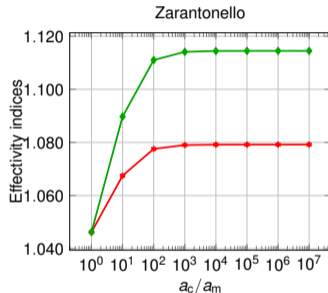
$$(a(r) = a_m + \frac{a_c - a_m}{\sqrt{1+r^2}})$$

J_ℓ^k



How large is the error? Robustness wrt the nonlinearities

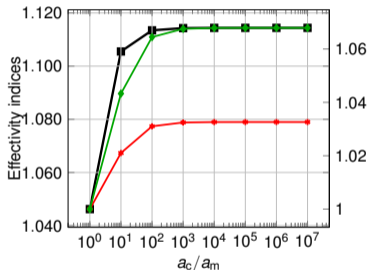
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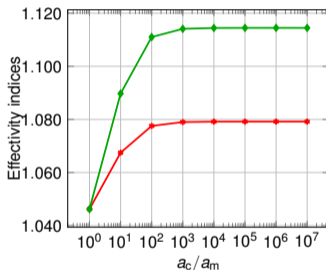
How large is the error? Robustness wrt the nonlinearities

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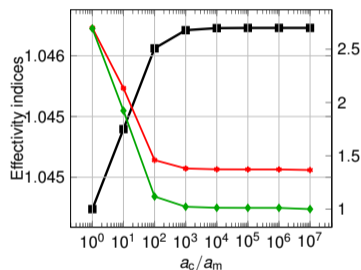
Picard



Zarantonello



Newton

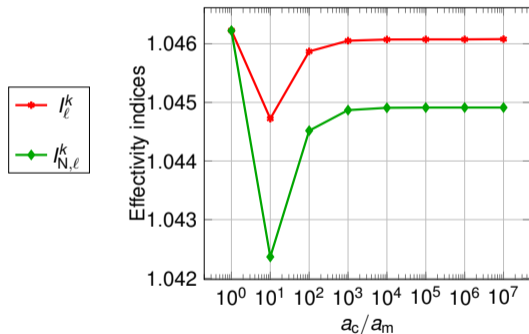


A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

How large is the error? Robustness wrt the nonlinearities

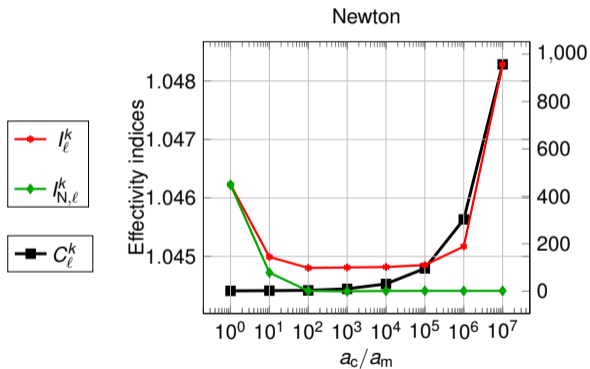
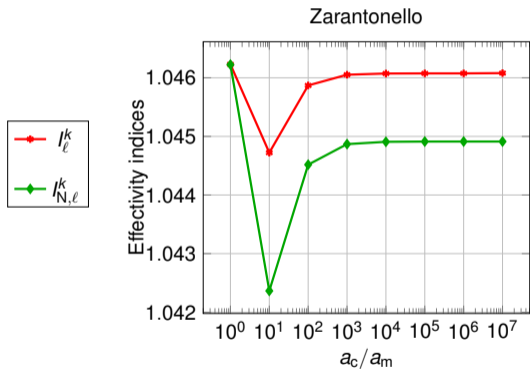
$$(a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}})$$

Zarantonello



How large is the error? Robustness wrt the nonlinearities

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



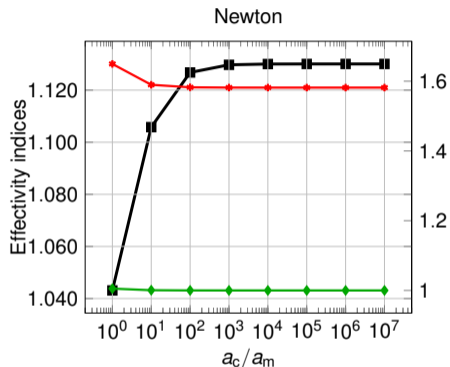
A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

Singular solution

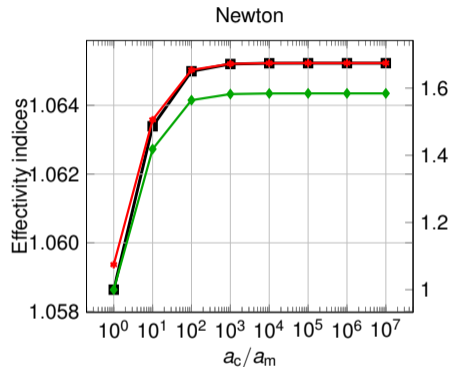
Setting

- 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

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Observation

Observation

Not all nonlinear problems admit an energy minimization structure.

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: 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

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

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

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

$$\left((u_{\langle \ell \rangle}^k - u_\ell^{k-1}), v \right)_{u_\ell^{k-1}} = - \underbrace{\left(\mathcal{R}(u_\ell^{k-1}), v \right)}_{\text{residual}} \quad \forall v \in H_0^1(\Omega)$$

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- error components

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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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One time step of the Richards equation

Setting

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

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

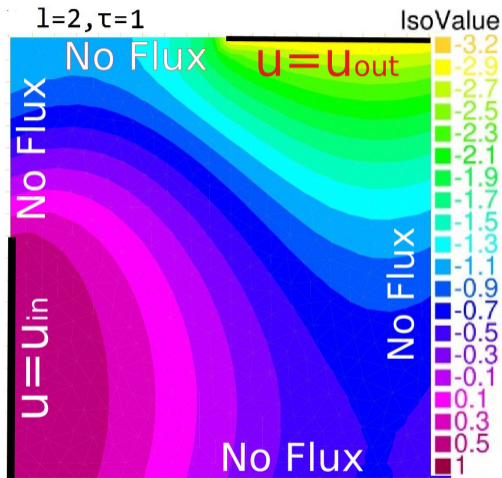
$$\mathbf{K} = \begin{bmatrix} 1 & 0.2 \\ 0.2 & 1 \end{bmatrix}, \quad \mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

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

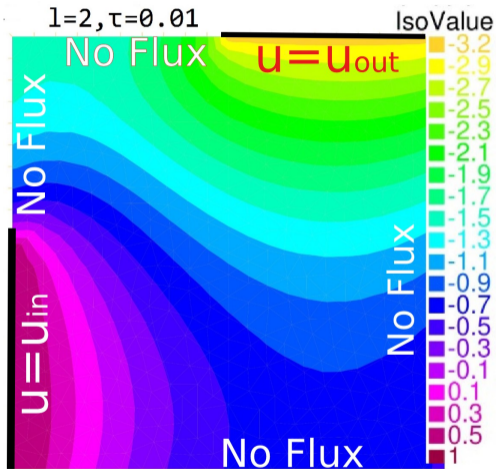
$$S(u) := \left(1 + (2 - u)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \quad \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^\lambda\right)^2, \quad \lambda = 0.5$$

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

One time step of the Richards equation: saturation u

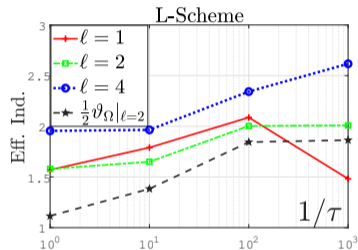
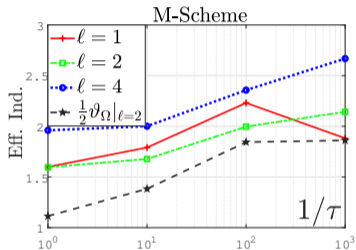
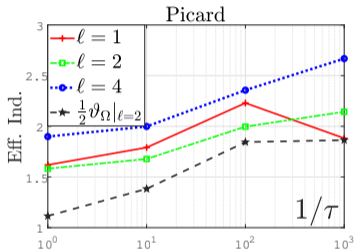


Time step length $\tau = 1$



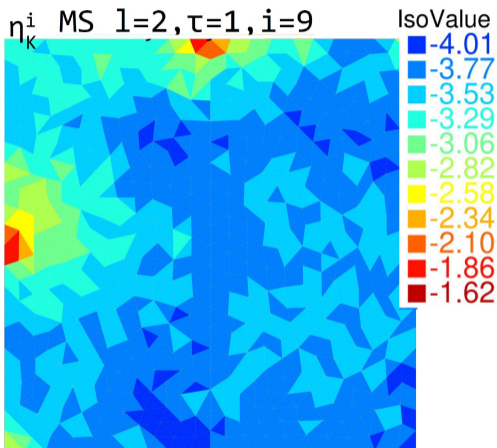
Time step length $\tau = 0.01$

How large is the error? Robustness wrt the nonlinearities

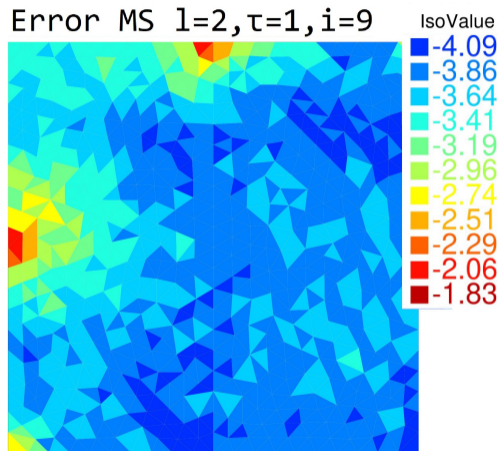


K. Mitra, M. Vohralík, preprint (2023)

Where is the error localized?

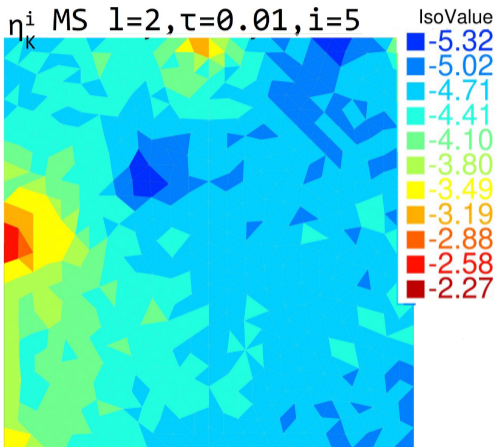


Estimated local error, $\tau = 1$

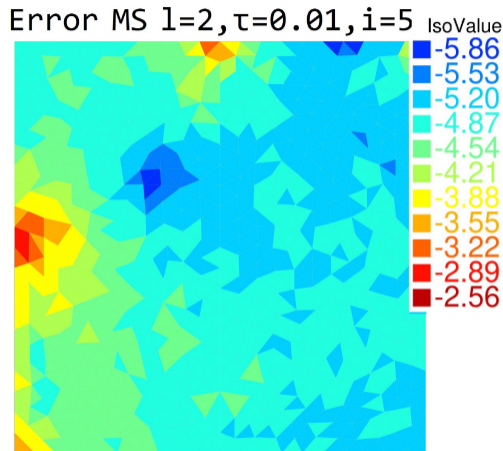


Exact local error, $\tau = 1$

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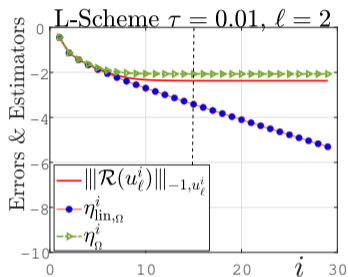
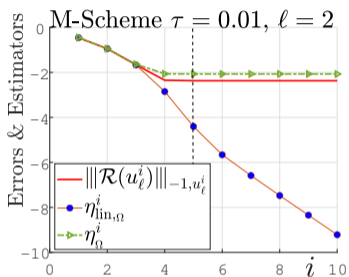
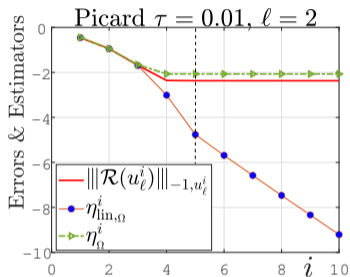


Estimated local error, $\tau = 0.01$



Exact local error, $\tau = 0.01$

Error components and adaptivity via stopping criteria



Time step length $\tau = 0.01$

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- a posteriori **certification** of the **error** for nonlinear problems
- **robustness** with respect to the **strength of nonlinearities**
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 HARNIST A., MITRA K., RAPPAPORT A., VOHRALÍK M. Robust augmented energy a posteriori estimates for Lipschitz and strongly monotone elliptic problems. HAL Preprint 04033438, 2023.


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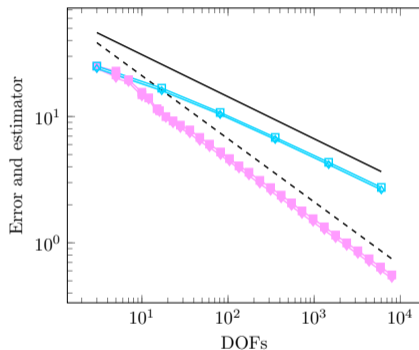
Thank you for your attention!

Outline

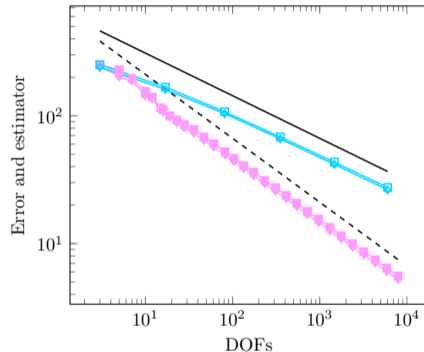
5 Adaptivity

6 Equilibrated flux reconstruction

Decreasing the error efficiently: optimal decay rate wrt DoFs



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



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

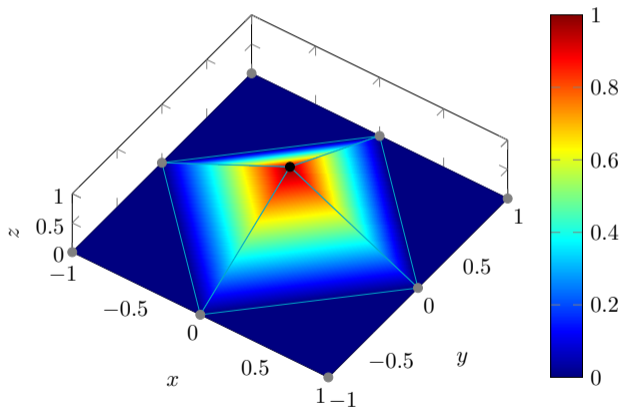
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Partition of unity

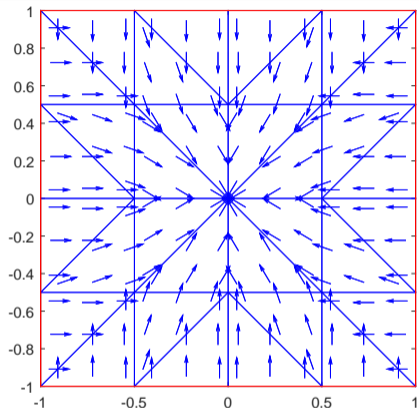
$$\sum_{\mathbf{a} \in \mathcal{V}_\ell} \psi^{\mathbf{a}} = 1$$



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

Equilibrated flux reconstruction

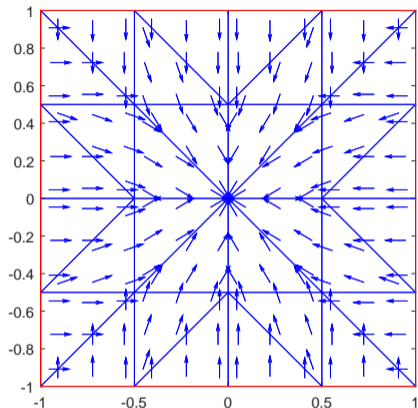
Destuynder and Métivet (1998), Braess & Schöberl (2008), Ern & Vohralík (2013)



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

Equilibrated flux reconstruction

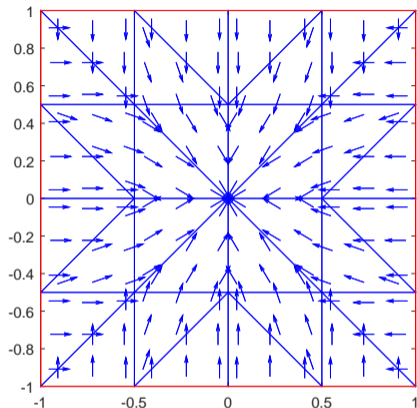
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Flux $\iota_\ell \notin H(\text{div}), \nabla \cdot \iota_\ell \neq f$

Equilibrated flux reconstruction

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

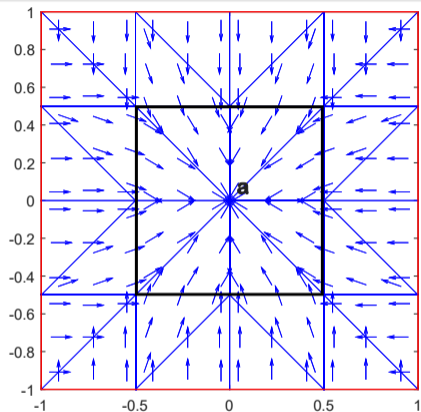


Flux $\boldsymbol{v}_\ell \notin \boldsymbol{H}(\text{div}), \nabla \cdot \boldsymbol{v}_\ell \neq f$

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

Equilibrated flux reconstruction

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



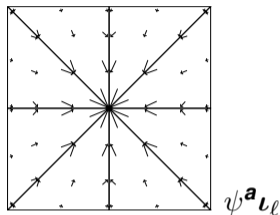
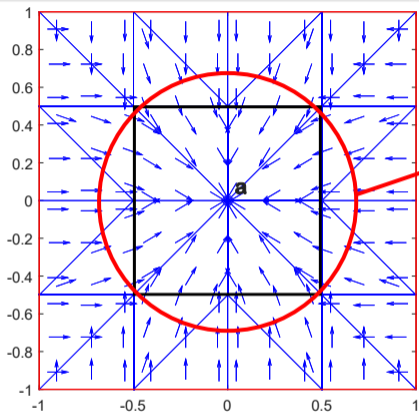
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$$\underbrace{\boldsymbol{v}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}$$

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

Equilibrated flux reconstruction

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

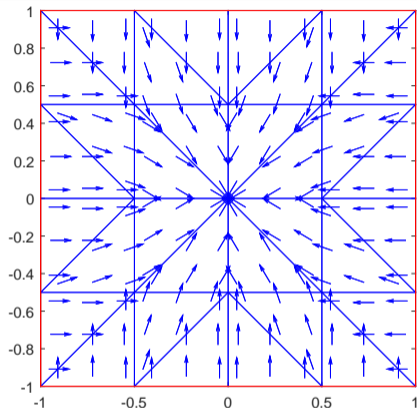


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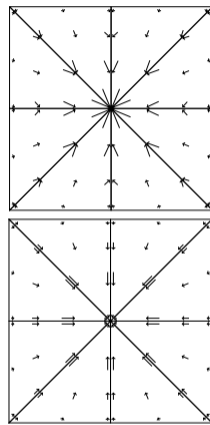
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$\psi^{\boldsymbol{a}}_{\boldsymbol{v}_\ell}$

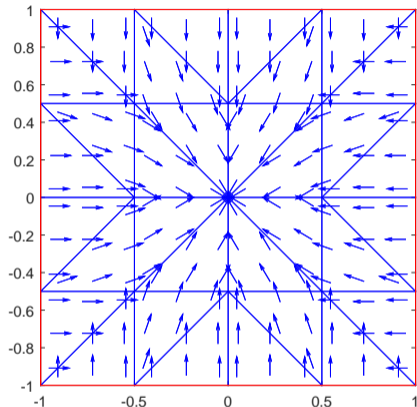
$\sigma_\ell^{\boldsymbol{a}}$

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

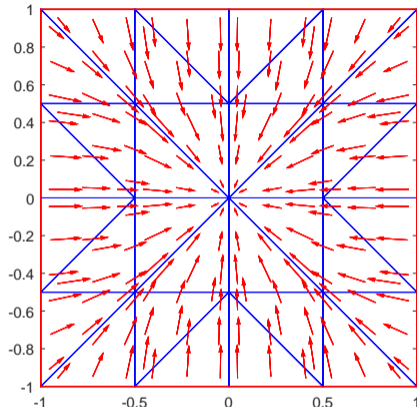
$$\sigma_\ell^{\boldsymbol{a}} := \arg \min_{\substack{\boldsymbol{v}_\ell \in \mathcal{RT}_{p+1}(\mathcal{T}_a) \cap \boldsymbol{H}_0(\text{div}, \omega_a) \\ \nabla \cdot \boldsymbol{v}_\ell = f \psi^{\boldsymbol{a}} + \boldsymbol{v}_\ell \cdot \nabla \psi^{\boldsymbol{a}}}} \|\psi^{\boldsymbol{a}}_{\boldsymbol{v}_\ell}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

Equilibrated flux reconstruction

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



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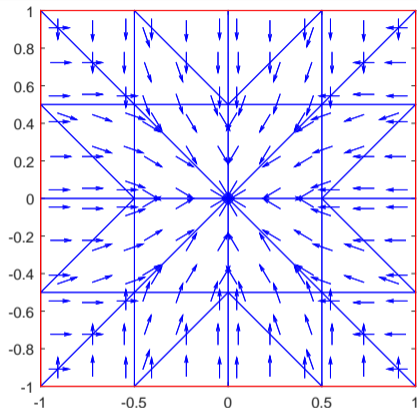


Equilibrated flux $\boldsymbol{\sigma}_\ell \in \mathbf{H}(\text{div})$, $\nabla \cdot \boldsymbol{\sigma}_\ell = f$

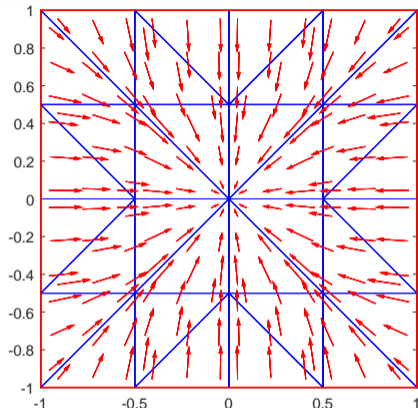
$$\underbrace{\boldsymbol{\nu}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)} \rightarrow \boldsymbol{\sigma}_\ell := \sum_{\mathbf{a} \in \mathcal{V}_\ell} \boldsymbol{\sigma}_\ell^{\mathbf{a}} \in \mathcal{RT}_{p+1}(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}), \nabla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

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