

A posteriori error estimates robust with respect to nonlinearities and final time

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Outline

- 1 Introduction
- 2 Equilibrated flux reconstruction
- 3 Steady linear problems
 - A posteriori error estimates
 - Recovering mass balance
- 4 Steady nonlinear problems
 - Gradient-dependent nonlinearities
 - A posteriori error estimates for an augmented energy difference
 - Numerical experiments
 - Gradient-independent nonlinearities
 - A posteriori error estimates for an iteration-dependent norm
 - Numerical experiments
- 5 Unsteady linear problems
- 6 The Richards equation (unsteady nonlinear degenerate parabolic problems)
- 7 Conclusions

Modelling flow of water and air through soil

The Richards equation

Find $u : \Omega \times (0, T) \rightarrow \mathbb{R}$ such that

$$\partial_t \mathbf{S}(u) - \nabla \cdot [\mathbf{K} \kappa(\mathbf{S}(u))(\nabla u + \mathbf{g})] = f \quad \text{in } \Omega \times (0, T),$$

$$u = 0 \quad \text{on } \partial\Omega \times (0, T),$$

$$(\mathbf{S}(u))(0) = s_0 \quad \text{in } \Omega.$$

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Setting

- u : pressure
- $s = S(u)$: saturation
- $\Omega \subset \mathbb{R}^d$, $1 \leq d \leq 3$, open bounded polytope with Lipschitz boundary $\partial\Omega$
- T : final time
- diffusion tensor \mathbf{K} , source term f , gravity \mathbf{g} , initial saturation s_0
- **nonlinear (degenerate) functions** S and κ

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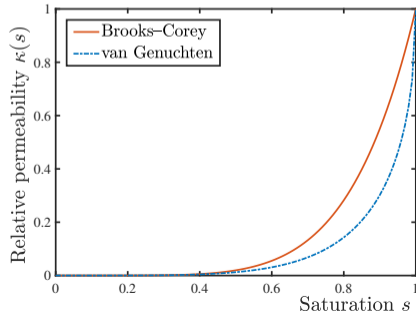
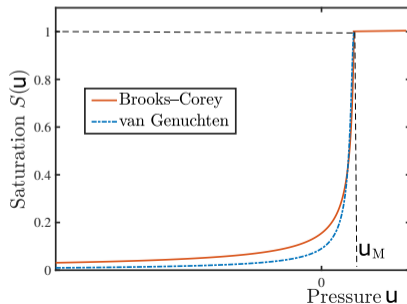
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Degeneracies

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Degeneracies

- parabolic–hyperbolic: $\kappa(0) = 0$ leads to

$$\partial_t S(u) = f$$

- parabolic–elliptic: $S'(u) = 0$ for $u > u_M$ leads to

$$-\nabla \cdot [\mathbf{K} \kappa(S(u))(\nabla u + \mathbf{g})] = f$$

A posteriori error estimates

Purpose

- provide sharp **computable bounds** on the unknown error between the unavailable exact solution and its numerical approximation
- predict the **error localization** (in space and in time)
- **adapt** the regularization parameters, linear solver, nonlinear solver, space mesh, time mesh ...

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Goals

Nonlinear problems

a posteriori error estimates

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

Goals

Nonlinear problems

Guaranteed a posteriori error estimates

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Goals

Nonlinear problems

Guaranteed a posteriori error estimates

efficient

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

Goals

Nonlinear problems

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

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

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

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

Goals

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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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Unsteady problems

Guaranteed a posteriori error estimates

$$\int_0^T \| \| u - u_\ell \| \|^2 \leq \sum_{n=1}^N \sum_{K \in \mathcal{T}_\ell^n} \eta_K^n(u_\ell)^2$$

Goals

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$$\int_0^T \|u - u_\ell\|^2 \leq \sum_{n=1}^N \sum_{K \in \mathcal{T}_\ell^n} \eta_K^n(u_\ell)^2 \leq C_{\text{eff}}^2 \int_0^T \|u - u_\ell\|^2,$$

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Guaranteed a posteriori error estimates **efficient** and **robust** with respect to the **final time**.

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Unsteady problems

Guaranteed a posteriori error estimates **locally space-time efficient** and **robust** with respect to the **final time**.

$$\eta_K^n(u_\ell)^2 \leq C_{\text{eff}}^2 \int_{t^{n-1}}^{t^n} \| \| u - u_\ell \| \|_{\omega_K}^2, \quad \text{for all } n \text{ and } K \in \mathcal{T}_\ell^n$$

Outline

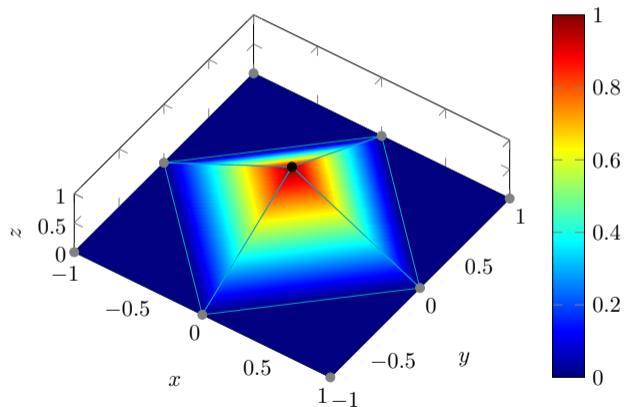
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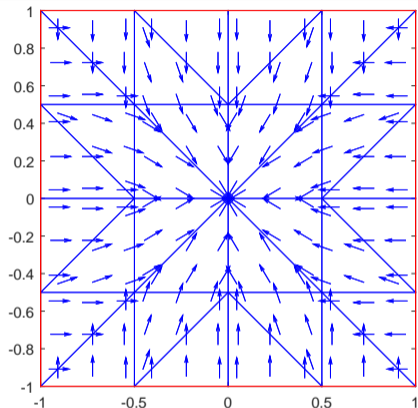
Partition of unity

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



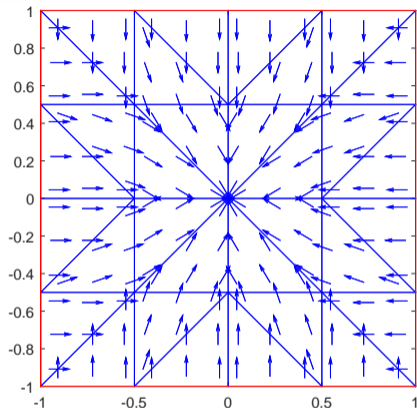
Hat basis function $\psi^{\mathbf{a}}$

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



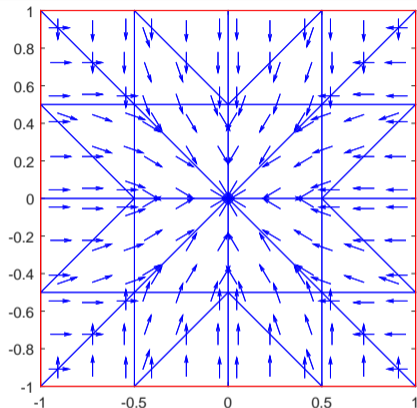
Flux $\iota_\ell \notin \mathbf{H}(\text{div})$ (e.g. FE flux $-\nabla u_\ell$)

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



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

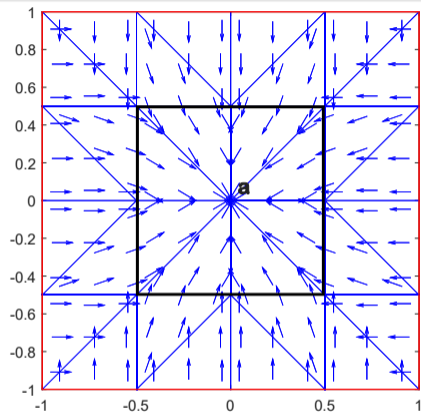
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Flux $\boldsymbol{v}_\ell \notin \boldsymbol{H}(\text{div}), \nabla \cdot \boldsymbol{v}_\ell \neq f$

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

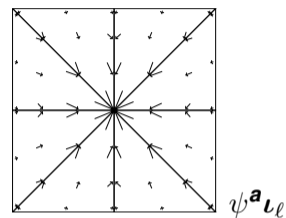
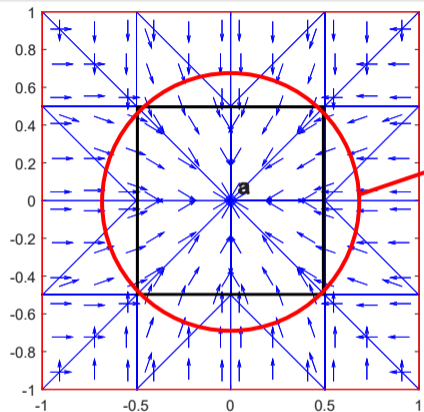
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$$\underbrace{\boldsymbol{v}_\ell \in \boldsymbol{RT}_\rho(\mathcal{T}_\ell), f \in \mathcal{P}_\rho(\mathcal{T}_\ell)}_{(f, \psi^{\boldsymbol{a}})_{\omega_{\boldsymbol{a}}} + (\boldsymbol{v}_\ell, \nabla \psi^{\boldsymbol{a}})_{\omega_{\boldsymbol{a}}}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

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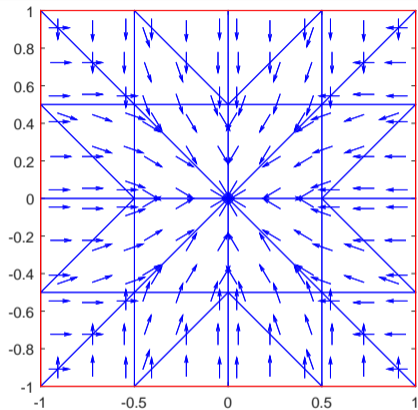


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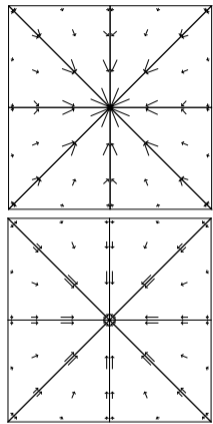
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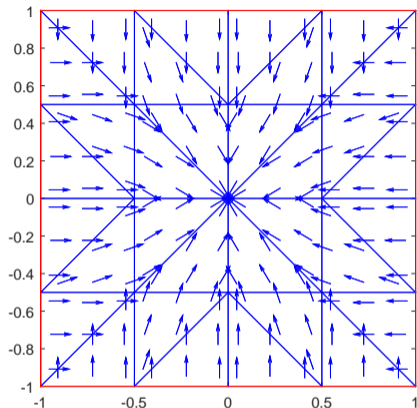
ψ^a

σ_ℓ^a

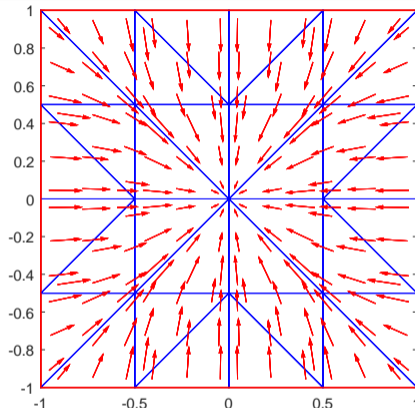
$$\underbrace{\boldsymbol{v}_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)}$$

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

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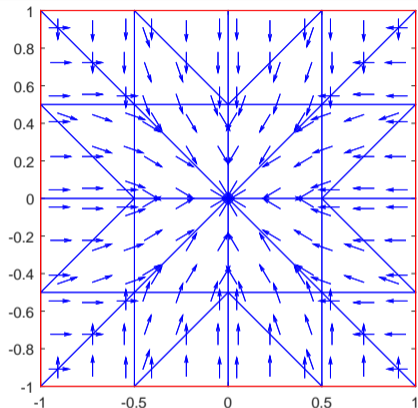
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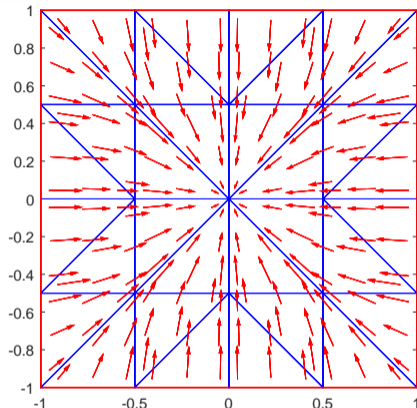
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$$\underbrace{\iota_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell), f \in \mathcal{P}_p(\mathcal{T}_\ell)} \rightarrow \sigma_\ell := \sum_{\mathbf{a} \in \mathcal{V}_\ell} \sigma_\ell^{\mathbf{a}} \in \mathcal{RT}_{p+1}(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}),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

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

$h_e \approx 1/ \mathcal{T}_e ^{1/2}$	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
$\approx h_0/2$	2	14%	12%	1.17
$\approx h_0/4$	3	7%	6%	1.17
$\approx h_0/8$	4	4%	3%	1.17
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A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

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

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$\approx h_0/2$	2	$8.9 \times 10^{-2}\%$		
$\approx h_0/4$	3	$5.9 \times 10^{-2}\%$		
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$\approx h_0/2$	2	$9.9 \times 10^{-2}\%$	$9.2 \times 10^{-2}\%$	1.07
$\approx h_0/4$	3	$5.9 \times 10^{-2}\%$	$5.9 \times 10^{-2}\%$	1.01
$\approx h_0/8$	4	$5.9 \times 10^{-2}\%$	$5.8 \times 10^{-2}\%$	1.02

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A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2012)

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

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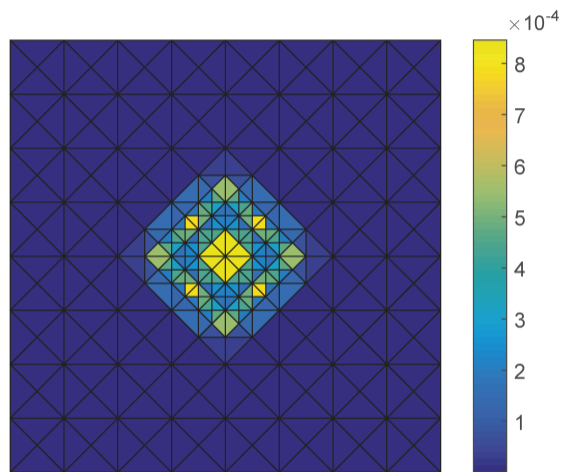
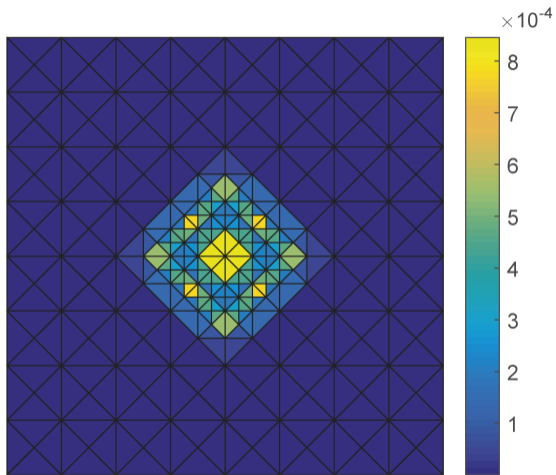
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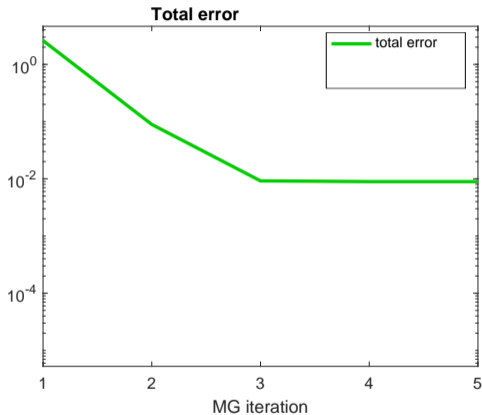
Where (in space) is the error **localized**? (steady linear Darcy)



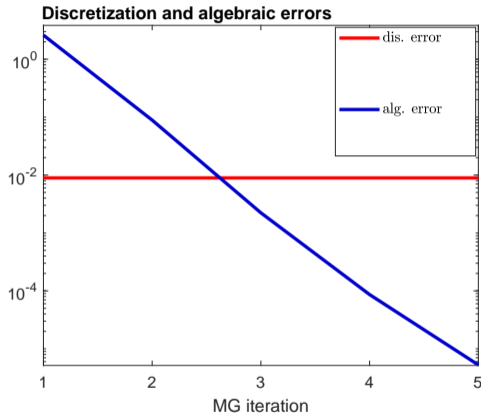
Estimated local error $\eta_K(u_\ell) = \|\nabla u_\ell + \sigma_\ell\|_K$

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

How large is the total error and its components? ($\mathbb{A}_l \mathbf{U}_l^i \neq \mathbf{F}_l$)



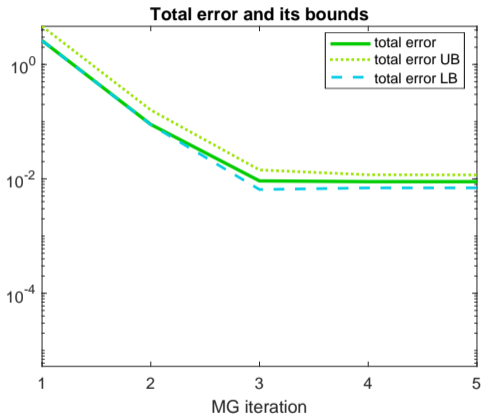
Total error



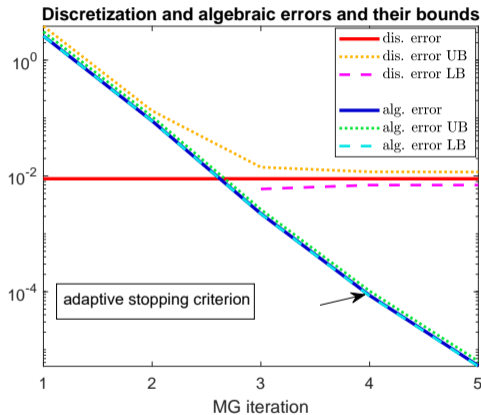
Error components

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

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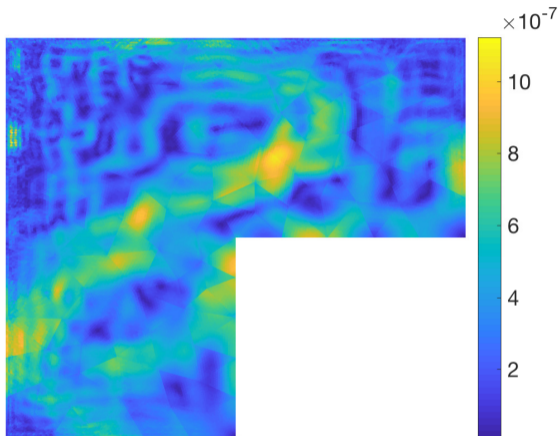
Total error



Error components and stopping criteria

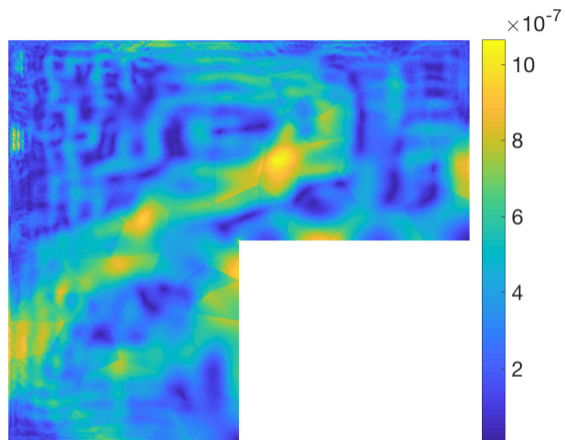
J. Papež, U. Růde, M. Vohralík, B. Wohlmuth, Computer Methods in Applied Mechanics and Engineering (2020)

Where (in space) is the algebraic error localized? ($\mathbb{A}_\ell \mathbf{U}_\ell^i \neq \mathbf{F}_\ell$)



Estimated local algebraic errors

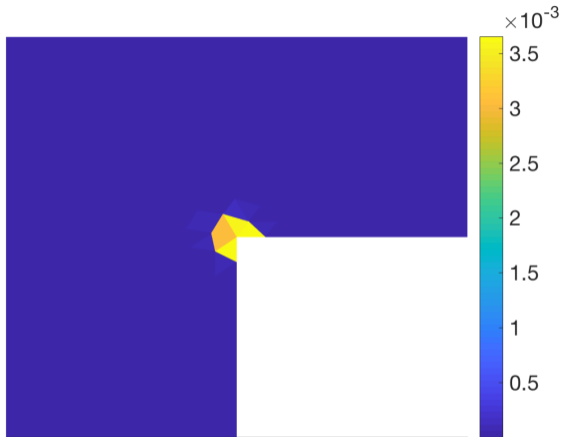
$$\eta_{\text{alg},\kappa}(u_\ell^i) = \|\sigma_{\text{alg},\ell}^i\|_\kappa$$



Exact local algebraic errors

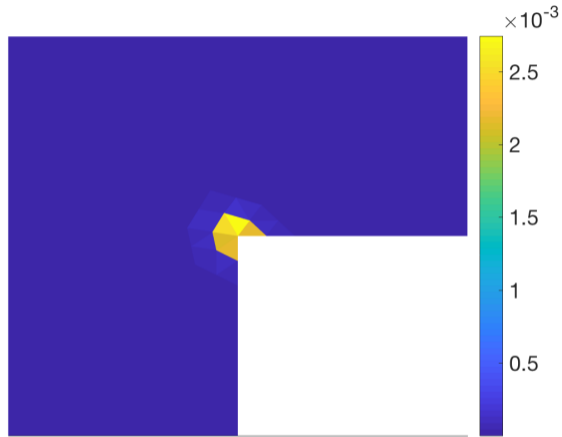
$$\|\nabla(u_\ell - u_\ell^i)\|_\kappa$$

Where (in space) is the total error localized? ($\mathbb{A}_\ell \mathbf{U}_\ell^i \neq \mathbf{F}_\ell$)



Estimated local total errors

$$\eta_{\mathcal{K}}(u_\ell^i) = \|\nabla u_\ell + \sigma_\ell^i\|_{\mathcal{K}}$$



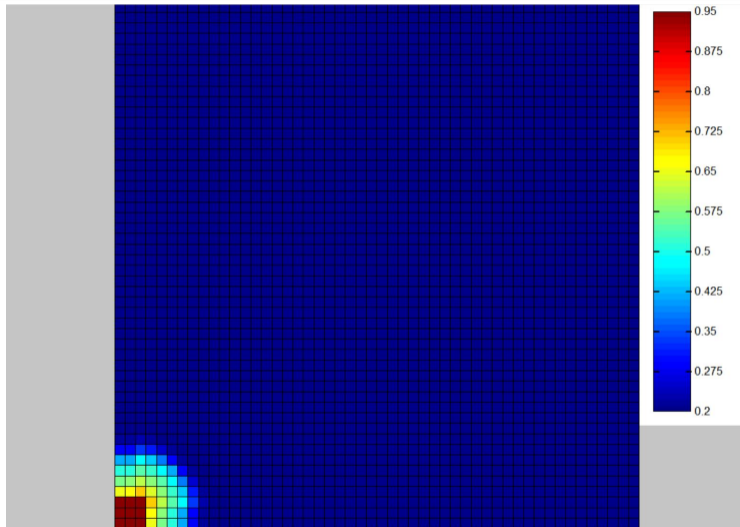
Exact local total errors

$$\|\nabla(u - u_\ell^i)\|_{\mathcal{K}}$$

Outline

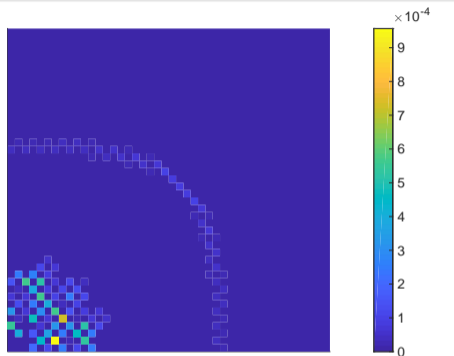
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- 2 Equilibrated flux reconstruction
- 3 **Steady linear problems**
 - A posteriori error estimates
 - **Recovering mass balance**
- 4 Steady nonlinear problems
 - Gradient-dependent nonlinearities
 - A posteriori error estimates for an augmented energy difference
 - Numerical experiments
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- 5 Unsteady linear problems
- 6 The Richards equation (unsteady nonlinear degenerate parabolic problems)
- 7 Conclusions

Two-phase flow, water saturation

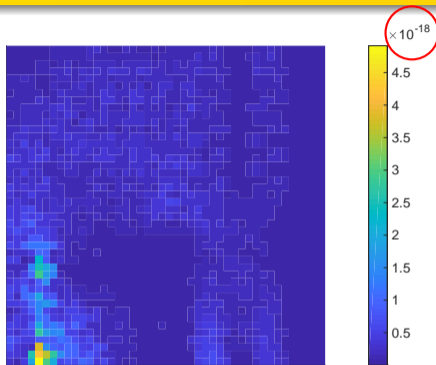


M. Vohralík, M. Wheeler, Computational Geosciences (2013)

Recovering mass balance: two-phase flow (inexact solver, water)



original mass balance misfit (m^2s^{-1})

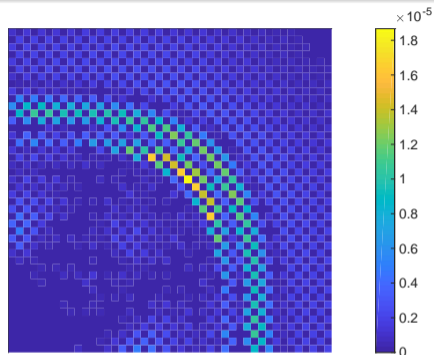


corrected mass balance misfit (m^2s^{-1})

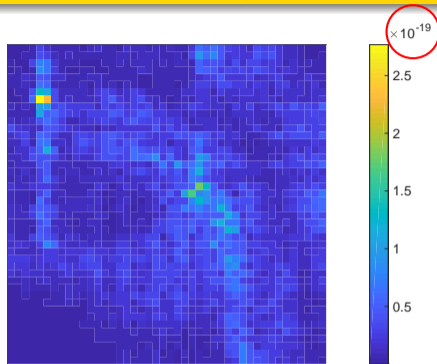
Setting

- fully implicit discretization of a two-phase oil–water flow
- cell-centered finite volumes on a square mesh
- time step 260, 1st Newton linearization, GMRes iteration 195

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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

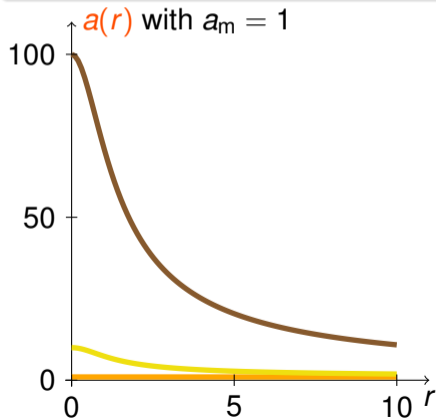
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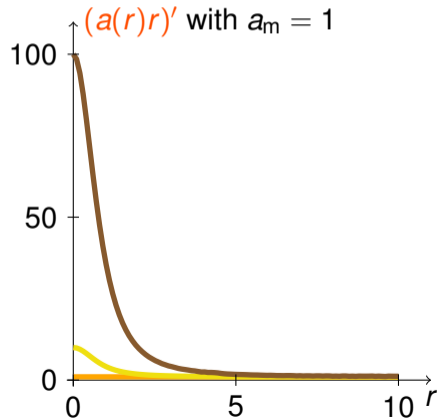
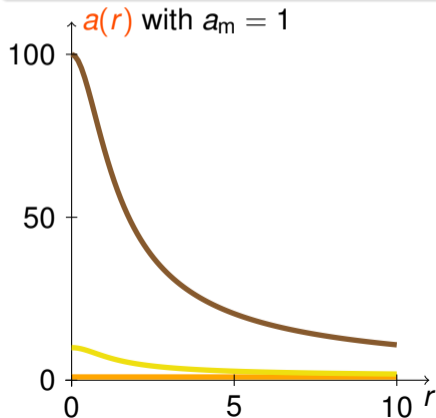


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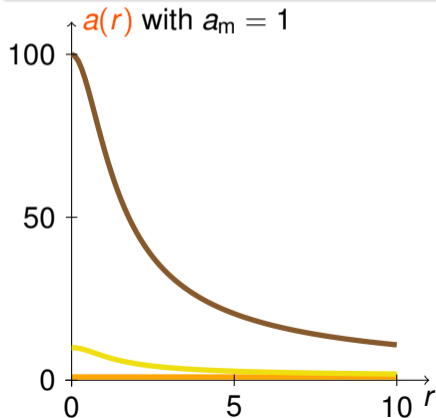


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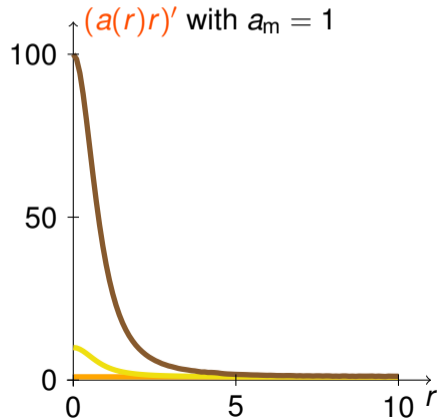
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Strength of the nonlinearity

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



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

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

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- \mathcal{T}_ℓ simplicial mesh of Ω
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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

Previous 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

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- Frittelli & Hager (1994), Verfürth (1995), Kim (1997), Houston, Süli & Wathen (2001), Gopal, Meunier & Zang (2011), Gopal, Meunier, Prasad & Wathen (2018), Heil & Winkler (2020)

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Need to **solve a nonlinear system**

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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)

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

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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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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

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

Outline

- 1 Introduction
- 2 Equilibrated flux reconstruction
- 3 Steady linear problems
 - A posteriori error estimates
 - Recovering mass balance
- 4 Steady nonlinear problems**
 - Gradient-dependent nonlinearities
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- 5 Unsteady linear problems
- 6 The Richards equation (unsteady nonlinear degenerate parabolic problems)
- 7 Conclusions

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

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$$\mathcal{E}_\ell^k := \frac{1}{2} \underbrace{(\mathcal{J}(u_\ell^k) - \mathcal{J}(u))}_{\text{energy difference}} + \lambda_\ell^k \frac{1}{2} \underbrace{(\mathcal{J}_\ell^{k-1}(u_\ell^k) - \mathcal{J}_\ell^{k-1}(u_{\langle \ell \rangle}^k))}_{\text{linearized en. diff.}}$$

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

A posteriori error estimates for an augmented energy difference

Augmented energy difference

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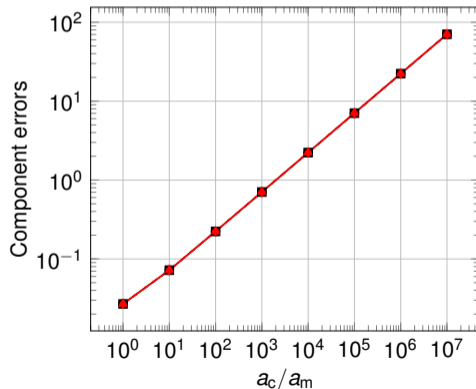
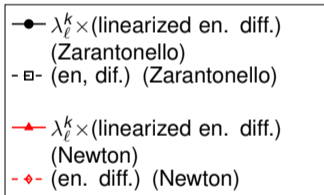
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- λ_ℓ^k computable weight to make the two components comparable

A posteriori error estimates for an augmented energy difference

Augmented energy difference

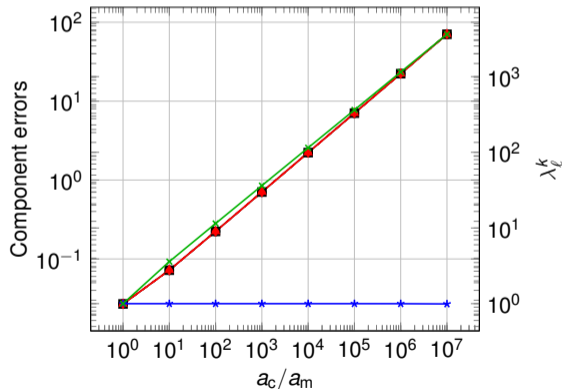
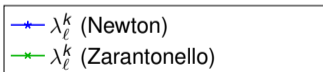
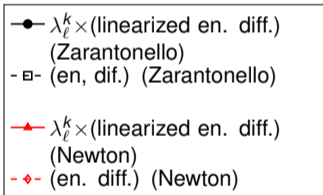
$$\mathcal{E}_\ell^k = \frac{1}{2} \text{energy difference} + \lambda_\ell^k \times \frac{1}{2} (\text{linearized energy difference})$$



A posteriori error estimates for an augmented energy difference

Augmented energy difference

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



A posteriori error estimates for an augmented energy difference

Augmented energy difference

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

Practically

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

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- 2 Equilibrated flux reconstruction
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- 7 Conclusions

Smooth solution

Setting

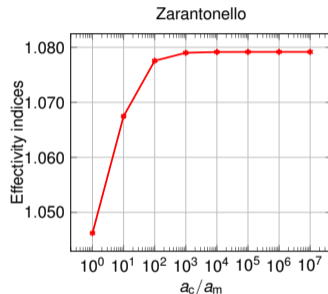
- unit square $\Omega = (0, 1)^2$
- known smooth solution $u(x, y) := 10 x(x - 1)y(y - 1)$
- $p = 1$
- effectivity indices

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

How large is the error? Robustness wrt the nonlinearities

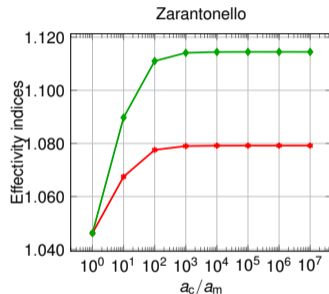
$$(a(r) = a_m + \frac{a_c - a_m}{\sqrt{1+r^2}})$$

J_ℓ^k



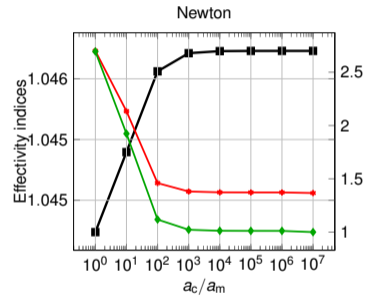
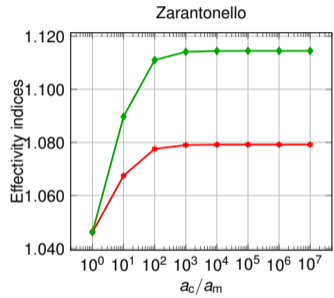
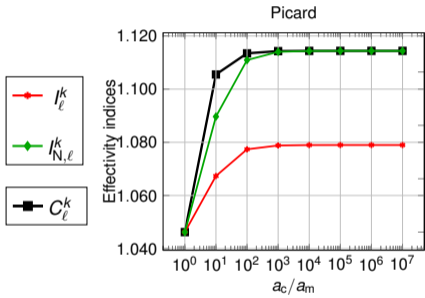
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How large is the error? Robustness wrt the nonlinearities

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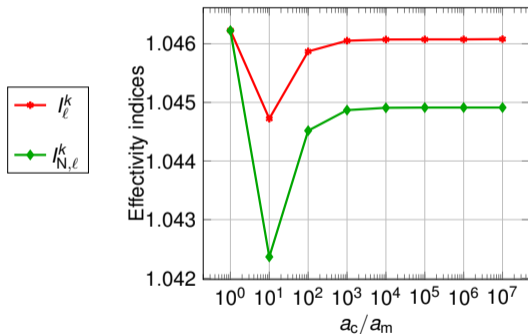
A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)



How large is the error? Robustness wrt the nonlinearities

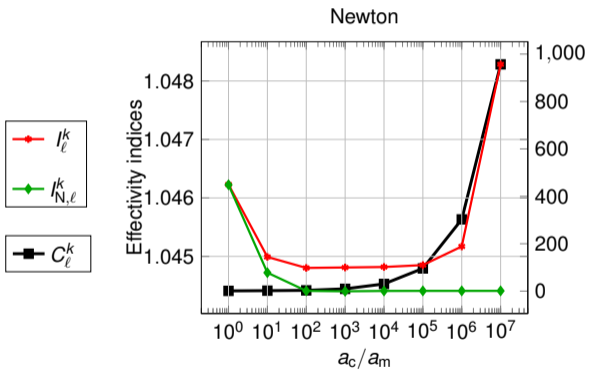
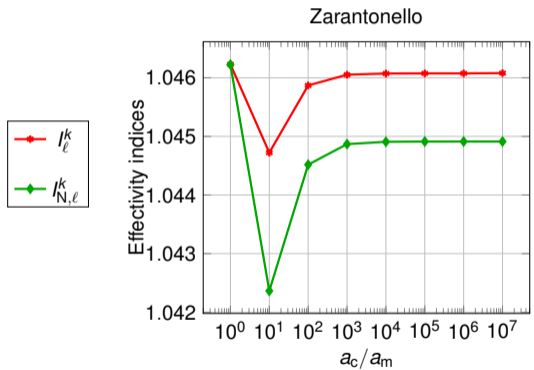
$$(a(r) = a_m + (a_c - a_m) \frac{1 - e^{-\frac{3}{2}r^2}}{1 + 2e^{-\frac{3}{2}r^2}})$$

Zarantonello



How large is the error? Robustness wrt the nonlinearities

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



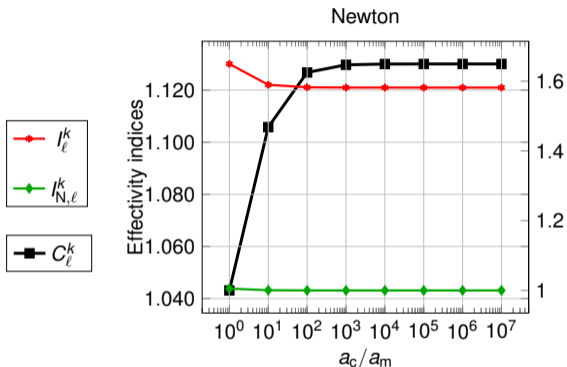
A. Harnist, K. Mitra, A. Rappaport, M. Vohralík, preprint (2023)

Singular solution

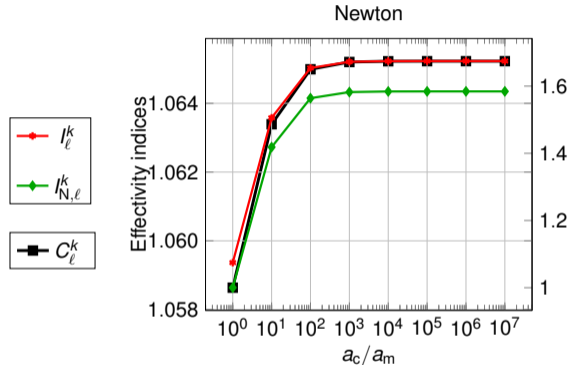
Setting

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

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

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

- $\tau > 0$ 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

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

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

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

Iteration-dependent norm

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

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

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- **orthogonal decomposition**
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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 $k \geq 1$,

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

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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 $k \geq 1$, there holds

$$\eta(u_\ell^k) \leq C_{\text{eff}}(d, \kappa_T) C_\ell^k \|\mathcal{R}(u_\ell^{k-1})\|_{-1, u_\ell^{k-1}} + \text{quadrature error terms},$$

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

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

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

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- ✓ C_K^k given by **local conditioning** of the linearization matrix \mathbf{A}_ℓ^{k-1} : typically **much better** than global conditioning (= worst-case scenario)

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One time step of the Richards equation

Setting

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

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

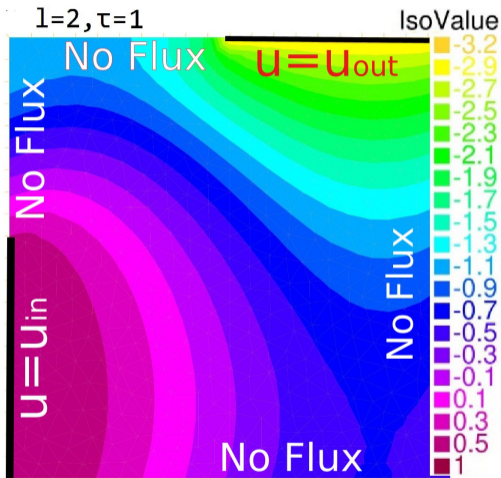
$$\mathbf{K} = \begin{bmatrix} 1 & 0.2 \\ 0.2 & 1 \end{bmatrix}, \quad \mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

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

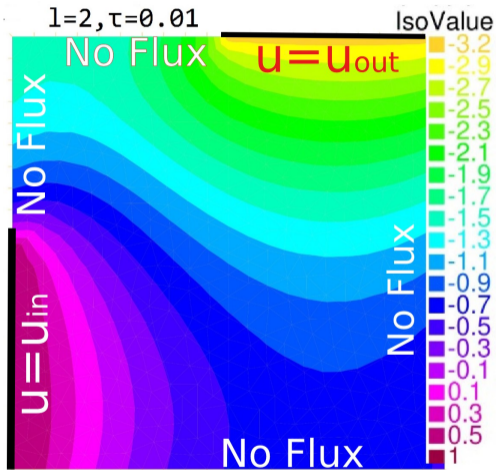
$$S(u) := \left(1 + (2 - u)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \quad \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^{\lambda}\right)^2, \quad \lambda = 0.5$$

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

One time step of the Richards equation: saturation u

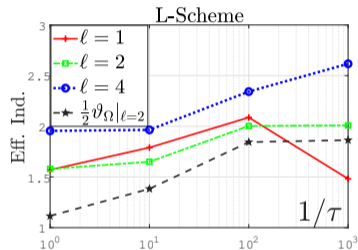
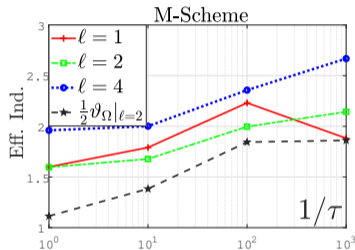
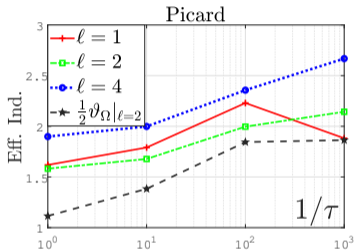


Time step length $\tau = 1$



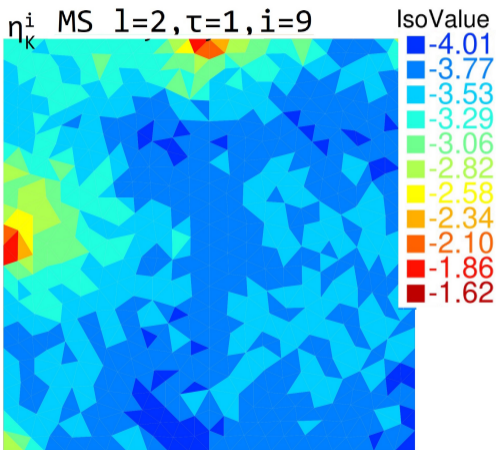
Time step length $\tau = 0.01$

How large is the error? Robustness wrt the nonlinearities

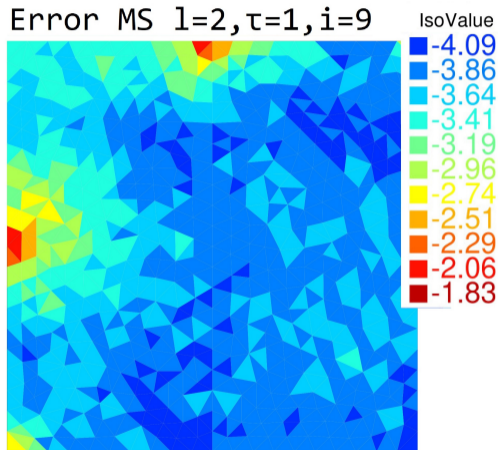


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Where is the error localized?

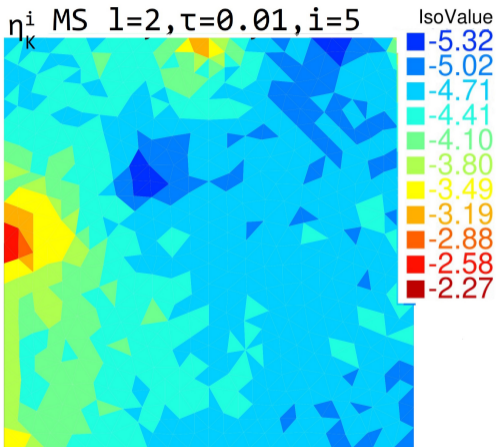


Estimated local error, $\tau = 1$

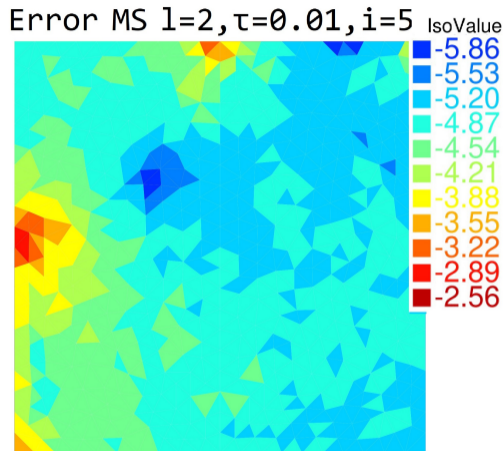


Exact local error, $\tau = 1$

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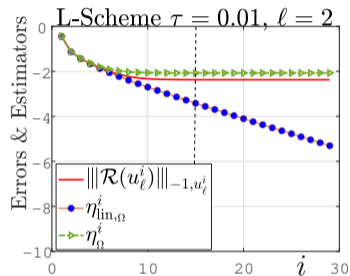
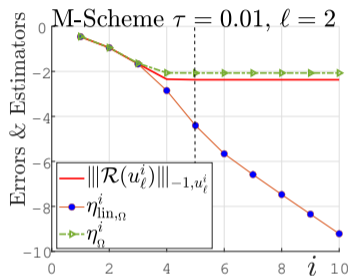
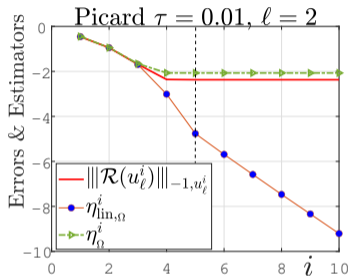


Estimated local error, $\tau = 0.01$



Exact local error, $\tau = 0.01$

Error components and adaptivity via stopping criteria



Time step length $\tau = 0.01$

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A model unsteady linear problem

The unsteady linear Darcy (heat) equation

Find $u : \Omega \times (0, T) \rightarrow \mathbb{R}$ such that

$$\partial_t u - \Delta u = f \quad \text{in } \Omega \times (0, T),$$

$$u = 0 \quad \text{on } \partial\Omega \times (0, T),$$

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- T : final time
- f and u_0 piecewise polynomial for simplicity

Spaces and norms

$$X := L^2(0, T; H_0^1(\Omega)),$$

$$\|v\|_X^2 := \int_0^T \|\nabla v\|^2 dt,$$

$$Y := L^2(0, T; H_0^1(\Omega)) \cap H^1(0, T; H^{-1}(\Omega)),$$

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$u \in Y$ with $u(0) = u_0$ such that

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Nonsymmetry

Trial space Y , test space X .

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Previous results

- Picasso / Verfürth (1998), work with the energy norm of X :
 - ✓ upper bound $\|u - u_\ell\|_X^2 \leq C^2 \sum_{n=1}^N \sum_{K \in \mathcal{T}_\ell^n} \eta_K^n(u_\ell)^2$
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Augmented Y norm

$$\|u - u_e\|_{\mathcal{E}_Y}^2 := \|u - \mathcal{I}u_e\|_Y^2 + \underbrace{\|u_e - \mathcal{I}u_e\|_X^2}_{\text{known, computable, measures time jumps}}$$

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Modelling flow of water and air through soil

The Richards equation

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- u : pressure
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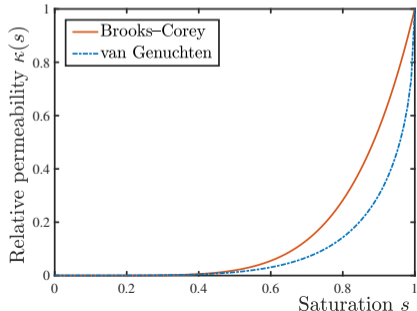
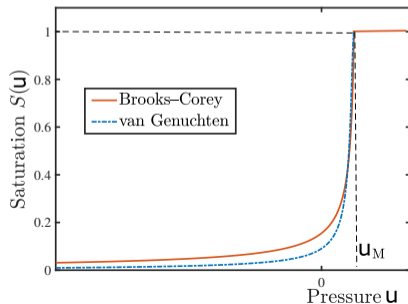
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A posteriori error estimates

Nonsymmetry, nonlinearity, degeneracy

- Use all the tools from the above cases.
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A posteriori error estimates

Nonsymmetry, nonlinearity, degeneracy

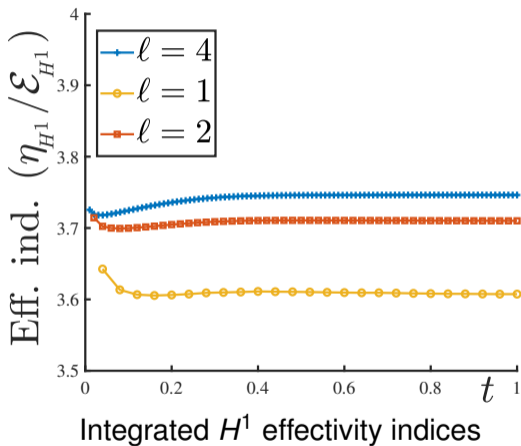
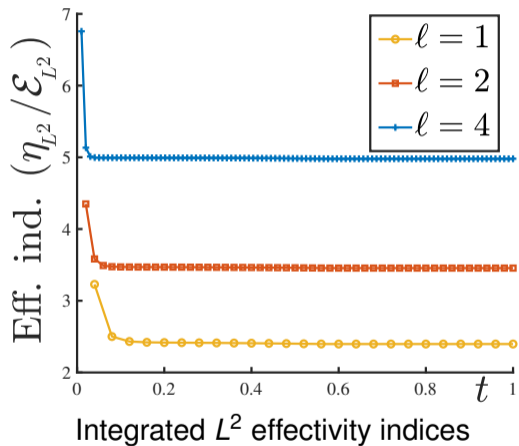
- Use all the tools from the above cases.
- Treatment of time-dependent nonlinearity: **sharp Gronwall lemma** not neglecting the integral terms.
- **Avoids** the appearance of the usual factor e^T , but gives rise to **integrated norms**.
- ✓ **local** in **space** and in **time** efficiency
- ✓ **robustness** w.r.t the final time T
- ✗ **heuristic** estimators for the **treatment of degeneracy**
- ✗ **norm change** between efficiency and reliability
- ✗ **no robustness** wrt the **strength of nonlinearities**

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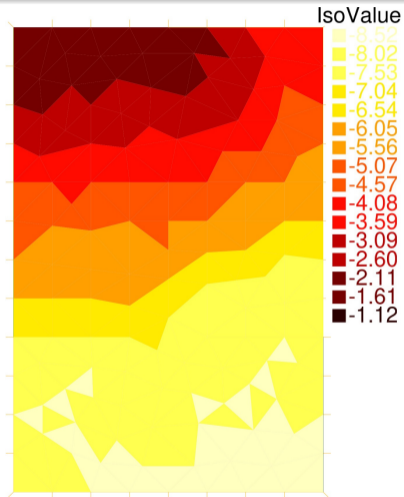
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- Details in K. Mitra, M. Vohralík, preprint (2022)

How large is the error? **Robustness** wrt the final time (known sol.)

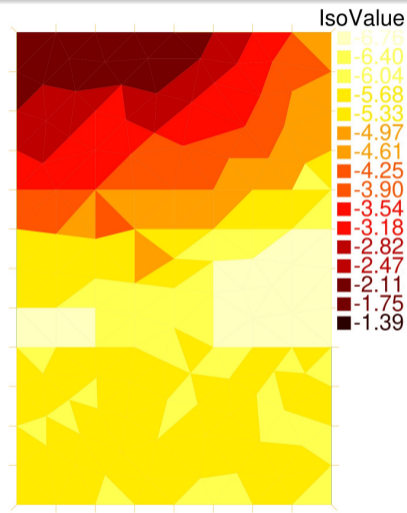


K. Mitra, M. Vohralík, preprint (2022)

Where (in space and time) is the error **localized**? (benchmark case)



Estimated local error



Exact local error

Realistic case

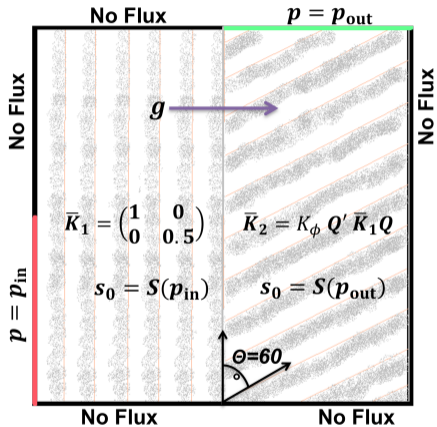
Setting

- unit square $\Omega = (0, 1)^2$
- $T = 1$
- $f(\mathbf{x}, u) = 0$, heterogeneous and anisotropic \mathbf{K} , $\mathbf{g} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$
- **Brooks–Corey**-type **saturation** and **permeability** laws

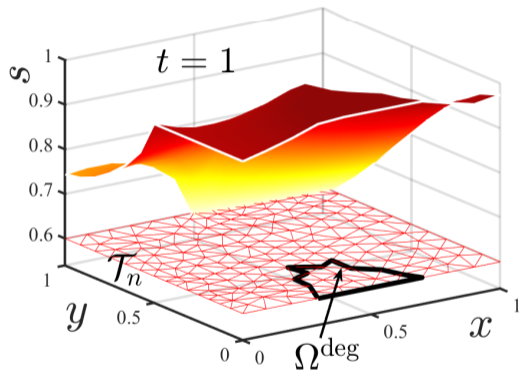
$$S(u) := \begin{cases} \frac{1}{(2-u)^{\frac{1}{3}}} & \text{if } u < 1, \\ 1 & \text{if } u \geq 1 \end{cases}, \quad \kappa(s) := s^3$$

- $(h, \tau) = (h_0, \tau_0)/\ell$ with $\ell \in \{1, 2, 4\}$, $h_0 = 0.2$, and $\tau_0 = 0.04$

Realistic case

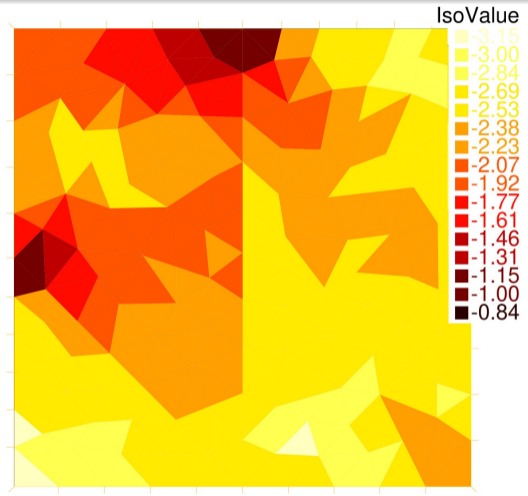


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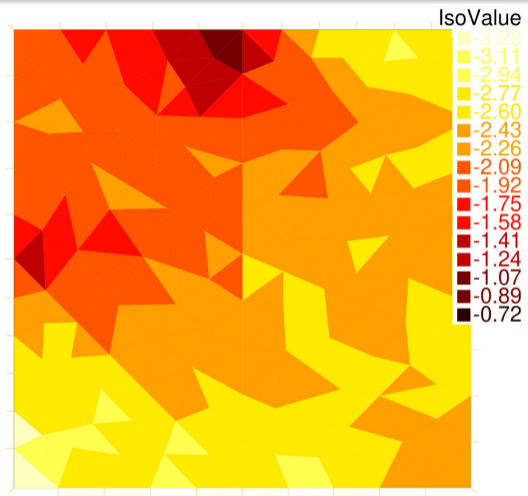


Numerical saturation for $\ell = 2$ at $t = 1$

Where (in space and time) is the error **localized**? (realistic test case)



Estimated local error



Exact local error

K. Mitra, M. Vohralík, preprint (2022)

Outline

- 1 Introduction
- 2 Equilibrated flux reconstruction
- 3 Steady linear problems
 - A posteriori error estimates
 - Recovering mass balance
- 4 Steady nonlinear problems
 - Gradient-dependent nonlinearities
 - A posteriori error estimates for an augmented energy difference
 - Numerical experiments
 - Gradient-independent nonlinearities
 - A posteriori error estimates for an iteration-dependent norm
 - Numerical experiments
- 5 Unsteady linear problems
- 6 The Richards equation (unsteady nonlinear degenerate parabolic problems)
- 7 Conclusions

Conclusions


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- a posteriori **certification** of the **error** for nonlinear and unsteady problems
- **robustness** with respect to the **strength of nonlinearities** and **final time** for model cases (nonlinear or unsteady)
- **localization** of the error in **space** and in **time**
- **theory** and **sound numerical performance** for the Richards equation


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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

- 8 Other error measures
- 9 Fenchel conjugate, dual energy, flux equilibration
- 10 Adaptivity
- 11 Two-phase flow

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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Fenchel conjugate, dual energy, flux equilibration

Definition (Fenchel conjugate)

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

Definition (Dual energy)

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

Definition (Flux equilibration)

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

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

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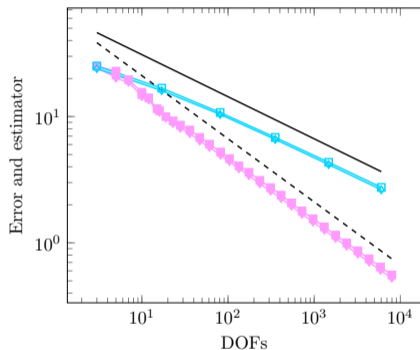
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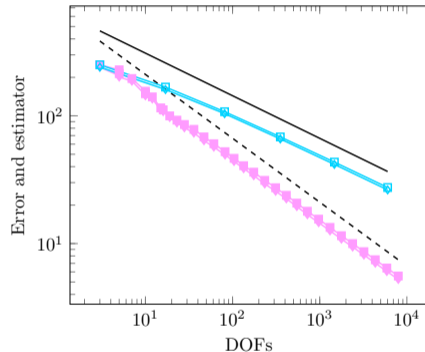
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Decreasing the error efficiently: optimal decay rate wrt DoFs



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

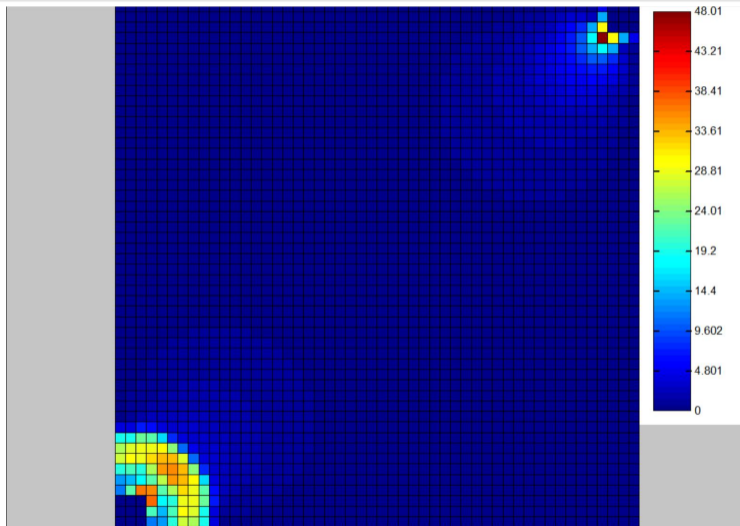


$$\frac{a_c}{a_m} = 10^6$$

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Where (in space and time) is the error **localized**? (two-phase flow)



M. Vohralík, M. Wheeler, Computational Geosciences (2013)

All error components (two-phase flow)

