

Adaptive inexact Newton methods with a posteriori stopping criteria for nonlinear diffusion PDEs

Martin Vohralík and Alexandre Ern

INRIA Paris-Rocquencourt

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Outline

1 Introduction

2 Adaptive inexact Newton method

- Quasi-linear elliptic problems
- Approximate solution and error measure
- A guaranteed a posteriori error estimate
- Stopping criteria and efficiency

3 Examples of application

- Arbitrary iterative algebraic solver
- Nonconforming FEs, fixed point/Newton, p -Laplacian
- DGs, fixed point/Newton, quasi-linear diffusion
- Summary

4 Numerical results

5 An extension to two-phase flows

6 Conclusions and future directions

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Inexact Newton method

System of nonlinear algebraic equations

Nonlinear operator $\mathcal{A}: \mathbb{R}^N \rightarrow \mathbb{R}^N$, vector $F \in \mathbb{R}^N$: find $U \in \mathbb{R}^N$ s.t.

$$\mathcal{A}(U) = F$$

Algorithm (Inexact linearization)

1 Choose initial vector U^0 . Set $k := 1$.

2 $U^{k-1} \Rightarrow$ matrix \mathbb{A}^k and vector F^k : find U^k s.t.

$$\mathbb{A}^k U^k \approx F^k.$$

3 1 Set $U^{k,0} := U^{k-1}$ and $i := 1$.

2 Do 1 algebraic solver step $\Rightarrow U^{k,i}$ s.t. ($R^{k,i}$ algebraic res.)

$$\mathbb{A}^k U^{k,i} = F^k - R^{k,i}.$$

3 Convergence? OK $\Rightarrow U^k := U^{k,i}$. KO $\Rightarrow i := i + 1$, back to 3.2.

4 Convergence? OK \Rightarrow finish. KO $\Rightarrow k := k + 1$, back to 2.

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Context and questions

Approximate solution

- approximate solution $U^{k,i}$ does **not solve** $\mathcal{A}(U^{k,i}) = F$

Numerical method

- underlying numerical method: the vector $U^{k,i}$ is associated with a (piecewise polynomial) **approximation** $u_h^{k,i}$

Partial differential equation

- underlying PDE, u its **weak solution**: $A(u) = f$

Question (Stopping criteria)

- *What is a good stopping criterion for the nonlinear solver?*
- *What is a good stopping criterion for the linear solver?*

Question (Error)

- *How big is the error $\|u - u_h^{k,i}\|$ on Newton step k and algebraic solver step i , how is it distributed?*

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Previous results

Inexact Newton method

- Eisenstat and Walker (1990's)
- Moret (1989)

Stopping criteria

- engineering literature, since 1950's
- Becker, Johnson, and Rannacher (1995), multigrid st. crit.
- Arioli (2000's)

A posteriori error estimates for nonlinear problems

- Ladevèze (since 1990's), guaranteed upper bound
- Han (1994), general framework
- Verfürth (1994), residual estimates
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- Chaillou and Suri (2006, 2007), distinguishing
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Quasi-linear elliptic problem

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

- quasi-linear diffusion problem

$$\sigma(x, v, \xi) = \underline{A}(x, v)\xi \quad \forall (x, v, \xi) \in \Omega \times \mathbb{R} \times \mathbb{R}^d$$

- Leray–Lions problem

$$\sigma(x, v, \xi) = \underline{A}(x, \xi)\xi \quad \forall (x, \xi) \in \Omega \times \mathbb{R}^d$$

- $p > 1$, $q := \frac{p}{p-1}$, $f \in L^q(\Omega)$

Example

p -Laplacian: Leray–Lions setting with $\underline{A}(x, \xi) = |\xi|^{p-2}\mathbf{I}$

Nonlinear operator $A : V := W_0^{1,p}(\Omega) \rightarrow V'$

$$\langle A(u), v \rangle_{V', V} := (\sigma(\cdot, u, \nabla u), \nabla v)$$

Weak formulation

Find $u \in V$ such that

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Approximate solution and error measure

Approximate solution

- $u_h^{k,i} \in V(\mathcal{T}_h) \not\subset V$, $u_h^{k,i}$ not necessarily in V
- $V(\mathcal{T}_h) := \{v \in L^p(\Omega), v|_K \in W^{1,p}(K) \quad \forall K \in \mathcal{T}_h\}$

Error measure

$$\mathcal{J}_u(u_h^{k,i}) := \sup_{\varphi \in V; \|\nabla \varphi\|_p=1} (\sigma(u, \nabla u) - \sigma(u_h^{k,i}, \nabla u_h^{k,i}), \nabla \varphi) + \mathcal{J}_{u,\text{NC}}(u_h^{k,i})$$

$$\mathcal{J}_{u,\text{NC}}(u_h^{k,i}) := \left\{ \sum_{K \in \mathcal{T}_h} \sum_{e \in \mathcal{E}_K} h_e^{1-q} \|[\![u - u_h^{k,i}]\!] \|_{q,e}^q \right\}^{1/q}$$

- weak difference of the fluxes (dual norm of the residual) + nonconformity (computable jump term)
- there holds $\mathcal{J}_u(u_h^{k,i}) = 0$ if and only if $u = u_h^{k,i}$
- physical relevance: strong difference of the fluxes + nonconformity

$$\mathcal{J}_u(u_h^{k,i}) \leq \mathcal{J}_u^{\text{up}}(u_h^{k,i}) := \|\sigma(u, \nabla u) - \sigma(u_h^{k,i}, \nabla u_h^{k,i})\|_q + \mathcal{J}_{u,\text{NC}}(u_h^{k,i})$$

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- $u_h^{k,i} \in V(\mathcal{T}_h) \not\subset V$, $u_h^{k,i}$ not necessarily in V
- $V(\mathcal{T}_h) := \{v \in L^p(\Omega), v|_K \in W^{1,p}(K) \quad \forall K \in \mathcal{T}_h\}$

Error measure

$$\mathcal{J}_u(u_h^{k,i}) := \sup_{\varphi \in V; \|\nabla \varphi\|_p=1} (\sigma(u, \nabla u) - \sigma(u_h^{k,i}, \nabla u_h^{k,i}), \nabla \varphi) + \mathcal{J}_{u,\text{NC}}(u_h^{k,i})$$

$$\mathcal{J}_{u,\text{NC}}(u_h^{k,i}) := \left\{ \sum_{K \in \mathcal{T}_h} \sum_{e \in \mathcal{E}_K} h_e^{1-q} \| [u - u_h^{k,i}] \|_{q,e}^q \right\}^{1/q}$$

- weak difference of the fluxes (dual norm of the residual) + nonconformity (computable jump term)
- there holds $\mathcal{J}_u(u_h^{k,i}) = 0$ if and only if $u = u_h^{k,i}$
- physical relevance: strong difference of the fluxes + nonconformity

$$\mathcal{J}_u(u_h^{k,i}) \leq \mathcal{J}_u^{\text{up}}(u_h^{k,i}) := \|\sigma(u, \nabla u) - \sigma(u_h^{k,i}, \nabla u_h^{k,i})\|_q + \mathcal{J}_{u,\text{NC}}(u_h^{k,i})$$

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A posteriori error estimate

Assumption A (Total flux reconstruction)

There exists a **flux reconstruction** $\mathbf{t}_h^{k,i} \in \mathbf{H}^q(\text{div}, \Omega)$ and an **algebraic remainder** $\rho_h^{k,i} \in L^q(\Omega)$ such that

$$\nabla \cdot \mathbf{t}_h^{k,i} = f_h - \rho_h^{k,i},$$

with the data approximation f_h s.t. $(f_h, 1)_K = (f, 1)_K \quad \forall K \in \mathcal{T}_h$.

Theorem (A posteriori error estimate)

Let

- $u \in V$ be the weak solution,
- $u_h^{k,i} \in V(\mathcal{T}_h)$ be arbitrary,
- Assumption A hold.

Then there holds

$$\mathcal{J}_u(u_h^{k,i}) \leq \bar{\eta}^{k,i},$$

where $\bar{\eta}^{k,i}$ is fully computable from $u_h^{k,i}$, $\mathbf{t}_h^{k,i}$, and $\rho_h^{k,i}$.

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Distinguishing error components

Assumption B (Discretization, linearization, and algebraic errors)

There exist fluxes $\mathbf{d}_h^{k,i}, \mathbf{l}_h^{k,i}, \mathbf{a}_h^{k,i} \in [L^q(\Omega)]^d$ such that

- (i) $\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i} + \mathbf{a}_h^{k,i} = \mathbf{t}_h^{k,i}$;
- (ii) as the linear solver converges, $\|\mathbf{a}_h^{k,i}\|_q \rightarrow 0$;
- (iii) as the nonlinear solver converges, $\|\mathbf{l}_h^{k,i}\|_q \rightarrow 0$.

Comments

- $\mathbf{d}_h^{k,i}$: *discretization flux reconstruction*
- $\mathbf{l}_h^{k,i}$: *linearization error flux reconstruction*
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Estimate distinguishing error components

Theorem (Estimate distinguishing different error components)

Let

- $u \in V$ be the weak solution,
- $u_h^{k,i} \in V(\mathcal{T}_h)$ be arbitrary,
- Assumptions A and B hold.

Then there holds

$$\mathcal{J}_u(u_h^{k,i}) \leq \eta^{k,i} := \eta_{\text{disc}}^{k,i} + \eta_{\text{lin}}^{k,i} + \eta_{\text{alg}}^{k,i} + \eta_{\text{rem}}^{k,i} + \eta_{\text{quad}}^{k,i} + \eta_{\text{osc}}^{k,i}.$$

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Estimators

- *discretization estimator*

$$\eta_{\text{disc},K}^{k,i} := 2^{1/p} \left(\|\bar{\sigma}_h^{k,i} + \mathbf{d}_h^{k,i}\|_{q,K} + \left\{ \sum_{e \in \mathcal{E}_K} h_e^{1-q} \|[\![u_h^{k,i}]\!] \|_{q,e}^q \right\}^{\frac{1}{q}} \right)$$

- *linearization estimator*

$$\eta_{\text{lin},K}^{k,i} := \|\mathbf{l}_h^{k,i}\|_{q,K}$$

- *algebraic estimator*

$$\eta_{\text{alg},K}^{k,i} := \|\mathbf{a}_h^{k,i}\|_{q,K}$$

- *algebraic remainder estimator*

$$\eta_{\text{rem},K}^{k,i} := h_\Omega \|\rho_h^{k,i}\|_{q,K}$$

- *quadrature estimator*

$$\eta_{\text{quad},K}^{k,i} := \|\boldsymbol{\sigma}(u_h^{k,i}, \nabla u_h^{k,i}) - \bar{\sigma}_h^{k,i}\|_{q,K}$$

- *data oscillation estimator*

$$\eta_{\text{osc},K}^{k,i} := C_{\text{P},p} h_K \|f - f_h\|_{q,K}$$

- $\eta_{\cdot}^{k,i} := \left\{ \sum_{K \in \mathcal{T}_h} (\eta_{\cdot,K}^{k,i})^q \right\}^{1/q}$

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Stopping criteria

Global stopping criteria

- stop whenever:

$$\eta_{\text{rem}}^{k,i} \leq \gamma_{\text{rem}} \max\{\eta_{\text{disc}}^{k,i}, \eta_{\text{lin}}^{k,i}, \eta_{\text{alg}}^{k,i}\},$$

$$\eta_{\text{alg}}^{k,i} \leq \gamma_{\text{alg}} \max\{\eta_{\text{disc}}^{k,i}, \eta_{\text{lin}}^{k,i}\},$$

$$\eta_{\text{lin}}^{k,i} \leq \gamma_{\text{lin}} \eta_{\text{disc}}^{k,i}$$

- $\gamma_{\text{rem}}, \gamma_{\text{alg}}, \gamma_{\text{lin}} \approx 0.1$

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$$\eta_{\text{lin},K}^{k,i} \leq \gamma_{\text{lin},K} \eta_{\text{disc},K}^{k,i} \quad \forall K \in \mathcal{T}_h$$

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Assumption for efficiency

Assumption C (Approximation property)

For all $K \in \mathcal{T}_h$, there holds

$$\|\bar{\sigma}_h^{k,i} + \mathbf{d}_h^{k,i}\|_{q,K} \lesssim \eta_{\sharp,\mathfrak{T}_K}^{k,i} + \eta_{\text{osc},\mathfrak{T}_K}^{k,i},$$

where

$$\begin{aligned} \eta_{\sharp,\mathfrak{T}_K}^{k,i} := & \left\{ \sum_{K' \in \mathfrak{T}_K} h_{K'}^q \|f_h + \nabla \cdot \bar{\sigma}_h^{k,i}\|_{q,K'}^q + \sum_{e \in \mathcal{E}_K^{\text{int}}} h_e \|[\bar{\sigma}_h^{k,i} \cdot \mathbf{n}_e]\|_{q,e}^q \right. \\ & \left. + \sum_{e \in \mathcal{E}_K} h_e^{1-q} \|[\mathbf{u}_h^{k,i}]\|_{q,e}^q \right\}^{\frac{1}{q}}. \end{aligned}$$

Global efficiency

Theorem (Global efficiency)

Let the mesh T_h be shape-regular and let the **global stopping criteria** hold. Recall that $\mathcal{J}_u(u_h^{k,i}) \leq \eta^{k,i}$. Then, under Assumption C,

$$\eta^{k,i} \lesssim \mathcal{J}_u(u_h^{k,i}) + \eta_{\text{quad}}^{k,i} + \eta_{\text{osc}}^{k,i},$$

where \lesssim means up to a constant **independent** of σ and q .

- **robustness** with respect to the **nonlinearity** thanks to the choice of the **dual norm** as error measure

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Let the mesh \mathcal{T}_h be shape-regular and let the local stopping criteria hold. Then, under Assumption C,

$$\begin{aligned} & \eta_{\text{disc},K}^{k,i} + \eta_{\text{lin},K}^{k,i} + \eta_{\text{alg},K}^{k,i} + \eta_{\text{rem},K}^{k,i} \\ & \lesssim \mathcal{J}_{u,\mathfrak{T}_K}^{\text{up}}(u_h^{k,i}) + \eta_{\text{quad},\mathfrak{T}_K}^{k,i} + \eta_{\text{osc},\mathfrak{T}_K}^{k,i} \end{aligned}$$

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- robustness and local efficiency for an upper bound on the dual norm

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Algebraic error flux reconstruction and algebraic remainder

Construction of $\mathbf{a}_h^{k,i}$ and $\rho_h^{k,i}$

- On linearization step k and algebraic step i , we have

$$\mathbb{A}^k \mathbf{U}^{k,i} = \mathbf{F}^k - \mathbf{R}^{k,i}.$$

- Do ν additional steps of the algebraic solver, yielding

$$\mathbb{A}^k \mathbf{U}^{k,i+\nu} = \mathbf{F}^k - \mathbf{R}^{k,i+\nu}.$$

- Construct the function $\rho_h^{k,i}$ from the algebraic residual vector $\mathbf{R}^{k,i+\nu}$ (lifting into appropriate discrete space).
- Suppose we can obtain discretization and linearization flux reconstructions $\mathbf{d}_h^{k,i}, \mathbf{l}_h^{k,i}$ on each algebraic step. Then set

$$\mathbf{a}_h^{k,i} := (\mathbf{d}_h^{k,i+\nu} + \mathbf{l}_h^{k,i+\nu}) - (\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i}).$$

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1 Introduction

2 Adaptive inexact Newton method

- Quasi-linear elliptic problems
- Approximate solution and error measure
- A guaranteed a posteriori error estimate
- Stopping criteria and efficiency

3 Examples of application

- Arbitrary iterative algebraic solver
- **Nonconforming FEs, fixed point/Newton, p -Laplacian**
- DGs, fixed point/Newton, quasi-linear diffusion
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4 Numerical results

5 An extension to two-phase flows

6 Conclusions and future directions

Nonconforming finite elements for the p -Laplacian

Discretization

Find $\textcolor{orange}{u}_h \in V_h$ such that

$$(\sigma(\nabla u_h), \nabla v_h) = (f_h, v_h) \quad \forall v_h \in V_h.$$

- $\sigma(\nabla u_h) = |\nabla u_h|^{p-2} \nabla u_h$
- V_h the Crouzeix–Raviart space
- $f_h := \Pi_0 f$
- leads to the system of **nonlinear algebraic equations**

$$\mathcal{A}(U) = F$$

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Linearization

Linearization

Find $\textcolor{brown}{u}_h^k \in V_h$ such that

$$(\sigma^{k-1}(\nabla u_h^k), \nabla \psi_e) = (f_h, \psi_e) \quad \forall e \in \mathcal{E}_h^{\text{int}}.$$

- $u_h^0 \in V_h$ yields the initial vector U^0
- fixed-point linearization

$$\sigma^{k-1}(\xi) := |\nabla u_h^{k-1}|^{p-2} \xi$$

- Newton linearization

$$\begin{aligned} \sigma^{k-1}(\xi) &:= |\nabla u_h^{k-1}|^{p-2} \xi + (p-2) |\nabla u_h^{k-1}|^{p-4} \\ &\quad (\nabla u_h^{k-1} \otimes \nabla u_h^{k-1})(\xi - \nabla u_h^{k-1}) \end{aligned}$$

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Algebraic solution

Algebraic solution

Find $\textcolor{orange}{u}_h^{k,i} \in V_h$ such that

$$(\sigma^{k-1}(\nabla u_h^{k,i}), \nabla \psi_e) = (f_h, \psi_e) - R_e^{k,i} \quad \forall e \in \mathcal{E}_h^{\text{int}}.$$

- algebraic residual vector $R^{k,i} = \{R_e^{k,i}\}_{e \in \mathcal{E}_h^{\text{int}}}$
- discrete system

$$\mathbb{A}^k U^k = F^k - R^{k,i}$$

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Flux reconstructions

Definition (Construction of $(\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i})$)

For all $K \in \mathcal{T}_h$,

$$(\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i})|_K := -\sigma^{\mathbf{k}-1}(\nabla u_h^{k,i})|_K + \frac{f_h|_K}{d}(\mathbf{x} - \mathbf{x}_K) - \sum_{e \in \mathcal{E}_K} \frac{\mathbf{R}_e^{k,i}}{d|D_e|}(\mathbf{x} - \mathbf{x}_K)|_{K_e}.$$

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where $\bar{\mathbf{R}}_e^{k,i} := (f_h, \psi_e) - (\sigma(\nabla u_h^{k,i}), \nabla \psi_e)$ $\forall e \in \mathcal{E}_h^{\text{int}}$.

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Set $\bar{\sigma}_h^{k,i} := \sigma(\nabla u_h^{k,i})$. Consequently, $\eta_{\text{quad}, K}^{k,i} = 0$ for all $K \in \mathcal{T}_h$.

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Verification of the assumptions – upper bound

Lemma (Assumptions A and B)

Assumptions A and B hold.

Comments

- $\|\mathbf{a}_h^{k,i}\|_{q,K} \rightarrow 0$ as the linear solver converges by definition.
- $\|\mathbf{l}_h^{k,i}\|_{q,K} \rightarrow 0$ as the nonlinear solver converges by the construction of $\mathbf{l}_h^{k,i}$.
- Both $(\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i})$ and $\mathbf{d}_h^{k,i}$ belong to $\mathbf{RTN}_0(\mathcal{S}_h) \Rightarrow \mathbf{a}_h^{k,i} \in \mathbf{RTN}_0(\mathcal{S}_h)$ and $\mathbf{t}_h^{k,i} \in \mathbf{RTN}_0(\mathcal{S}_h)$.

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Verification of the assumptions – efficiency

Lemma (Assumption C)

Assumption C holds.

Comments

- $\mathbf{d}_h^{k,i}$ close to $\sigma(\nabla u_h^{k,i})$
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Verification of the assumptions – efficiency

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Discontinuous Galerkin for the quasi-linear diffusion

Discretization

Find $\textcolor{red}{u}_h \in V_h := \mathbb{P}_m(\mathcal{T}_h)$, $m \geq 1$, such that, for all $v_h \in V_h$,

$$\begin{aligned} & (\sigma(u_h, \nabla u_h), \nabla v_h) - \sum_{e \in \mathcal{E}_h} \left\{ \langle \{\!\{ \sigma(u_h, \nabla u_h) \}\!\} \cdot \mathbf{n}_e, [\![v_h]\!] \rangle_e \right. \\ & \left. + \theta \langle \{\!\{ \underline{\mathbf{A}}(u_h) \nabla v_h \}\!\} \cdot \mathbf{n}_e, [\![u_h]\!] \rangle_e \right\} + \sum_{e \in \mathcal{E}_h} \langle \bar{\alpha}_e h_e^{-1} [\![u_h]\!], [\![v_h]\!] \rangle_e = (f, v_h). \end{aligned}$$

- $\theta \in \{-1, 0, 1\}$
- $\bar{\alpha}_e := \|\underline{\mathbf{A}}\|_{L^\infty(\mathbb{R})} \chi_e$, χ_e large enough
- leads to the system of **nonlinear algebraic equations**

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Linearization

Find $\textcolor{red}{u_h^k} \in V_h$ such that, for all $K \in \mathcal{T}_h$ and all $j \in \mathcal{C}_K := \{1, \dots, \dim(\mathbb{P}_m(K))\}$,

$$\begin{aligned} & (\sigma^{k-1}(u_h^k, \nabla u_h^k), \nabla \psi_{K,j}) - \sum_{e \in \mathcal{E}_h} \{ \langle \{\sigma^{k-1}(u_h^k, \nabla u_h^k)\} \cdot \mathbf{n}_e, [\![\psi_{K,j}]\!] \rangle_e \\ & + \theta \langle \{\underline{\mathbf{A}}^{k-1}(u_h^k) \nabla \psi_{K,j}\} \cdot \mathbf{n}_e, [u_h^k] \rangle_e \} + \sum_{e \in \mathcal{E}_h} \langle \bar{\alpha}_e h_e^{-1} [u_h^k], [\![\psi_{K,j}]\!] \rangle_e = (f, \psi_{K,j}). \end{aligned}$$

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$$\begin{aligned} \sigma^{k-1}(v, \xi) &:= \underline{\mathbf{A}}(u_h^{k-1})\xi + (v - u_h^{k-1}) \partial_v \underline{\mathbf{A}}(u_h^{k-1}) \nabla u_h^{k-1}, \\ \underline{\mathbf{A}}^{k-1}(v) &:= \underline{\mathbf{A}}(u_h^{k-1}) + \partial_v \underline{\mathbf{A}}(u_h^{k-1})(v - u_h^{k-1}) \end{aligned}$$

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- algebraic residual vector $R^{k,i} = \{R_{K,j}^{k,i}\}_{K \in \mathcal{T}_h, j \in \mathcal{C}_K}$
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Flux reconstructions

Definition (Construction of $(\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i}) \in \mathbf{RTN}_l(\mathcal{T}_h)$, $l := m-1/m$)

For all $K \in \mathcal{T}_h$ and all $e \in \mathcal{E}_K$,

$$\langle (\mathbf{d}_h^{k,i} + \mathbf{l}_h^{k,i}) \cdot \mathbf{n}_e, q_h \rangle_e := \langle -\{\sigma^{k-1}(u_h^{k,i}, \nabla u_h^{k,i})\} \cdot \mathbf{n}_e + \bar{\alpha}_e h_e^{-1} \llbracket u_h^{k,i} \rrbracket, q_h \rangle_e,$$

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for all $q_h \in \mathbb{P}_l(e)$ and all $\mathbf{r}_h \in [\mathbb{P}_{l-1}(K)]^d$.

Definition (Construction of $\mathbf{d}_h^{k,i} \in \mathbf{RTN}_l(\mathcal{T}_h)$, $l := m-1$ or $l := m$)

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Verification of the assumptions – upper bound

Definition (Construction of f_h , $\bar{\sigma}_h^{k,i}$)

Set $f_h := \Pi_I f$ and $\bar{\sigma}_h^{k,i} := \mathbf{I}_I^{\text{RTN}}(\sigma(u_h^{k,i}, \nabla u_h^{k,i}))$.

Lemma (Assumptions A and B)

Assumptions A and B hold.

Comments

- $\|\mathbf{a}_h^{k,i}\|_{q,K} \rightarrow 0$ as the linear solver converges by definition.
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- Both $(\mathbf{d}_h^{k,i} + \mathbf{I}_h^{k,i})$ and $\mathbf{d}_h^{k,i}$ belong to $\mathbf{RTN}_I(\mathcal{T}_h) \Rightarrow \mathbf{a}_h^{k,i} \in \mathbf{RTN}_I(\mathcal{T}_h)$ and $\mathbf{t}_h^{k,i} \in \mathbf{RTN}_I(\mathcal{T}_h)$.

Verification of the assumptions – efficiency

Lemma (Assumption C)

Assumption C holds.

Comments

- $\mathbf{d}_h^{k,i}$ close to $\overline{\sigma}_h^{k,i}$
- approximation properties of Raviart–Thomas–Nédélec spaces

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Summary

Discretization methods

- nonconforming finite elements
- discontinuous Galerkin
- finite elements
- various finite volumes
- mixed finite elements

Linearizations

- fixed point
- Newton

Linear solvers

- independent of the linear solver
- ... all Assumptions A to C verified

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Numerical experiment I

Model problem

- p -Laplacian

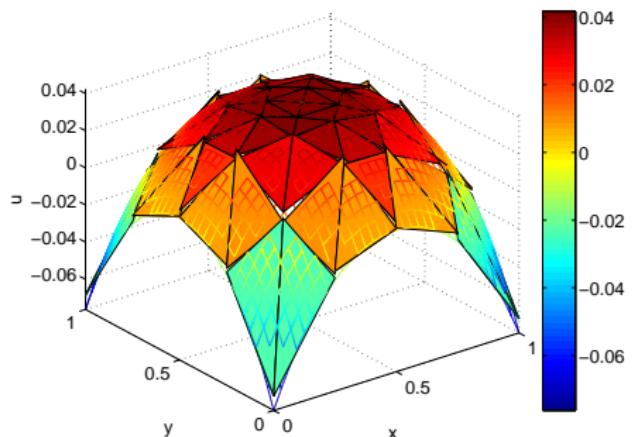
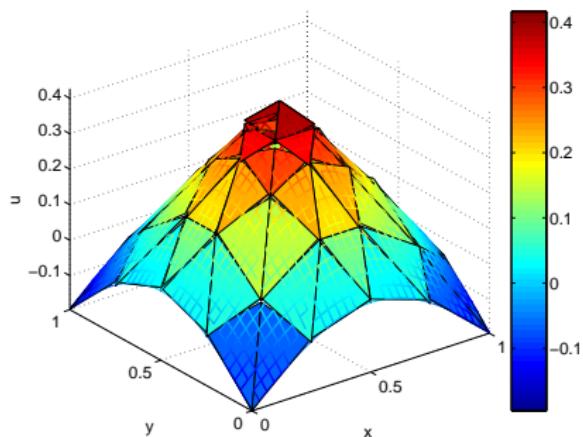
$$\begin{aligned}\nabla \cdot (|\nabla u|^{p-2} \nabla u) &= f \quad \text{in } \Omega, \\ u &= u_0 \quad \text{on } \partial\Omega\end{aligned}$$

- weak solution (used to impose the Dirichlet BC)

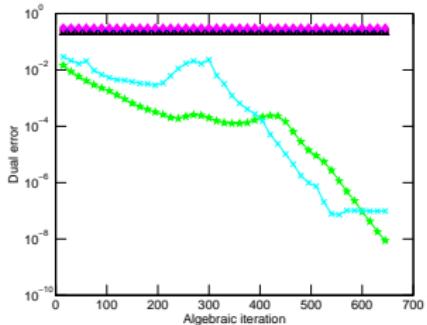
$$u(x, y) = -\frac{p-1}{p} \left((x - \frac{1}{2})^2 + (y - \frac{1}{2})^2 \right)^{\frac{p}{2(p-1)}} + \frac{p-1}{p} \left(\frac{1}{2} \right)^{\frac{p}{p-1}}$$

- tested values $p = 1.5$ and 10
- nonconforming finite elements

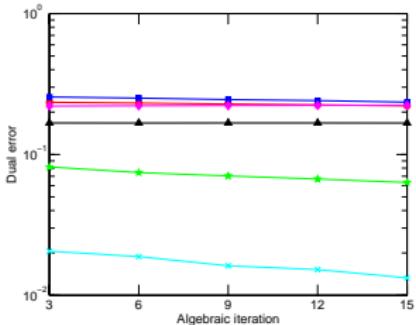
Analytical and approximate solutions

Case $p = 1.5$ Case $p = 10$

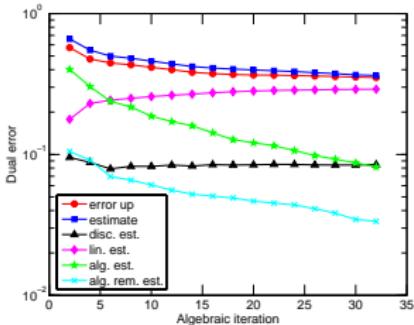
Error and estimators as a function of CG iterations, $p = 10$, 6th level mesh, 6th Newton step.



Newton

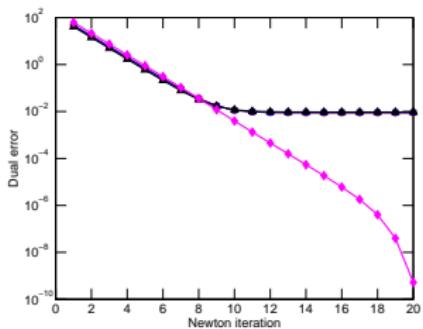


inexact Newton

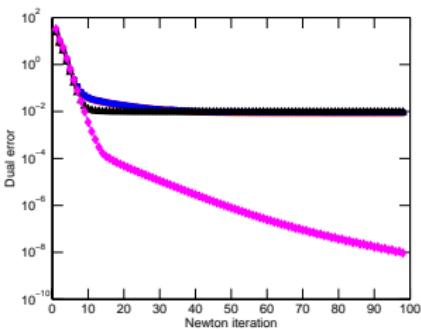


ad. inexact Newton

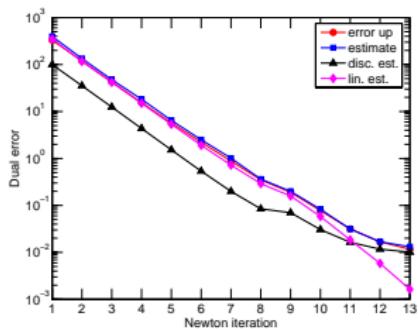
Error and estimators as a function of Newton iterations, $p = 10$, 6th level mesh



Newton

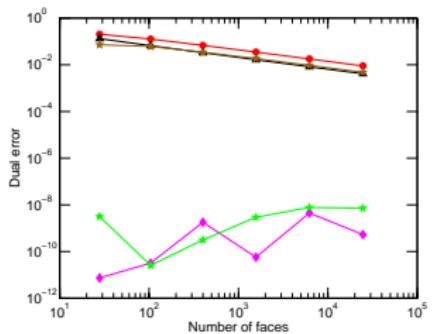


inexact Newton

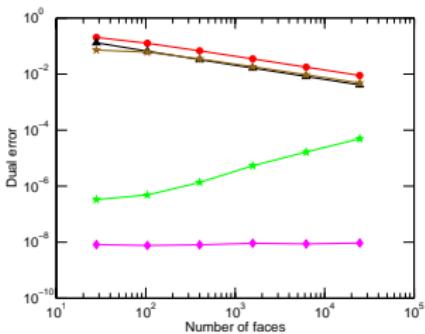


ad. inexact Newton

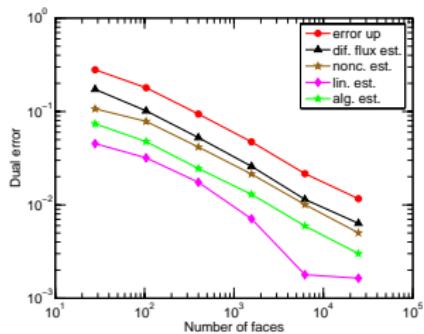
Error and estimators, $p = 10$



Newton

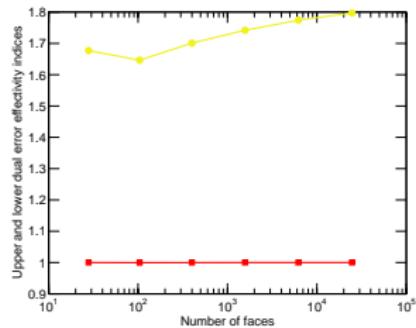


inexact Newton

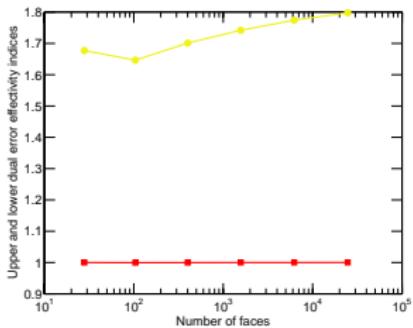


ad. inexact Newton

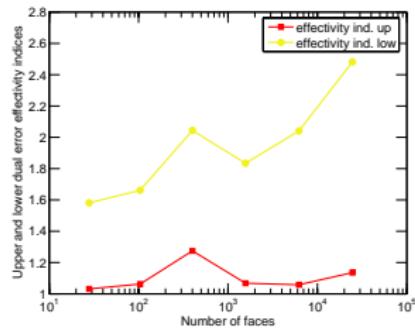
Effectivity indices, $p = 10$



Newton

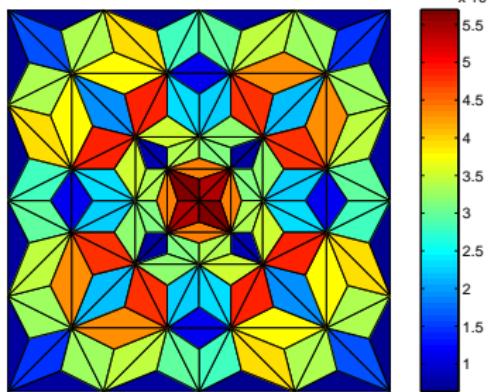


inexact Newton

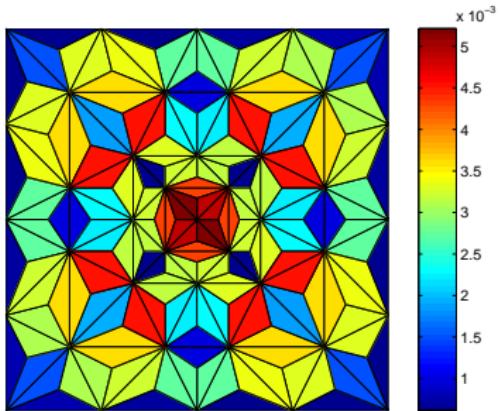


ad. inexact Newton

Error distribution, $p = 10$

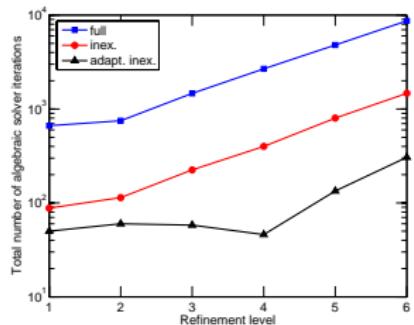
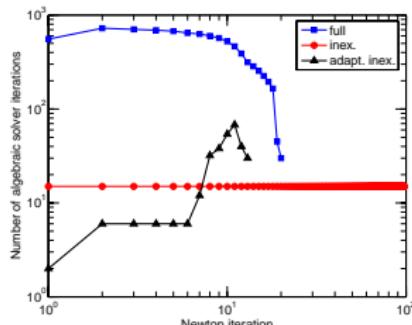
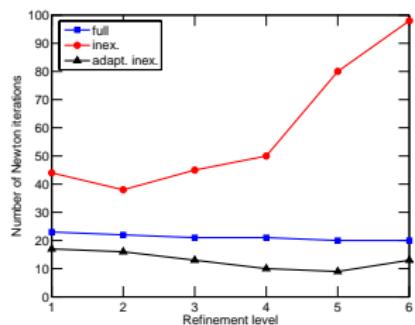


Estimated error distribution



Exact error distribution

Newton and algebraic iterations, $p = 10$

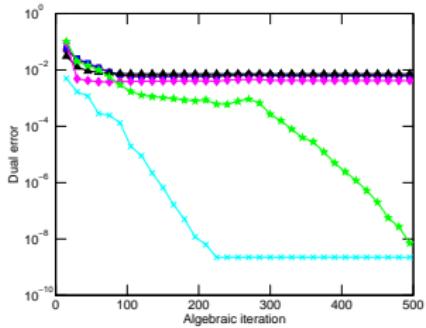


Newton it. / refinement

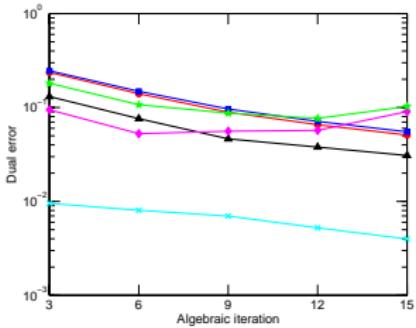
alg. it. / Newton step

alg. it. / refinement

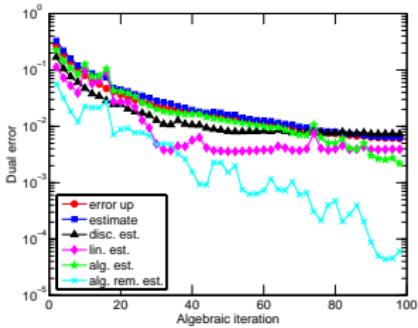
Error and estimators as a function of CG iterations, $p = 1.5$, 6th level mesh, 1st Newton step.



Newton

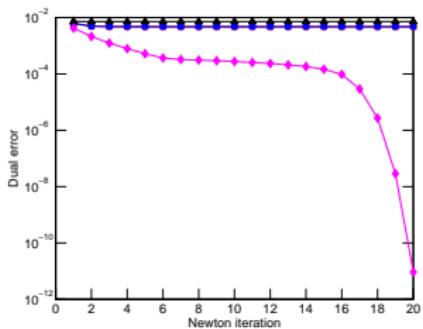


inexact Newton

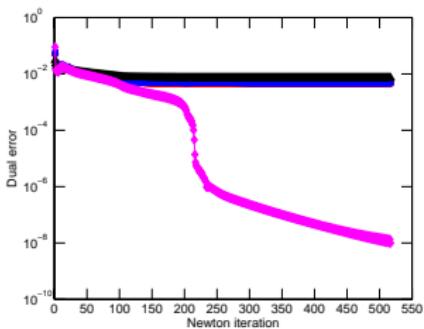


ad. inexact Newton

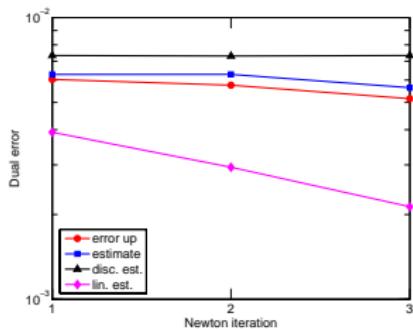
Error and estimators as a function of Newton iterations, $p = 1.5$, 6th level mesh



Newton

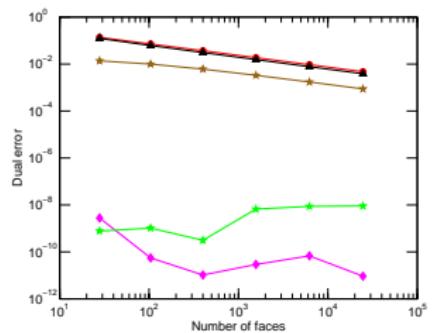


inexact Newton

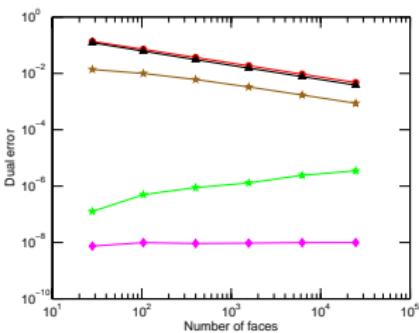


ad. inexact Newton

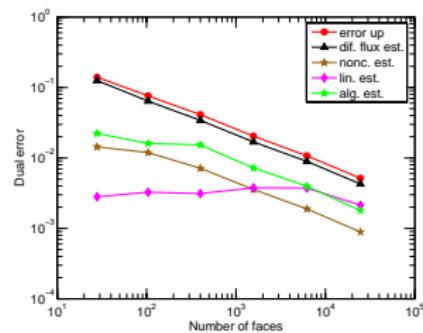
Error and estimators, $p = 1.5$



Newton

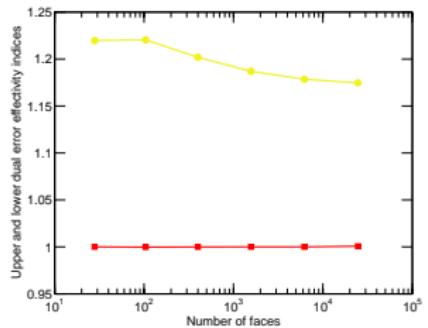


inexact Newton

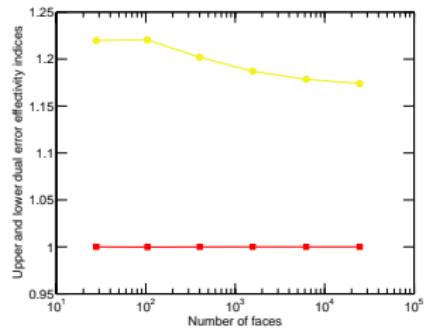


ad. inexact Newton

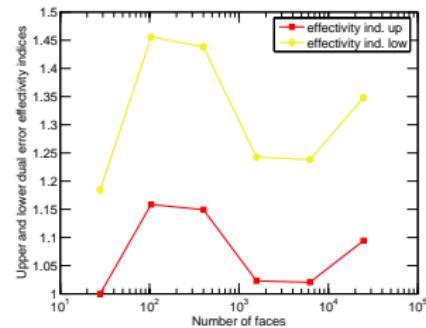
Effectivity indices, $p = 1.5$



Newton

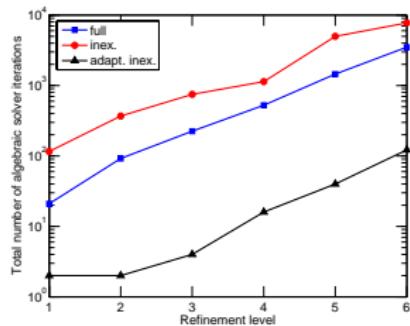
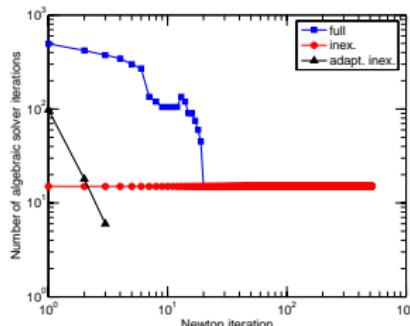
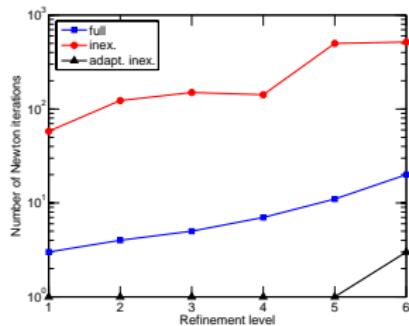


inexact Newton



ad. inexact Newton

Newton and algebraic iterations, $p = 1.5$



Newton it. / refinement

alg. it. / Newton step

alg. it. / refinement

Numerical experiment II

Model problem

- p -Laplacian

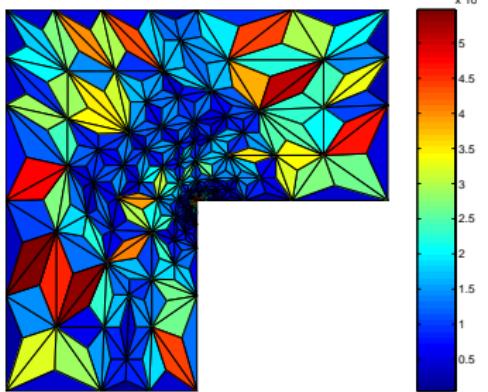
$$\begin{aligned} \nabla \cdot (|\nabla u|^{p-2} \nabla u) &= f && \text{in } \Omega, \\ u &= u_0 && \text{on } \partial\Omega \end{aligned}$$

- weak solution (used to impose the Dirichlet BC)

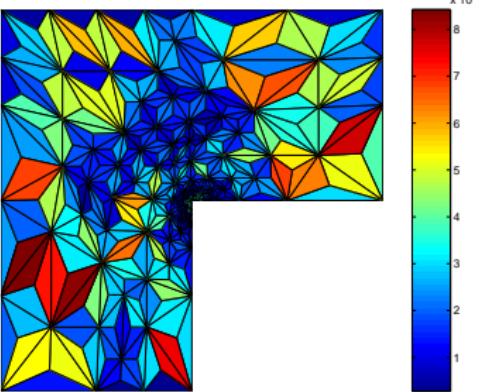
$$u(r, \theta) = r^{\frac{7}{8}} \sin(\theta \frac{7}{8})$$

- $p = 4$, L-shape domain, singularity in the origin
(Carstensen and Klose (2003))
- nonconforming finite elements

Error distribution on an adaptively refined mesh

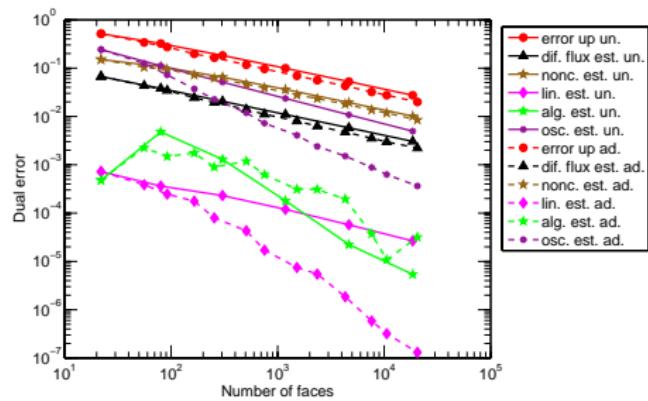


Estimated error distribution

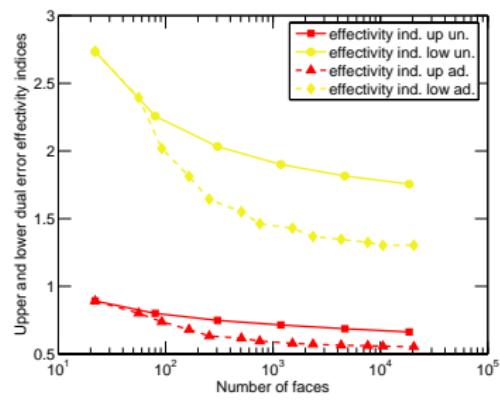


Exact error distribution

Estimated and actual errors and the effectivity index



Estimated and actual errors



Effectivity index

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Two-phase flow in porous media

Two-phase flow in porous media

$$\partial_t(\phi s_\alpha) + \nabla \cdot \mathbf{u}_\alpha = q_\alpha, \quad \alpha \in \{n, w\},$$

$$-\frac{k_{r,\alpha}(s_w)}{\mu_\alpha} \mathbf{K}(\nabla p_\alpha + \rho_\alpha g \nabla z) = \mathbf{u}_\alpha, \quad \alpha \in \{n, w\},$$

$$s_n + s_w = 1,$$

$$p_n - p_w = p_c(s_w)$$

Mathematical issues

- coupled system
- unsteady, nonlinear
- elliptic–parabolic degenerate type
- dominant advection

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Two-phase flow in porous media

Theorem (A posteriori error estimate distinguishing the error components)

Let

- n be the *time step*,
- k be the *linearization step*,
- i be the *algebraic solver step*,

with the approximations $(s_{\alpha,h\tau}^{k,i}, p_{\alpha,h\tau}^{k,i})$. Then

$$\| (s_{\alpha} - s_{\alpha,h\tau}^{k,i}, p_{\alpha} - p_{\alpha,h\tau}^{k,i}) \| \|_{I_n} \leq \eta_{\text{sp},\alpha}^{n,k,i} + \eta_{\text{tm},\alpha}^{n,k,i} + \eta_{\text{lin},\alpha}^{n,k,i} + \eta_{\text{alg},\alpha}^{n,k,i}.$$

Error components

- $\eta_{\text{sp},\alpha}^{n,k,i}$: spatial discretization
- $\eta_{\text{tm},\alpha}^{n,k,i}$: temporal discretization
- $\eta_{\text{lin},\alpha}^{n,k,i}$: linearization
- $\eta_{\text{alg},\alpha}^{n,k,i}$: algebraic solver

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Theorem (A posteriori error estimate distinguishing the error components)

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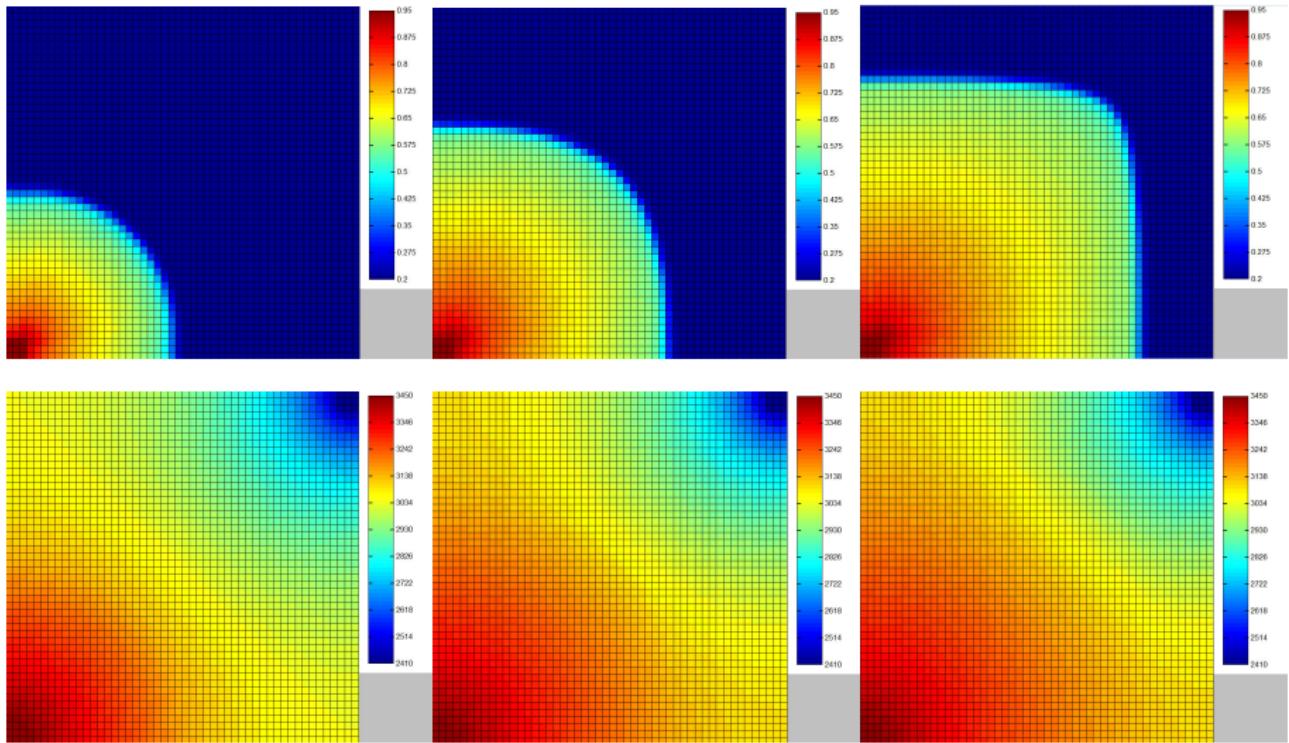
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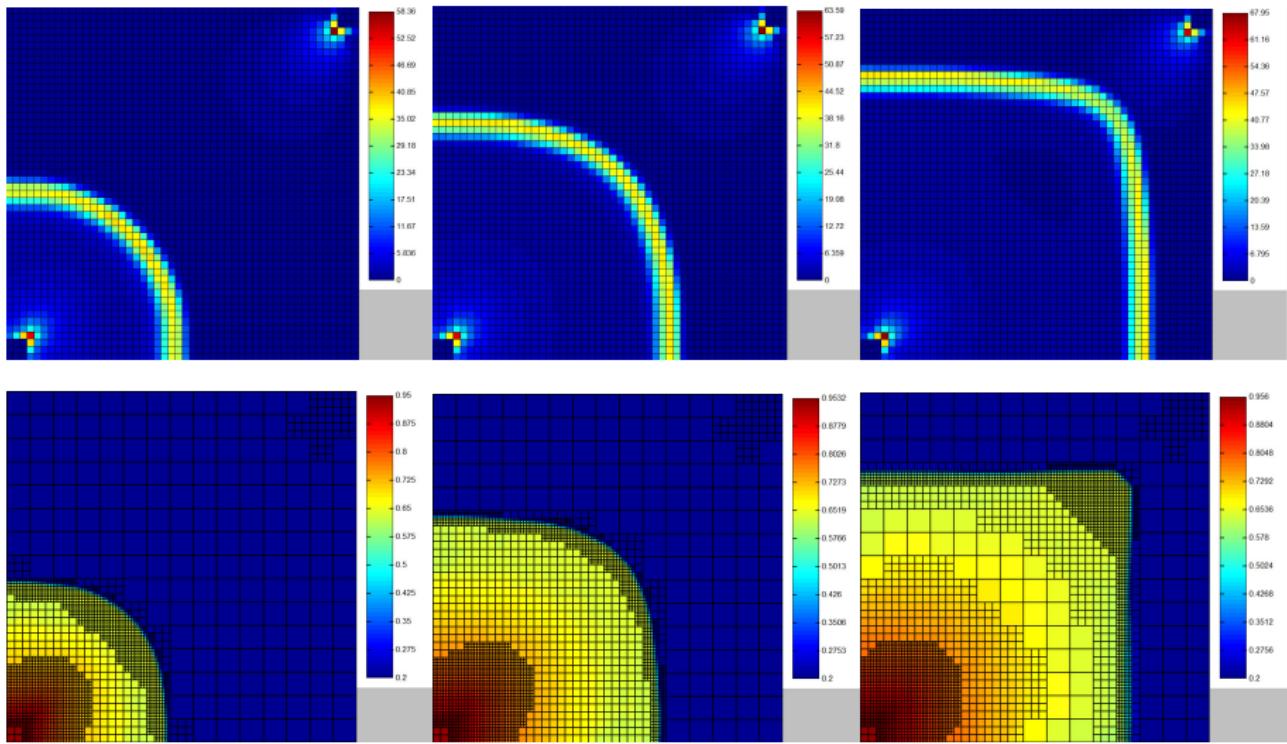
Error components

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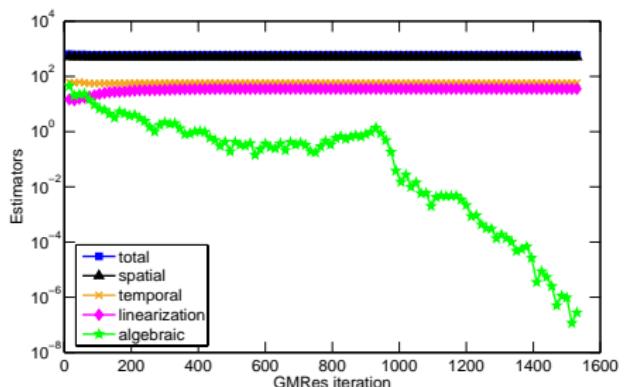
Water saturation/water pressure evolution



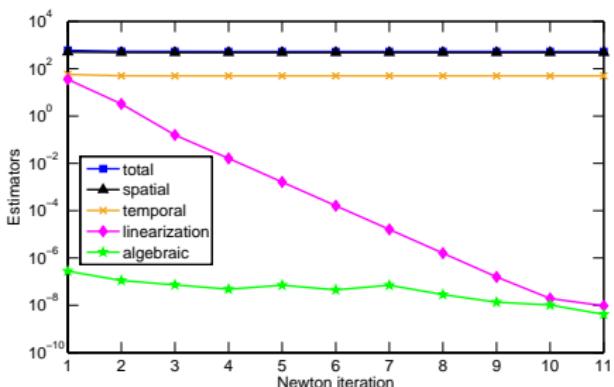
Estimators/meshes evolution



Estimators and stopping criteria

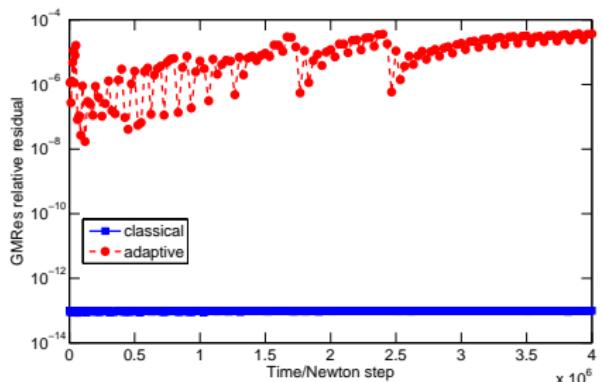


Estimators in function of
GMRes iterations

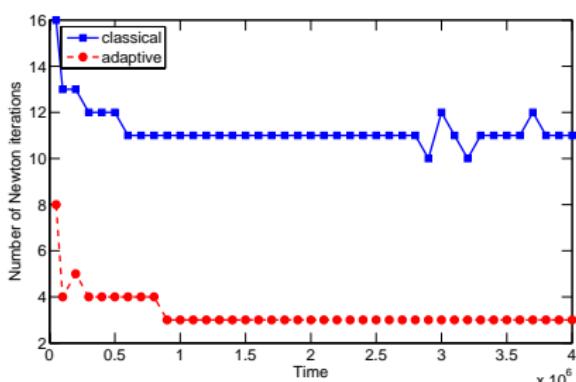


Estimators in function of
Newton iterations

GMRes relative residual/Newton iterations

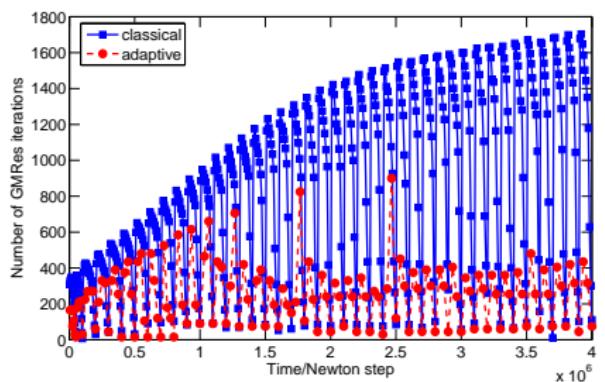


GMRes relative residual

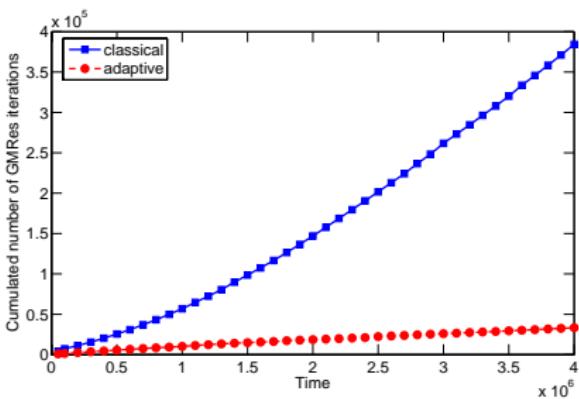


Newton iterations

GMRes iterations



Per time and Newton step



Cumulated

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- only a **necessary number** of **linearization iterations**
- **“smart online decisions”**: algebraic step / linearization step / space mesh refinement
- important **computational savings**
- guaranteed and robust upper bound via **a posteriori error estimates**

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- other coupled nonlinear systems
- convergence and optimality

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Bibliography

- ERN A., VOHRALÍK M., Adaptive inexact Newton methods for discretizations of nonlinear diffusion PDEs. I. General theory and a posteriori stopping criteria, HAL Preprint 00681422 & Adaptive inexact Newton methods for discretizations of nonlinear diffusion PDEs. II. Applications, HAL Preprint 00681426.
- JIRÁNEK P., STRAKOŠ Z., VOHRALÍK M., A posteriori error estimates including algebraic error and stopping criteria for iterative solvers. *SIAM J. Sci. Comput.* **32** (2010), 1567–1590.
- EL ALAOUI L., ERN A., VOHRALÍK M., Guaranteed and robust a posteriori error estimates and balancing discretization and linearization errors for monotone nonlinear problems. *Comput. Methods Appl. Mech. Engrg.* **200** (2011), 2782–2795.

Thank you for your attention!