

# Reliable, efficient, and robust a posteriori estimates for nonlinear elliptic problems

An orthogonal decomposition result based on iterative linearization

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- 1 Introduction
- 2 Main analytical results
- 3 Scope of the results
- 4 Numerical results

# 1 Outline

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- 2 Main analytical results
- 3 Scope of the results
- 4 Numerical results

## Nonlinear elliptic problems

For  $d \in \mathbb{N}$ , let  $\Omega \subset \mathbb{R}^d$  be an open and bounded polytope. Let  $u \in H_0^1(\Omega)$  solve the **nonlinear** elliptic operator equation: for  $\mathcal{R} : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$ ,

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Assumption 1  $\mathcal{R}$  is monotone & Lipschitz\*

For a numerical approximation  $u_\ell \in H_0^1(\Omega)$ , and constants  $\lambda_M > \lambda_m > 0$ ,

$$\lambda_m \operatorname{dist}(u_\ell, u) \leq \sup_{\varphi \in H_0^1(\Omega)} \frac{\langle \mathcal{R}(u_\ell) - \mathcal{R}(u), \varphi \rangle}{\|\nabla \varphi\|} \leq \lambda_M \operatorname{dist}(u_\ell, u).$$

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Then the estimate [Chaillou & Suri (2006), Kim (2007), Houston *et al* (2008), Garau *et al* (2011),...],

$$\lambda_m \operatorname{dist}(u_\ell, u) \leq \eta(u_\ell) \leq C \lambda_M \operatorname{dist}(u_\ell, u)$$

is not robust with respect to  $\lambda_M/\lambda_m$

## 1 Dual norm of the residual estimate

**Reliable**, and **locally efficient** a posteriori error estimates **robust** with respect to the strength of the **nonlinearity**  $\lambda_M/\lambda_m$

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- ▶ The dual norm of the residual might be too weak an error measure

## 1 A linear example

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Consider the diffusion eq:  $\langle \mathcal{R}(u), \varphi \rangle := (f, \varphi) - (\mathcal{D}\nabla u, \nabla \varphi) = 0$ .

Let  $\lambda_m |\mathbf{y}|^2 \leq \mathbf{y}^T \mathcal{D} \mathbf{y} \leq \lambda_M |\mathbf{y}|^2$ , for all  $\mathbf{y} \in \mathbb{R}^d$ .

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$$\|\nabla(u - u_\ell)\| \leq \frac{\lambda_M}{\lambda_m} \|\nabla(u - \varphi_\ell)\| \quad \forall \varphi_\ell \in V_\ell.$$

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However, defining the energy norm  $\|\|\varphi\|\|_{1,\mathcal{D}} = \|\mathcal{D}^{\frac{1}{2}}\nabla\varphi\|$  one has

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$$\|u - u_\ell\|_{1,\mathcal{D}} \leq \|u - \varphi_\ell\|_{1,\mathcal{D}}, \quad \forall \varphi_\ell \in V_\ell.$$

This motivates rather the error measure

$$\|\mathcal{R}(u_\ell)\|_{-1,\mathcal{D}} := \sup_{\varphi \in H_0^1(\Omega)} \frac{\langle \mathcal{R}(u_\ell), \varphi \rangle}{\|\varphi\|_{1,\mathcal{D}}} = \|u - u_\ell\|_{1,\mathcal{D}}$$

which also results in robust estimates

## 1 Moving to the nonlinear case

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Example (nonlinear diffusion):  $\langle \mathcal{R}(u), \varphi \rangle := (f, \varphi) - (\mathcal{D}(u)\nabla u, \nabla\varphi) = 0.$

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Then  $\|\mathcal{R}(\cdot)\|_{-1, \mathcal{D}(u)}$  cannot be defined since  $u \in H_0^1(\Omega)$  is unknown.

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## Linearization iterations

We generally solve nonlinear equations by linearization iterations, i.e., by finding a sequence  $\{u_\ell^i\}_{i \in \mathbb{N}} \subset V_\ell \subset H_0^1(\Omega)$ .

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**Example (Fixed point iteration)** For each  $i \in \mathbb{N}$  and  $u_\ell^i \in V_\ell$ , let  $u_\ell^{i+1} \in V_\ell$  solve  $(\mathcal{D}(u_\ell^i)\nabla u_\ell^{i+1}, \nabla \varphi_\ell) = (f, \varphi_\ell)$  for all  $\varphi_\ell \in V_\ell.$

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This is the FE approximation of  $u_{\langle \ell \rangle}^{i+1} \in H_0^1(\Omega)$  solving the linear problem

$$\langle \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_{\langle \ell \rangle}^{i+1}), \varphi \rangle := (f, \varphi) - (\mathcal{D}(u_\ell^i)\nabla u_{\langle \ell \rangle}^{i+1}, \nabla \varphi) = 0 \quad \forall \varphi \in H_0^1(\Omega).$$

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Then defining the **iteration-dependent energy norm**

$$\begin{cases} \|\varphi\|_{1, u_\ell^i} := \|\mathcal{D}(u_\ell^i)^{\frac{1}{2}} \nabla \varphi\| & \text{for } \varphi \in H_0^1(\Omega), \\ \|\varsigma\|_{-1, u_\ell^i} = \sup_{\varphi \in H_0^1(\Omega)} \langle \varsigma, \varphi \rangle / \|\varphi\|_{1, u_\ell^i} & \text{for } \varsigma \in H^{-1}(\Omega), \end{cases}$$

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we have (under conditions) robust estimates of

$$\left\| \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}) \right\|_{-1, u_\ell^i} = \left\| u_{\langle \ell \rangle}^{i+1} - u_\ell^{i+1} \right\|_{1, u_\ell^i}$$

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Noting that

$$\langle \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}), \varphi \rangle := -(\mathcal{D}(u_\ell^i)\nabla(u_\ell^{i+1} - u_\ell^i), \nabla \varphi) + \langle \mathcal{R}(u_\ell^i), \varphi \rangle$$

can we provide a robust estimate for  $\|\mathcal{R}(u_\ell^i)\|_{-1, u_\ell^i}?$

- 1 Introduction
- 2 Main analytical results
  - Decomposition of error
  - A posteriori error estimates
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## 2 An orthogonal decomposition result

### Theorem 1 Decomposition of the total error

Under Assumption 1, provided that the linearization iterations  $\{u_\ell^i\}_{i \in \mathbb{N}} \subset V_\ell$  are generated by FE approximations of  $u_{\langle \ell \rangle}^i \in H_0^1(\Omega)$  solving

$$\langle \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_{\langle \ell \rangle}^{i+1}), \varphi \rangle := -\mathcal{L}(u_\ell^i; u_{\langle \ell \rangle}^{i+1} - u_\ell^i, \varphi) + \langle \mathcal{R}(u_\ell^i), \varphi \rangle = 0 \quad \forall \varphi \in H_0^1(\Omega)$$

and  $i \geq 0$ , for a symmetric, bounded, coercive, bilinear form  $\mathcal{L}(u_\ell^i, \cdot, \cdot)$ ,

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**Remark** We would consider  $\mathcal{L} : H_0^1(\Omega) \times H_0^1(\Omega) \mapsto \mathbb{R}$  corresponding to **linear reaction-diffusion** problems, i.e.,

$$\mathcal{L}(u_\ell^i; v, w) := \underbrace{(L(\mathbf{x}, u_\ell^i) v, w)}_{\text{known reaction coeff.}} + \underbrace{(\mathfrak{a}(\mathbf{x}, u_\ell^i) \nabla v, \nabla w)}_{\text{known diffusion coeff.}}$$

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we have

$$\underbrace{\|\mathcal{R}(u_\ell^i)\|_{-1, u_\ell^i}^2}_{\text{total error}} = \underbrace{\|\mathcal{R}_{\text{lin}}^{u_\ell^i}(u_{\langle \ell \rangle}^{i+1})\|_{-1, u_\ell^i}^2}_{\text{discretization error of the linearization step}} + \underbrace{\|u_\ell^{i+1} - u_\ell^i\|_{1, u_\ell^i}^2}_{\text{linearization error}}.$$

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**Proof:** Since  $u_\ell^{i+1} - u_\ell^i \in V_\ell$ ,

$$\begin{aligned} \left\| \mathcal{R}(u_\ell^i) \right\|_{-1, u_\ell^i}^2 &= \left\| u_\ell^i - u_{\langle \ell \rangle}^{i+1} \right\|_{1, u_\ell^i}^2 = \left\| (u_\ell^i - u_\ell^{i+1}) + (u_\ell^{i+1} - u_{\langle \ell \rangle}^{i+1}) \right\|_{1, u_\ell^i}^2 \\ &= \left\| u_{\langle \ell \rangle}^{i+1} - u_\ell^{i+1} \right\|_{1, u_\ell^i}^2 + \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}^2 + \underbrace{2 \mathfrak{L}(u_\ell^i; u_{\langle \ell \rangle}^{i+1} - u_\ell^{i+1}, u_\ell^{i+1} - u_\ell^i)}_{=0, \text{ due to Galerkin orthogonality}} \\ &= \left\| \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}) \right\|_{-1, u_\ell^i}^2 + \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}^2. \end{aligned}$$

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- The linearization error is computed directly, we define

$$\eta_{\text{lin}, \Omega}^i := \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}.$$

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- The linearization error is computed directly, we define

$$\eta_{\text{lin}, \Omega}^i := \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}.$$

- For estimating  $\left\| \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}) \right\|_{-1, u_\ell^i}$  we introduce  $\eta_{\text{disc}, \Omega}^i$ , following the analysis on robust estimates of **singularly perturbed reaction-diffusion problems** in [Verfürth (1998)], [Ainsworth & Vejchodský (2011, 2014)] [Smears & Vohralík (2020)]

Theorem 2 Reliable, efficient, and robust a posteriori estimates

Global reliability

$$\|\mathcal{R}(u_\ell^i)\|_{-1, u_\ell^i}^2 \leq [\eta_\Omega^i]^2 := \sum_{K \in \mathcal{T}_\ell} ([\eta_{\text{disc}, K}^i]^2 + [\eta_{\text{lin}, K}^i]^2).$$

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$$[\eta_\Omega^i]^2 \lesssim \|\mathcal{R}(u_\ell^i)\|_{-1, u_\ell^i}^2 + (\text{data oscillation terms}).$$

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Local efficiency

For  $\omega \subset \Omega$ , there exists a neighbourhood  $\mathfrak{T}_\omega \subseteq \Omega$  such that

$$[\eta_\omega^i]^2 \lesssim \|\mathcal{R}(u_\ell^{i+1})\|_{-1, u_\ell^i, \mathfrak{T}_\omega}^2 + [\eta_{\text{lin}, \mathfrak{T}_\omega}^i]^2 + (\text{data oscillation terms}).$$

- 1 Introduction
- 2 Main analytical results
- 3 Scope of the results**
  - Gradient-dependent diffusivity
  - Gradient-independent diffusivity
- 4 Numerical results

### 3 Class of problems

#### Class 1: gradient-dependent diffusivity problems

For all  $\varphi \in H_0^1(\Omega)$ ,  $\mathcal{R} : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$  is defined as

$$\langle \mathcal{R}(u), \varphi \rangle := \langle f(\mathbf{x}, u), \varphi \rangle - (\boldsymbol{\sigma}(\mathbf{x}, \nabla u), \nabla \varphi)$$

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Assumption 1 is satisfied if  $f(\mathbf{x}, \cdot)$ ,  $\boldsymbol{\sigma}(\mathbf{x}, \cdot)$  are monotone and Lipschitz

$$(\boldsymbol{\sigma}(\mathbf{x}, \mathbf{y}) - \boldsymbol{\sigma}(\mathbf{x}, \mathbf{z})) \cdot (\mathbf{y} - \mathbf{z}) \geq \lambda_m |\mathbf{y} - \mathbf{z}|^2 \quad \text{for } \mathbf{x} \in \Omega \text{ and } \mathbf{y}, \mathbf{z} \in \mathbb{R}^d,$$

$$|\boldsymbol{\sigma}(\mathbf{x}, \mathbf{y}) - \boldsymbol{\sigma}(\mathbf{x}, \mathbf{z})| \leq \lambda_M |\mathbf{y} - \mathbf{z}| \quad \text{for } \mathbf{x} \in \Omega \text{ and } \mathbf{y}, \mathbf{z} \in \mathbb{R}^d.$$

with

$$\text{dist}(u, v) = \|\nabla(u - v)\|$$

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with

$$\text{dist}(u, v) = \|\nabla(u - v)\|$$

**Example** (Mean curvature flow) For  $a(\cdot)$  satisfying ellipticity condition

and  $b(\cdot) > 0$ :  $\boldsymbol{\sigma}(\mathbf{x}, \mathbf{y}) = a(\mathbf{x}) + \frac{b(\mathbf{x})\mathbf{y}}{(1+|\mathbf{y}|^2)^{\frac{1}{2}}}$

### 3 Linearization schemes: practical examples

#### Linearization operator

Considering the linearization operator

$$\mathfrak{L}(u_\ell^i; v, w) := (L(\mathbf{x}, u_\ell^i) v, w) + (\alpha(\mathbf{x}, u_\ell^i) \nabla v, \nabla w),$$

the coefficient functions for commonly used linearization schemes are

Scheme	$L(\mathbf{x}, v)$	$\alpha(\mathbf{x}, v)/\tau$
Kačanov (fixed point)	$\partial_\xi f(\mathbf{x}, v)$	$A(\mathbf{x},  \nabla v )$
Zarantonello	0	$\Lambda$ (constant) $> 0$

#### Class 2: gradient-independent diffusivity problems

For all  $\varphi \in H_0^1(\Omega)$ ,  $\mathcal{R} : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$  is defined as

$$\langle \mathcal{R}(u), \varphi \rangle := \langle f(\mathbf{x}, u), \varphi \rangle - \tau(\bar{\mathbf{K}}(\mathbf{x})(\mathcal{D}(\mathbf{x}, u)\nabla u + \mathbf{q}(\mathbf{x}, u)), \nabla\varphi)$$

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Assumption 1 is satisfied if  $\tau > 0$  is small and

- ▶  $\mathcal{D} : \Omega \times \mathbb{R} \rightarrow \mathbb{R}^+$  is bounded and Lipschitz
- ▶  $\bar{\mathbf{K}} : \Omega \rightarrow \mathbb{R}^{d \times d}$  is symmetric positive definite
- ▶  $f : \Omega \times \mathbb{R} \rightarrow \mathbb{R}$  is monotone and Lipschitz upto the boundary
- ▶  $\mathbf{q} : \Omega \times \mathbb{R} \rightarrow \mathbb{R}^d$  is bounded and satisfies a Lipschitz condition\*

with

$$\text{dist}(u, v) = \left\| \bar{\mathbf{K}}^{\frac{1}{2}} \nabla \int_u^v \mathcal{D} \right\|$$

## Class 2: gradient-independent diffusivity problems

For all  $\varphi \in H_0^1(\Omega)$ ,  $\mathcal{R} : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$  is defined as

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Semilinear equations  $-\Delta u = f(\mathbf{x}, u)$ 

Such equations pop up in quantum mechanics (special solutions to nonlinear Klein–Gordon equations), gravitation influences on stars, membrane buckling problems...

## Time-discrete nonlinear advection-reaction-diffusion equations

with time-step  $\tau > 0$ , the following evolutions equations reduce to this case

poro-Fischer equations:  $\partial_t u = \Delta u^m + \lambda u(1 - u)$

the Richards equation:  $\partial_t S(u) = \nabla \cdot [\bar{\mathbf{K}}(\mathbf{x})\kappa(S(u))(\nabla u + \mathbf{g})] + f(\mathbf{x}, u)$

biofilm equations:  $\partial_t u_k = \mu_k \Delta \Phi_k(u_k) + f_k((u_k)_{k=1}^n)$

### 3 Linearization schemes: practical examples

#### Abstract linearization

Considering the linearization operator

$$\mathfrak{L}(u_\ell^i; v, w) := (L(\mathbf{x}, u_\ell^i) v, w) + (\alpha(\mathbf{x}, u_\ell^i) \nabla v, \nabla w),$$

the coefficient functions for commonly used linearization schemes are

Scheme	$L(\mathbf{x}, v)$	$\alpha(\mathbf{x}, v)/\tau$
Picard (fixed point)	$\partial_\xi f(\mathbf{x}, v)$	$\bar{\mathbf{K}}(\mathbf{x}) \mathcal{D}(\mathbf{x}, v)$
Jäger–Kačur	$\max_{\xi \in \mathbb{R}} \left( \frac{f(\mathbf{x}, \xi) - f(\mathbf{x}, v)}{\xi - v} \right)$	$\bar{\mathbf{K}}(\mathbf{x}) \mathcal{D}(\mathbf{x}, v)$
$L$ -scheme	$L$ (constant) $\geq \frac{1}{2} \sup \partial_\xi f$	$\bar{\mathbf{K}}(\mathbf{x}) \mathcal{D}(\mathbf{x}, v)$
$M$ -scheme	$\partial_\xi f(\mathbf{x}, v) + M\tau$ (constant)	$\bar{\mathbf{K}}(\mathbf{x}) \mathcal{D}(\mathbf{x}, v)$

- 1 Introduction
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  - Gradient-independent diffusivity
  - The Newton scheme

### Effectivity indices

Global effectivity index: Eff. Ind. :=  $\eta_{\Omega}^i / \|\mathcal{R}(u_{\ell}^i)\|_{-1, u_{\ell}^i}$

Local effectivity index: (Eff. Ind.) $_K := \eta_K^i / \|\mathcal{R}(u_{\ell}^i)\|_{-1, u_{\ell}^i, K}$ ,  $K \in \mathcal{T}_{\ell}$ ,

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## 4 Gradient-independent diffusivity case: the Richards equation | 14

For  $\Omega = (0, 1) \times (0, 1)$  we study

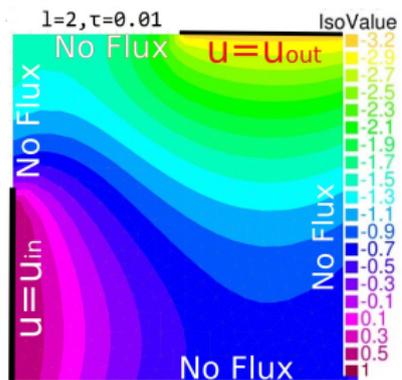
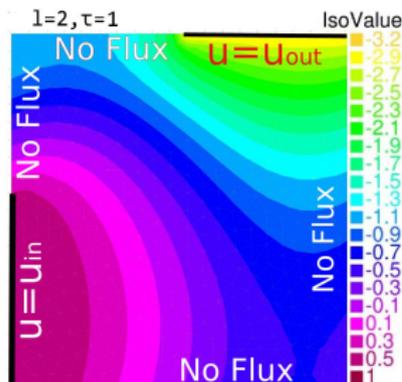
$$\langle \mathcal{R}(u_\ell), \varphi \rangle = (S(\bar{u}) - S(u_\ell), \varphi) - \tau(\bar{\mathbf{K}}\kappa(S(u_\ell))[\nabla u_\ell - \mathbf{g}], \nabla \varphi)$$

where the van Genuchten parametrization for  $S$ ,  $\kappa$  is used:

$$\begin{cases} S(\xi) := \left(1 + (2 - \xi)^{\frac{1}{1-\lambda}}\right)^{-\lambda}, \\ \kappa(s) := \sqrt{s} \left(1 - (1 - s^{\frac{1}{\lambda}})^{\lambda}\right)^2, \end{cases}$$

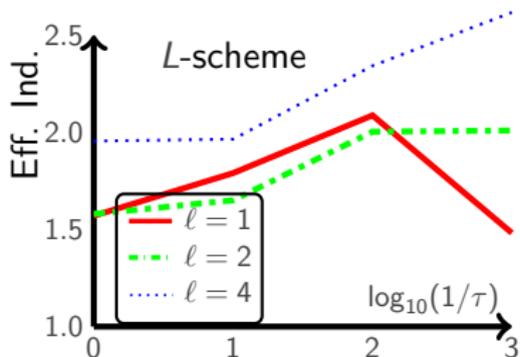
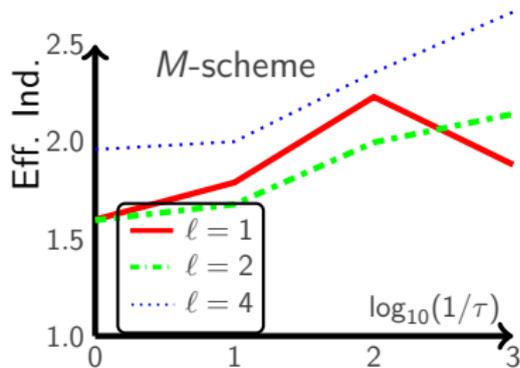
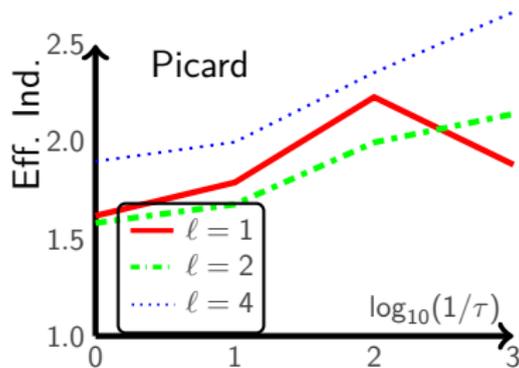
with  $\lambda = 0.5$ ,  $u_\ell^0 = 0$ ,

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

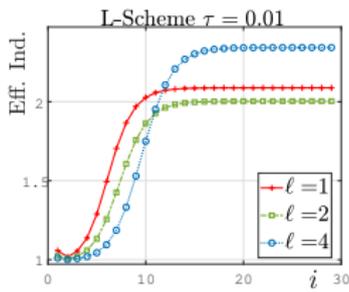
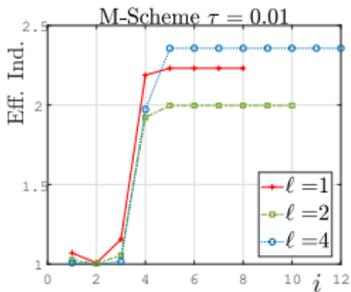
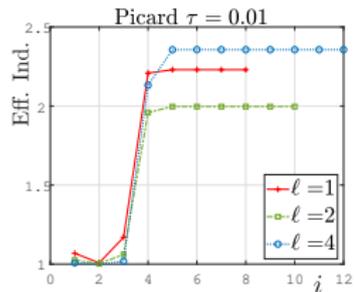
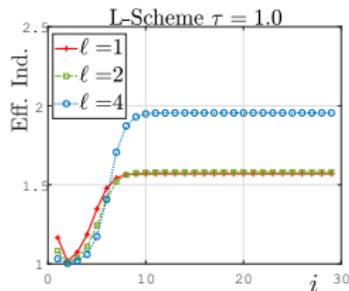
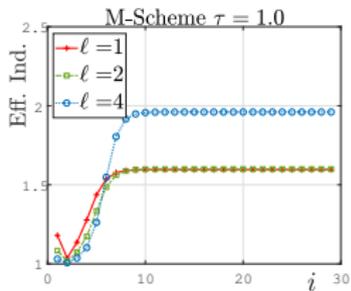
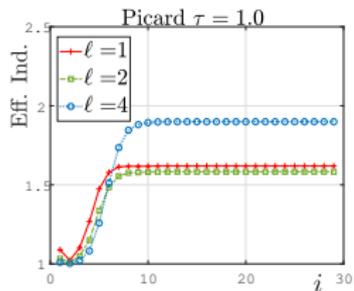


#### 4 Robustness with respect to $\lambda_M/\lambda_m$ represented by $1/\tau$

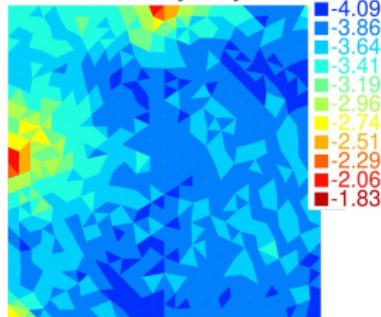
| 14



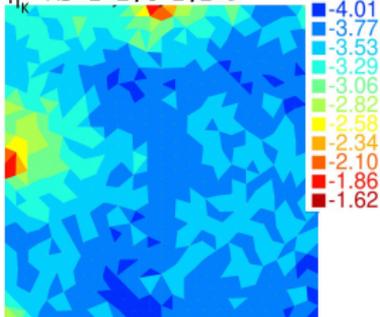
## 4 Global effectivity



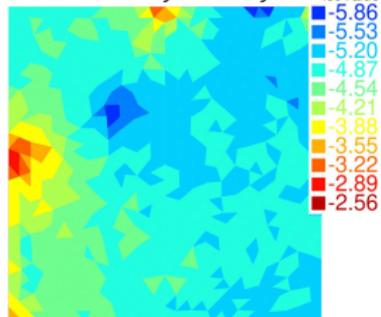
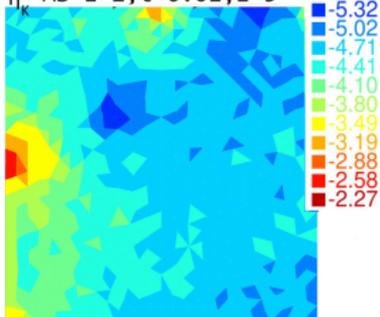
## 4 Distribution of error vs. estimates

Error MS  $l=2, \tau=1, i=9$ 

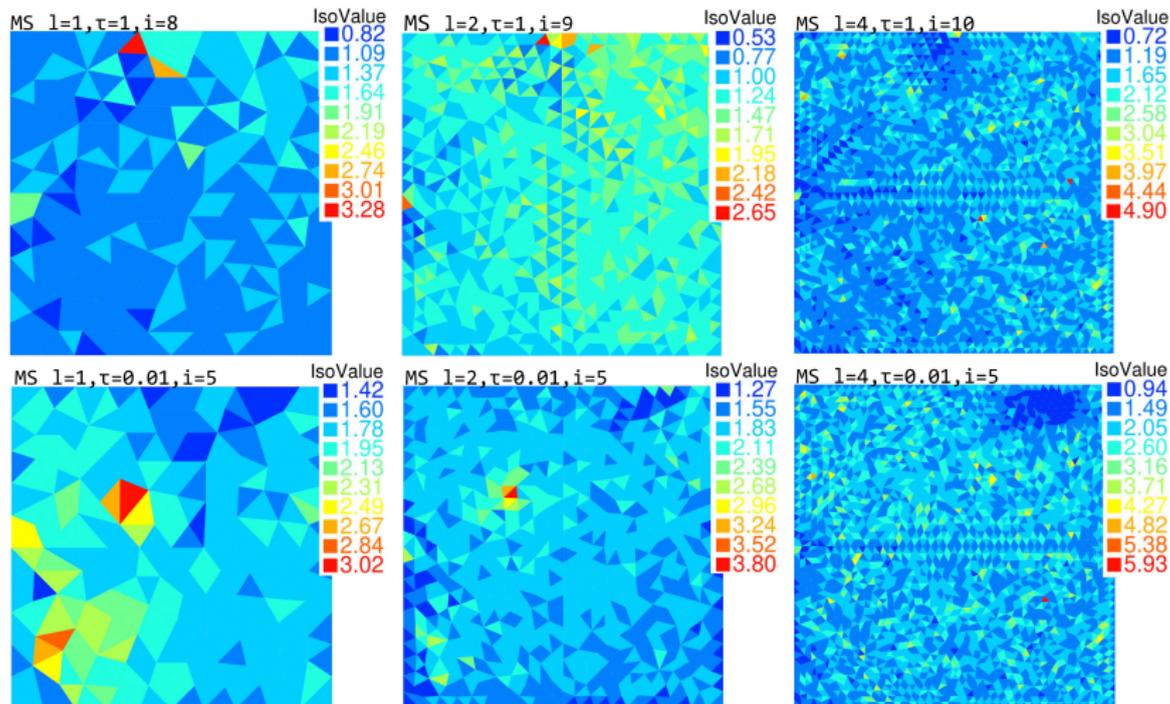
Error

 $\eta_k^i$  MS  $l=2, \tau=1, i=9$ 

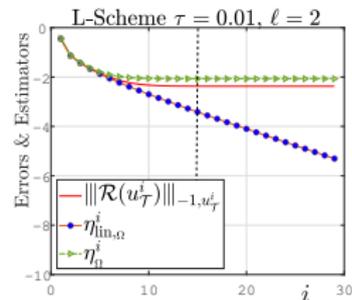
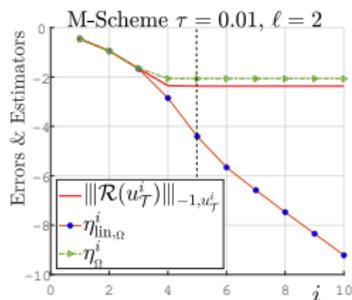
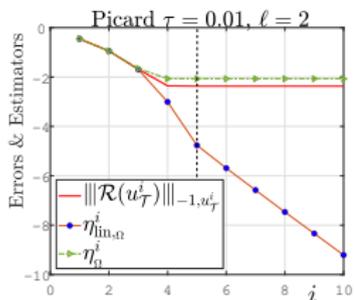
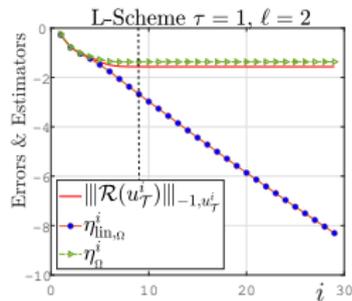
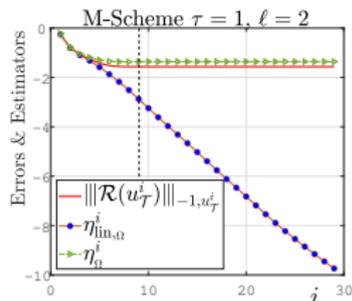
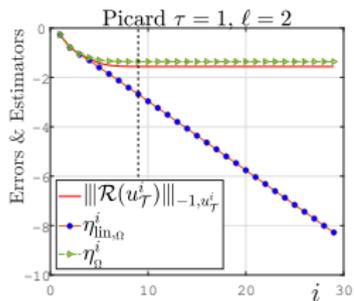
Estimate

Error MS  $l=2, \tau=0.01, i=5$  $\eta_k^i$  MS  $l=2, \tau=0.01, i=5$ 

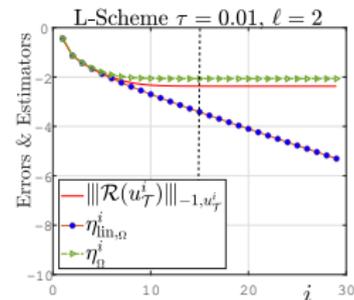
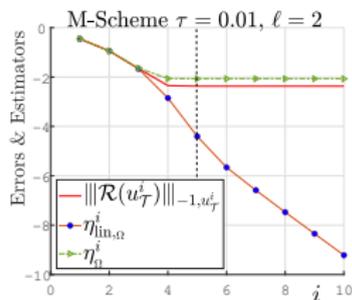
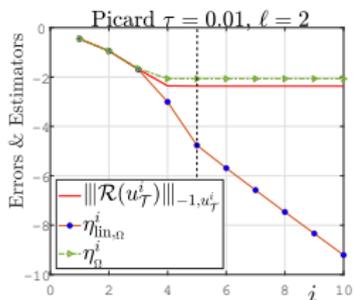
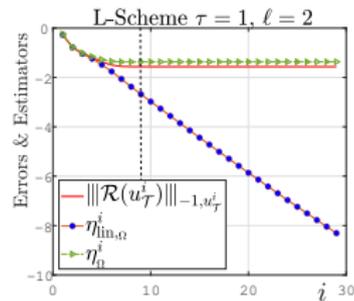
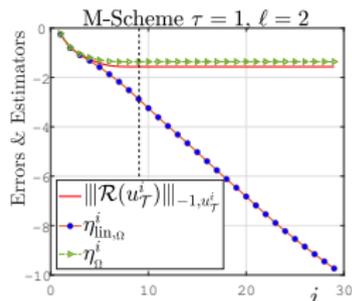
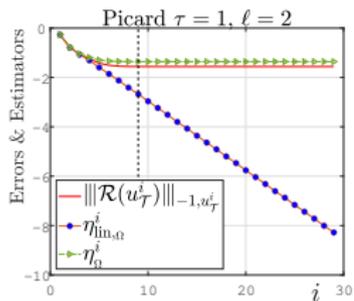
## 4 Local effectivity



## 4 Error with linearization iterations



## 4 Error with linearization iterations



Adaptive iteration stopping criteria:

$$\eta_{\text{lin}, \Omega}^i \leq 0.05 [\eta_\Omega^i].$$

## 4 The Newton scheme

For the Newton scheme, the linearization operator

$$\mathfrak{L}(u_\ell^i; v, w) := (L(\mathbf{x}, u_\ell^i) v, w) + (\mathfrak{a}(\mathbf{x}, u_\ell^i) \nabla v, \nabla w) + (\mathbf{w}(\mathbf{x}, u_\ell^i) v, \nabla w),$$

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is **non-symmetric**. However, if for some  $C_N \in [0, 2)$  we have

$$\mathbf{w}(\mathbf{x}, u_\ell^i) \mathfrak{a}^{-1}(\mathbf{x}, u_\ell^i) \mathbf{w}(\mathbf{x}, u_\ell^i) \leq C_N^2 L(\mathbf{x}, u_\ell^i), \quad \forall \mathbf{x} \in \Omega, \text{ and } i \in \mathbb{N},$$

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then,

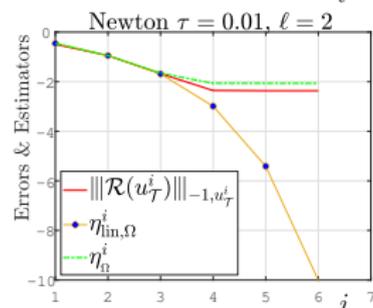
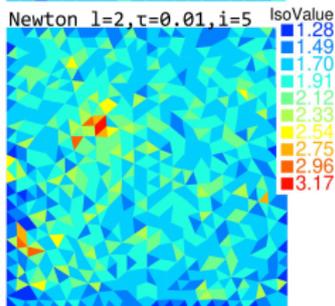
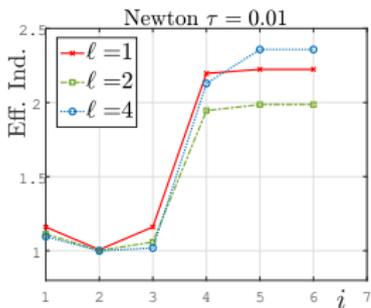
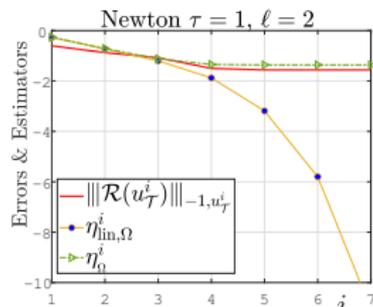
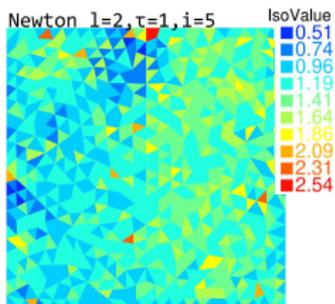
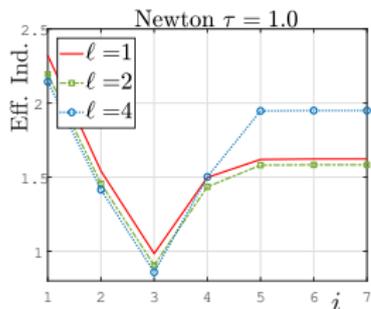
$$C_m(C_N) \left[ \left\| \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}) \right\|_{-1, u_\ell^i}^2 + \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}^2 \right] \leq \left\| \mathcal{R}(u_\ell^i) \right\|_{-1, u_\ell^i}^2$$

$$\leq C_M(C_N) \left[ \left\| \mathcal{R}_{\text{lin}}^{u_\ell^i}(u_\ell^{i+1}) \right\|_{-1, u_\ell^i}^2 + \left\| u_\ell^{i+1} - u_\ell^i \right\|_{1, u_\ell^i}^2 \right]$$

with  $C_m(C_N), C_M(C_N) \rightarrow 1$  if  $C_N \searrow 0$ .

## 4 The Newton scheme: numerical results

For gradient independent diffusivity case, we have



Global Effectivity

Local Effectivity

Error with iterations

## 4 Thank you for your time

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d'akujem Tak Dankie kiitos  
Спасибо תודה धन्यवाद terima kasih  
Asante Gracias شكرا mulțumesc hvala  
salamat 謝謝 Thank you Danke Hvala  
ありがとう Obrigado Merci Grazie 谢谢  
dank u ευχαριστώ Благодаря Děkuji  
ačiū Tack хвала Sağol تشکر از شما  
Дзякуй 감사합니다 dziękuję Спасибі  
paldies teşekkür ederim তোমাকে ধন্যবাদ