

Guaranteed and robust a posteriori error estimates for the reaction–diffusion and heat equations

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Outline

1 Introduction

2 The reaction–diffusion equation

- Equivalence between error and dual norm of the residual
- Guaranteed upper bound
- Local efficiency and robustness
- Numerical experiments

3 The heat equation

- Equivalence between error and dual norm of the residual
- High-order discretization & Radau reconstruction
- Guaranteed upper bound
- Local space-time efficiency and robustness
- Numerical experiments

4 Conclusions and future directions

An optimal a posteriori estimate for steady-state problems

Guaranteed upper bound

- $\|u - u_h\|_{?, \Omega}^2 \leq \sum_{K \in \mathcal{T}} \eta_K(u_h)^2$
- no undetermined constant: **error control**

Global efficiency

- $\sum_{K \in \mathcal{T}} \eta_K(u_h)^2 \leq C_{\text{eff}} \|u - u_h\|_{?, \Omega}^2$
- mathematical **equivalence** between the unknown **error** and known **estimate**

Robustness

- C_{eff} independent of the domain Ω , solutions u, u_h , **data**, meshes size and form

Small evaluation cost

- estimators $\eta_K(u_h)$ can be evaluated cheaply (locally) from u_h

Asymptotic exactness

- $\sum_{K \in \mathcal{T}} \eta_K(u_h)^2 / \|u - u_h\|_{?, \Omega}^2 \rightarrow 1$
- overestimation factor goes to one with increasing effort

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Local efficiency

- $\eta_K(u_h) \leq C_{\text{eff}} \|u - u_h\|_{?, \text{neighbors of } K}$
- **local** error lower bound (optimal space **mesh refinement**)

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Previous results (reaction–diffusion equation)

- Verfürth (1998) / Ainsworth and Babuška (1999): robustness wrt. singular perturbation
- Grosman (2006): robustness & anisotropic meshes, polynomial degree $p = 1$
- Cheddadi, Fučík, Prieto, Vohralík (2009): guaranteed upper bound & robustness, $p = 1$
- Ainsworth and Vejchodský (2011, 2014): guaranteed upper bound & robustness but requires submesh (complicated), (2019) without submesh (simple but with restrictions), $p = 1$
- Kopteva (2017): guaranteed upper bound, robustness, & anisotropic meshes, $p = 1$

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Find $u : \Omega \subset \mathbb{R}^d \rightarrow \mathbb{R}$, $d \geq 1$, such that

$$\begin{aligned}-\varepsilon^2 \Delta u + \kappa^2 u &= f \quad \text{in } \Omega, \\ u &= 0 \quad \text{on } \partial\Omega\end{aligned}$$

- $f \in L^2(\Omega)$, $\varepsilon > 0$, $\kappa \geq 0$ fixed real parameters

Singular perturbation

- $\varepsilon \ll \kappa$

Weak solution

Find $u \in H_0^1(\Omega)$ such that

$$\varepsilon^2 (\nabla u, \nabla v) + \kappa^2 (u, v) = (f, v) \quad \forall v \in H_0^1(\Omega)$$

Finite element approximation

Find $u_h \in V_h := \mathcal{P}_p(\mathcal{T}) \cap H_0^1(\Omega)$, $p \geq 1$, such that

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Equivalence between error and residual

Energy norm

$$\|\varphi\|^2 := \varepsilon^2 \|\nabla \varphi\|^2 + \kappa^2 \|\varphi\|^2 \quad \varphi \in H_0^1(\Omega)$$

$$\|\varphi\| = \sup_{v \in H_0^1(\Omega); \|v\|=1} \{ \varepsilon^2 (\nabla \varphi, \nabla v) + \kappa^2 (\varphi, v) \} \quad \varphi \in H_0^1(\Omega)$$

Residual of $u_h \in H_0^1(\Omega)$

- $\mathcal{R}(u_h) \in H^{-1}(\Omega)$, the misfit of u_h in the weak formulation:

$$\langle \mathcal{R}(u_h), v \rangle := (f, v) - \varepsilon^2 (\nabla u_h, \nabla v) - \kappa^2 (u_h, v) \quad v \in H_0^1(\Omega)$$

- dual norm of the residual

$$\|\mathcal{R}(u_h)\|_{H^{-1}(\Omega)} := \sup_{v \in H_0^1(\Omega); \|v\|=1} \langle \mathcal{R}(u_h), v \rangle$$

Energy error

is the dual norm of the residual

$$\|u - u_h\| = \sup_{v \in H_0^1(\Omega); \|v\|=1} \{ (f, v) - \varepsilon^2 (\nabla u_h, \nabla v) - \kappa^2 (u_h, v) \} = \|\mathcal{R}(u_h)\|_{H^{-1}(\Omega)}$$

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Energy error is the dual norm of the residual

$$\|(u - u_h)\| = \sup_{v \in H_0^1(\Omega); \|v\|=1} \{ \varepsilon^2 (\nabla(u - u_h), \nabla v) + \kappa^2 (u - u_h, v) \} = \|\mathcal{R}(u_h)\|_{H^{-1}(\Omega)}$$



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Upper bound: motivation

Bound on the residual

- let $\sigma_h \in H(\text{div}, \Omega)$ and $\phi_h \in L^2(\Omega)$ be such that $\boxed{\nabla \cdot \sigma_h + \kappa^2 \phi_h = f}$

σ_h : equilibrated flux reconstruction, $\approx -\varepsilon^2 \nabla u$

ϕ_h : potential reconstruction, $\approx u$

Green theorem $(\nabla \cdot \sigma_h, v) + (\sigma_h, \nabla v) = 0$ for $v \in H_0^1(\Omega)$:

$$\begin{aligned} \langle \mathcal{R}(u_h), v \rangle &= (f, v) - \varepsilon^2 (\nabla u_h, \nabla v) - \kappa^2 (u_h, v) \\ &= -(\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h, \varepsilon \nabla v) - (\kappa(u_h - \phi_h), \kappa v) \\ &\leq [\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|^2 + \|\kappa(u_h - \phi_h)\|^2]^{\frac{1}{2}} \|v\| \end{aligned}$$

- then

$$\|(u - u_h)\| = \|\mathcal{R}(u_h)\|_{H^{-1}(\Omega)} \leq [\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|^2 + \|\kappa(u_h - \phi_h)\|^2]^{\frac{1}{2}}$$

- how to obtain suitable practical (inexpensive) σ_h and ϕ_h ?
- counter-example where $\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|$ can largely overestimate the error

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$$\begin{aligned} \langle \mathcal{R}(u_h), v \rangle &= (f, v) - \varepsilon^2 (\nabla u_h, \nabla v) - \kappa^2 (u_h, v) \\ &= -(\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h, \varepsilon \nabla v) - (\kappa(u_h - \phi_h), \kappa v) \\ &\leq [\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|^2 + \|\kappa(u_h - \phi_h)\|^2]^{\frac{1}{2}} \|v\| \end{aligned}$$

- then

$$\|u - u_h\| = \|\mathcal{R}(u_h)\|_{H^{-1}(\Omega)} \leq [\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|^2 + \|\kappa(u_h - \phi_h)\|^2]^{\frac{1}{2}}$$

- how to obtain suitable **practical** (inexpensive) σ_h and ϕ_h ?
- counter-example** where $\|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|$ can largely overestimate the error

Upper bound: motivation

Bound on the residual

- let $\sigma_h \in H(\text{div}, \Omega)$ and $\phi_h \in L^2(\Omega)$ be such that $\boxed{\nabla \cdot \sigma_h + \kappa^2 \phi_h = f}$
- σ_h : **equilibrated flux reconstruction**, $\approx -\varepsilon^2 \nabla u$
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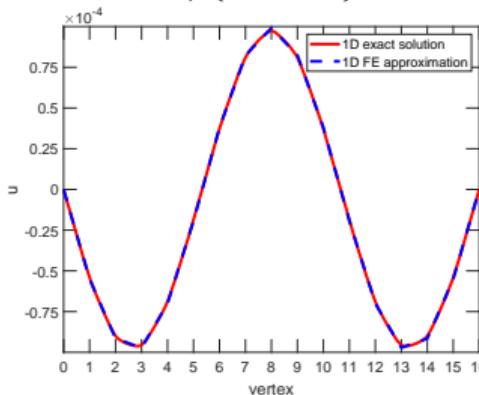
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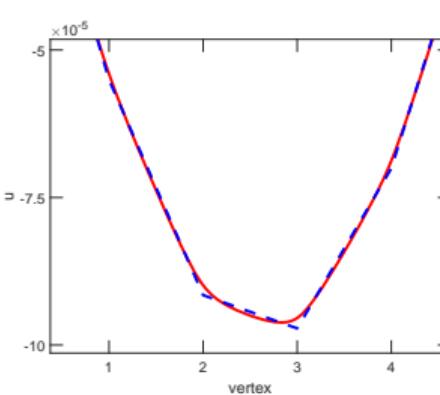
Counter-example ($-\varepsilon^2 \Delta u + \kappa^2 u = f$ in Ω , $u = 0$ on $\partial\Omega$)

Data

- $\Omega := (-1/2, 1/2)$, $d = 1$
- odd integer m
- f : piecewise affine Lagrange int. of $\cos(m\pi x)$
- uniform mesh \mathcal{T} with $2N = (m+1)^2$ intervals
- $h = 1/(m+1)^2$



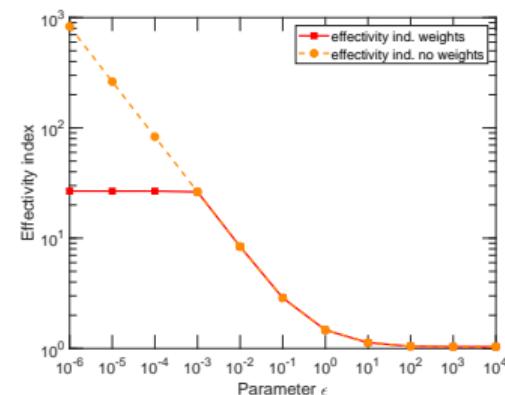
$$\varepsilon = 1, \kappa = 10^2, m = 3$$



detailed view

Finite element approximation

- $V_h := \mathcal{P}_1(\mathcal{T}) \cap H_0^1(\Omega)$
- $\mu_h := \frac{6}{2+\cos(m\pi h)} \frac{1-\cos(m\pi h)}{h^2}$
- $u_h = (\varepsilon^2 \mu_h + \kappa^2)^{-1} f$



$$\kappa = 10^2, m = 3, \text{eff. ind.}$$

Equilibrated flux and potential reconstructions

Definition (Flux σ_h and potential ϕ_h)

For each vertex $\mathbf{a} \in \mathcal{V}$, let

$$(\sigma_h^\mathbf{a}, \phi_h^\mathbf{a}) := \arg \min_{(v_h, q_h) \in \mathcal{RTN}_p(\mathcal{T}^\mathbf{a}) \times \mathcal{P}_p(\mathcal{T}^\mathbf{a})} J_{\Omega_h}(v_h, q_h)$$

$$J_{\Omega_h}(v_h, q_h) := \kappa^2 \|\varepsilon v_h \nabla u_h + \varepsilon^{-1} v_h\|_{\omega_\mathbf{a}}^2 + \|\kappa [\Pi_h(\psi_\mathbf{a} u_h) - q_h]\|_{\omega_\mathbf{a}}^2$$

Comments

- **local discrete** constrained minimization problems
- choose the locally best possible estimators
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$$\mathcal{J}_{\Omega_h}(v_h, q_h) := \kappa^2 \| \nabla v_h \nabla u_h + \kappa^{-1} v_h \|^2_{\omega_\mathbf{a}} + \| \kappa [\Pi_h(\nabla u_h) - q_h] \|^2_{\omega_\mathbf{a}}$$

with the weight $w_\mathbf{a} := \min \left\{ 1, C_s \sqrt{\frac{\kappa}{\kappa h_{\omega_\mathbf{a}}}} \right\}$. Combine

$$\sigma_h := \sum_{\mathbf{a} \in \mathcal{V}} \sigma_h^\mathbf{a} \in \mathcal{RTM}_p(\mathcal{T}^\mathbf{a} \cap H_0(\text{div}, Q)), \quad \phi_h := \sum_{\mathbf{a} \in \mathcal{V}} \phi_h^\mathbf{a} \in \mathcal{P}_p(\mathcal{T}).$$

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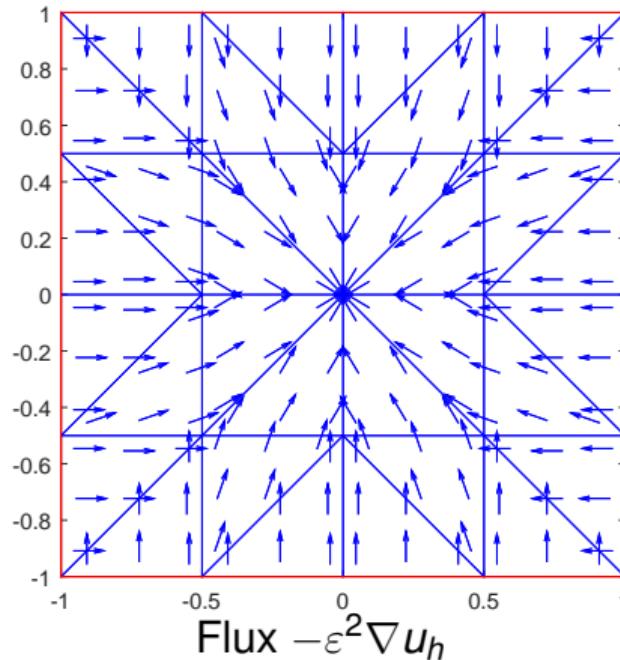
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Comments

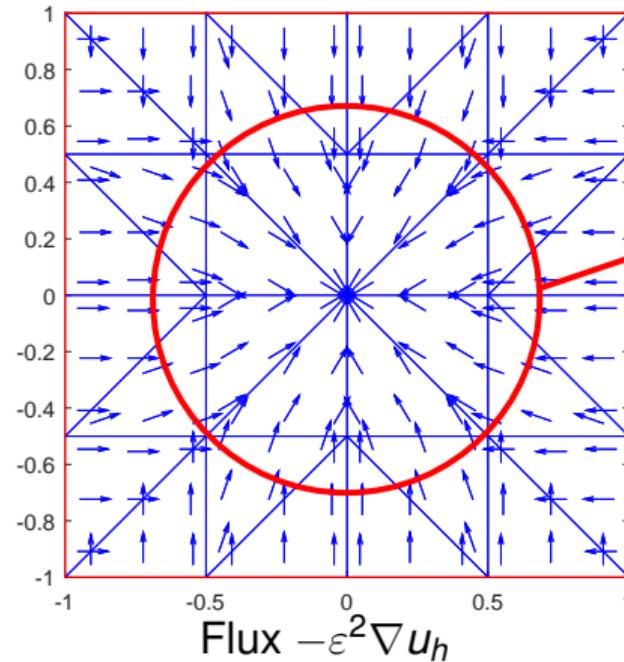
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Equilibrated flux reconstruction

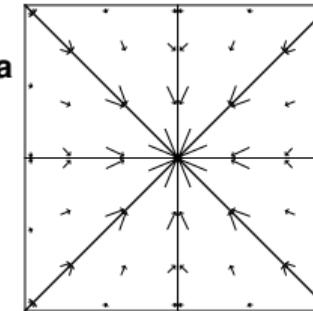


$$\underbrace{-\varepsilon^2 \nabla u_h \in \mathcal{RTN}_p(\mathcal{T}), f \in L^2(\Omega)}_{(f, \psi_a)_{\omega_a} - \varepsilon^2 (\nabla u_h, \nabla \psi_a)_{\omega_a} - \kappa^2 (u_h, \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}^{\text{int}}}$$

Equilibrated flux reconstruction



vertex patch T^a

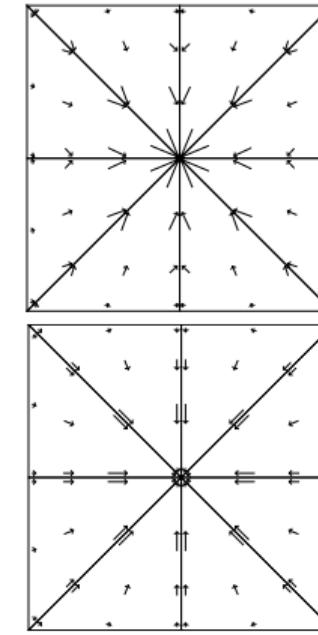
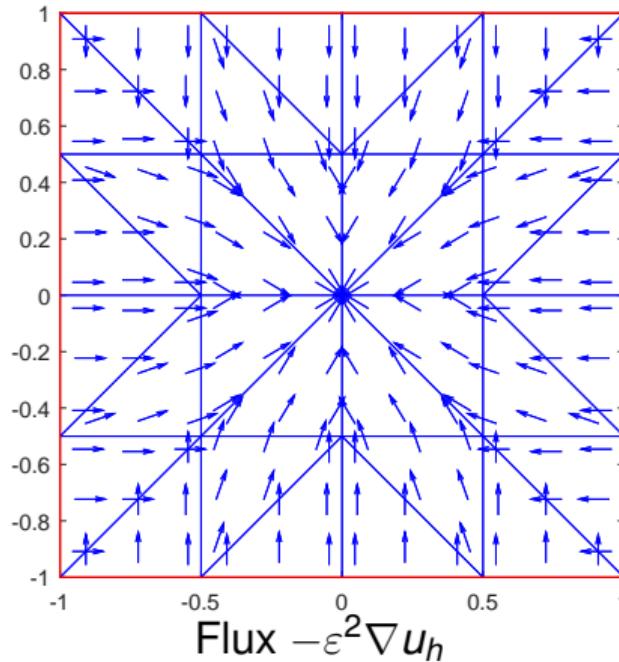


$$-\psi_a \varepsilon^2 \nabla u_h$$

$$-\varepsilon^2 \nabla u_h \in \mathcal{RTN}_p(T), f \in L^2(\Omega)$$

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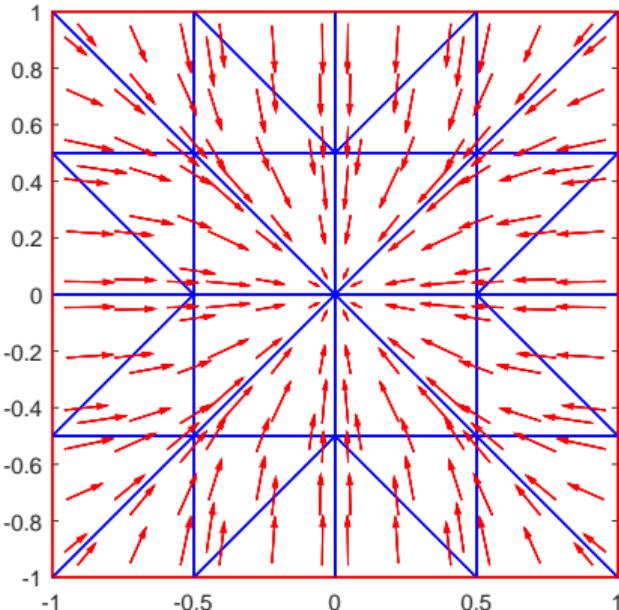
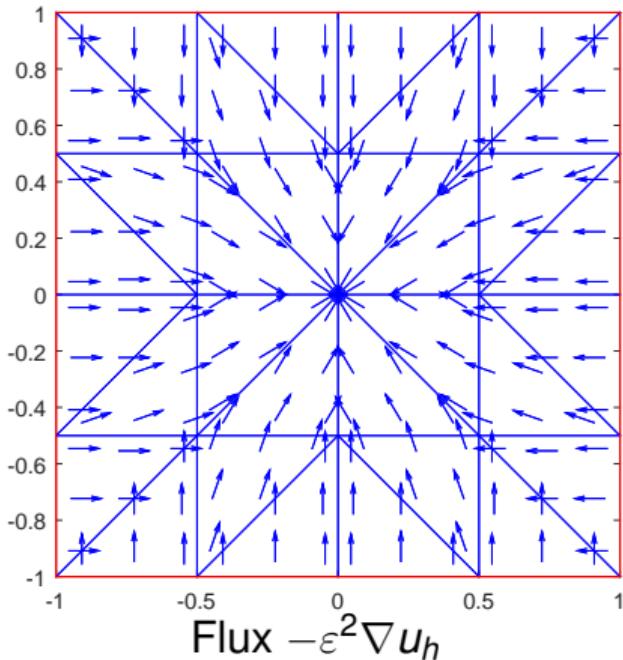


$$\sigma_h^a$$

$$-\varepsilon^2 \nabla u_h \in \mathcal{RTN}_p(\mathcal{T}), f \in L^2(\Omega)$$

$$(f, \psi_a)_{\omega_a} - \varepsilon^2 (\nabla u_h, \nabla \psi_a)_{\omega_a} - \kappa^2 (u_h, \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}^{\text{int}}$$

Equilibrated flux reconstruction



Equilibrated flux reconstruction σ_h

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Guaranteed a posteriori error estimate

Theorem (Guaranteed a posteriori error estimate)

Let \mathbf{u} be the weak solution and let $\mathbf{u}_h \in V_h$ be its finite element approximation. Let $\boldsymbol{\sigma}_h \in \mathcal{RTN}_p(\mathcal{T}) \cap \mathbf{H}(\text{div}, \Omega)$ and $\phi_h \in \mathcal{P}_p(\mathcal{T})$ be the flux and potential reconstructions. Then

$$\| \mathbf{u} - \mathbf{u}_h \|^2 \leq \sum_{K \in \mathcal{T}} [w_K \|\varepsilon \nabla \mathbf{u}_h + \varepsilon^{-1} \boldsymbol{\sigma}_h\|_K + \|\kappa (\mathbf{u}_h - \phi_h)\|_K + \tilde{w}_K \|f - \Pi_h f\|_K]^2$$

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- guaranteed upper bound on the unknown error
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1 Introduction

2 The reaction–diffusion equation

- Equivalence between error and dual norm of the residual
- Guaranteed upper bound
- **Local efficiency and robustness**
- Numerical experiments

3 The heat equation

- Equivalence between error and dual norm of the residual
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Let f be a piecewise polynomial for simplicity. Then, for all $K \in \mathcal{T}$,

$$w_K \|\varepsilon \nabla u_h + \varepsilon^{-1} \sigma_h\|_K + \|\kappa (u_h - \phi_h)\|_K \leq C_{\text{eff}} \|u - u_h\|_{\omega_K},$$

where the constant C_{eff} only depends on the space dimension d , the shape-regularity constant $\vartheta_{\mathcal{T}}$ of the mesh \mathcal{T} , and on the polynomial degree p of u_h .

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- the computable elementwise estimators are **local** lower bounds for the unknown error
- C_{eff} independent of the parameters ε and $\kappa \Rightarrow$ **robustness**

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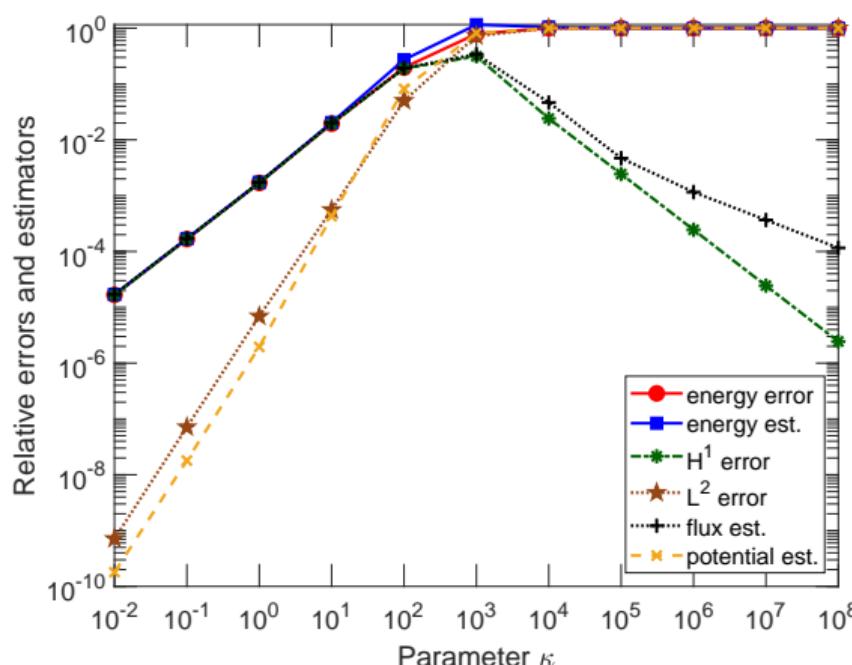
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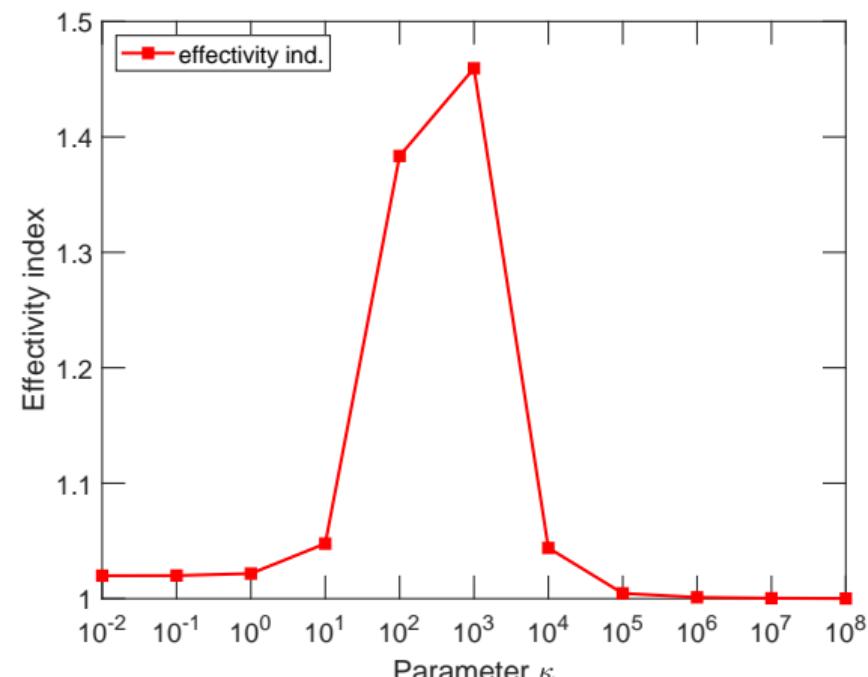
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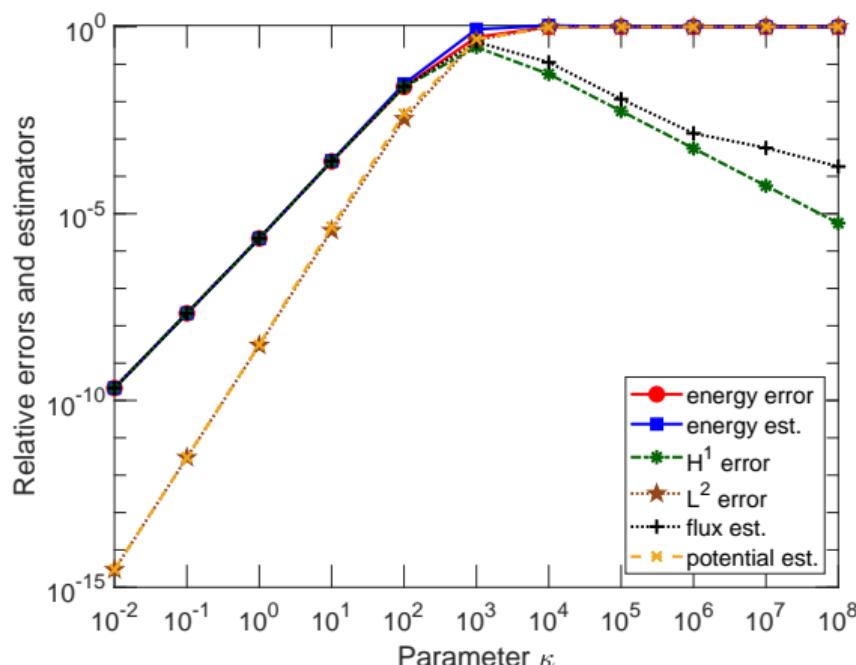
Boundary layer, solution $u(x, y) = e^{-\frac{\kappa}{\varepsilon}x} + e^{-\frac{\kappa}{\varepsilon}y}$, $p = 1$



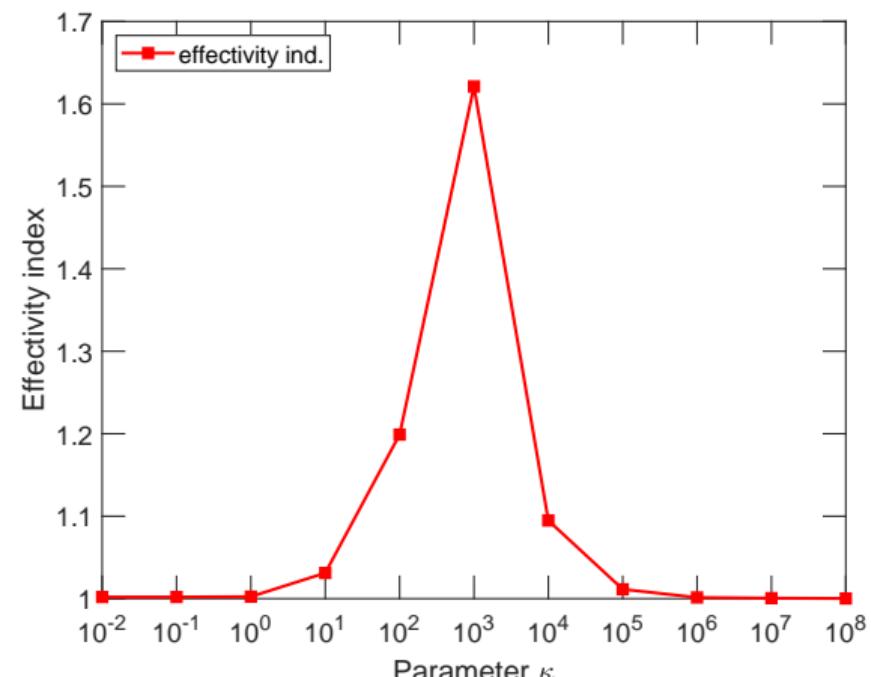
Relative energy errors and estimates

Effectivity indices $\eta(u_h)/\|u - u_h\|$

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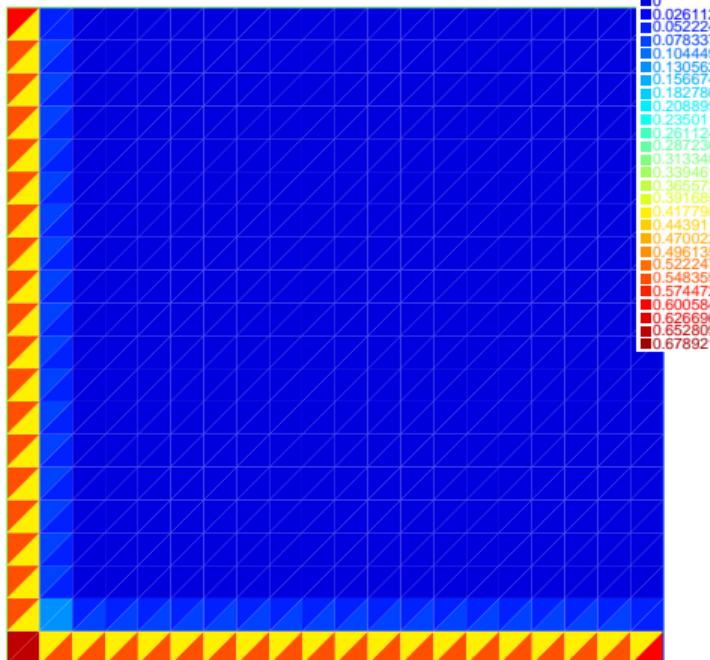


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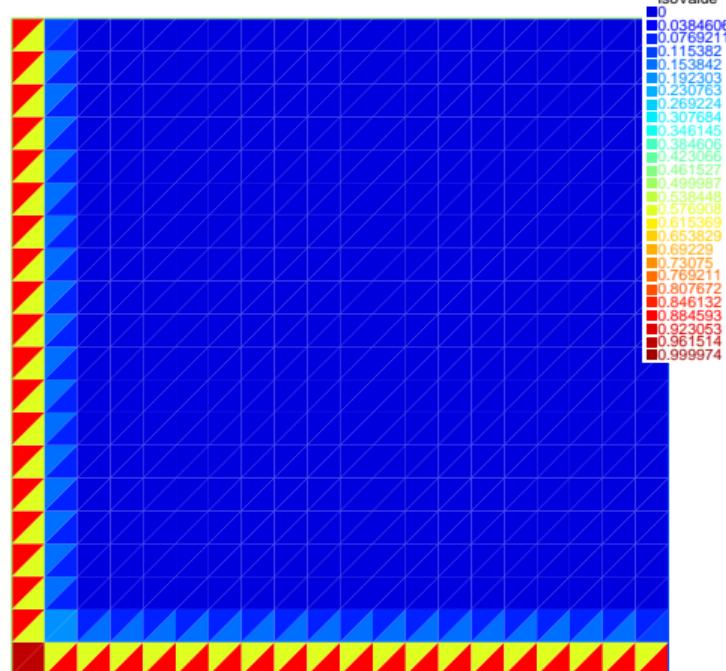
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energy errors



Exact en. error on each mesh element

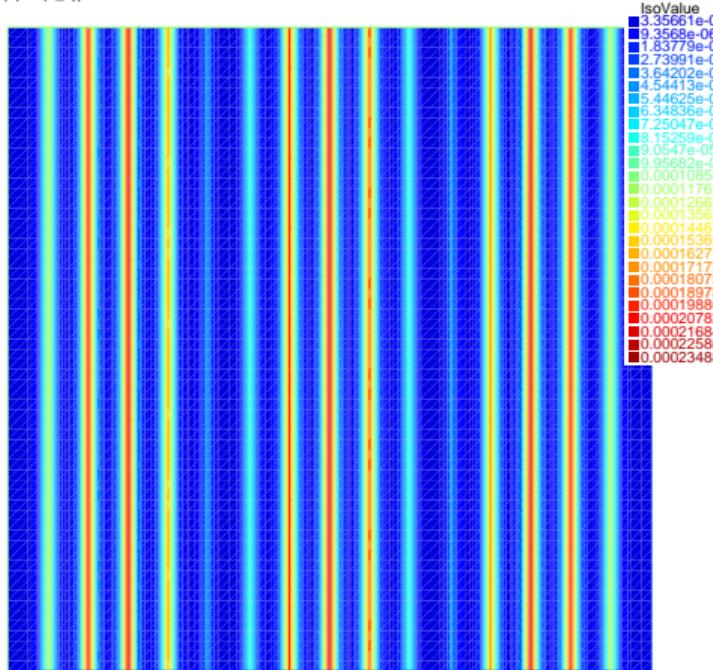
estimators



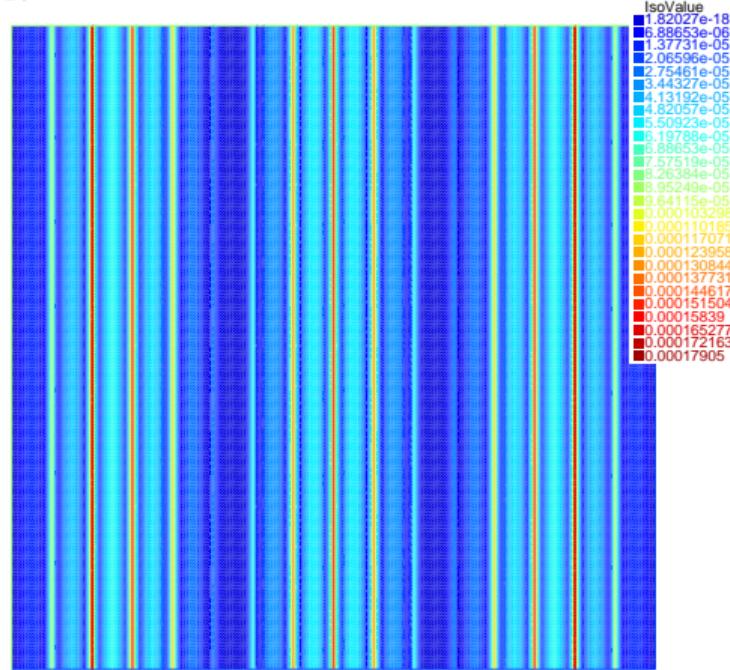
Estimated en. error on each mesh element

Counter-example with $\varepsilon = 1$, $\kappa = 10^2$, $m = 3$, $p = 1$

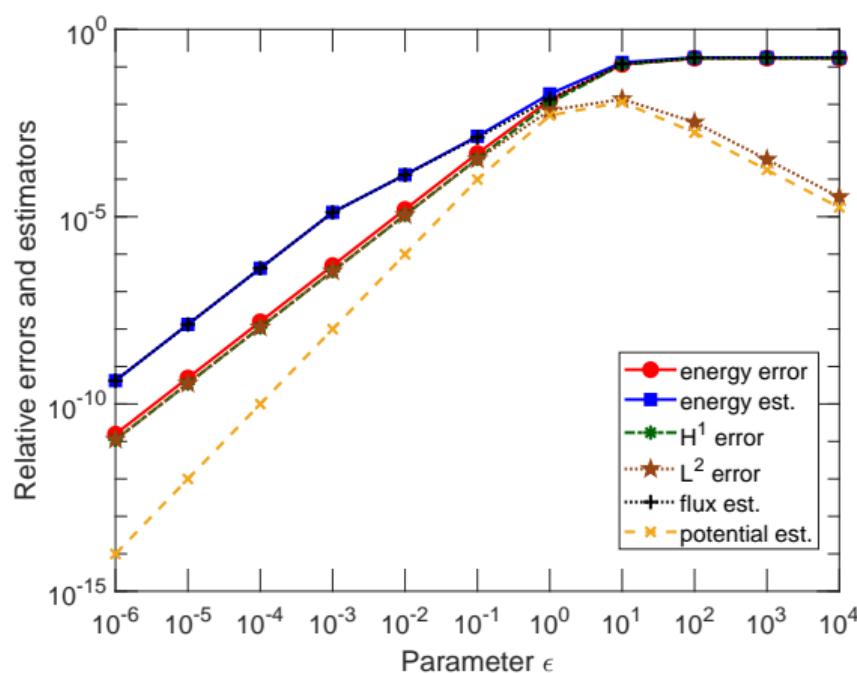
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Pointwise H^1 -seminorm errors

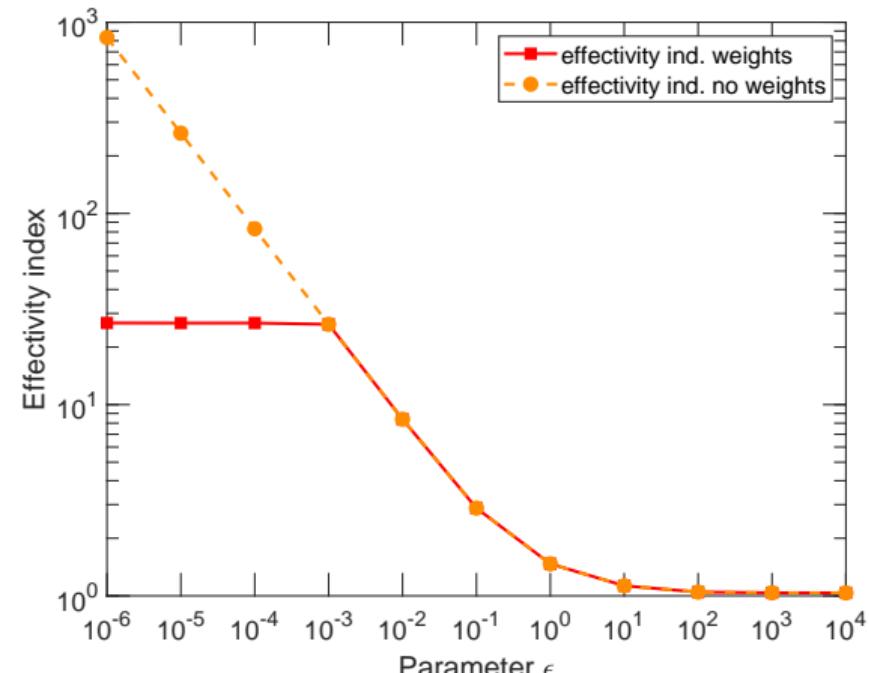
kappa*|u-u_h|

Pointwise $\kappa \times L^2$ -norm errors

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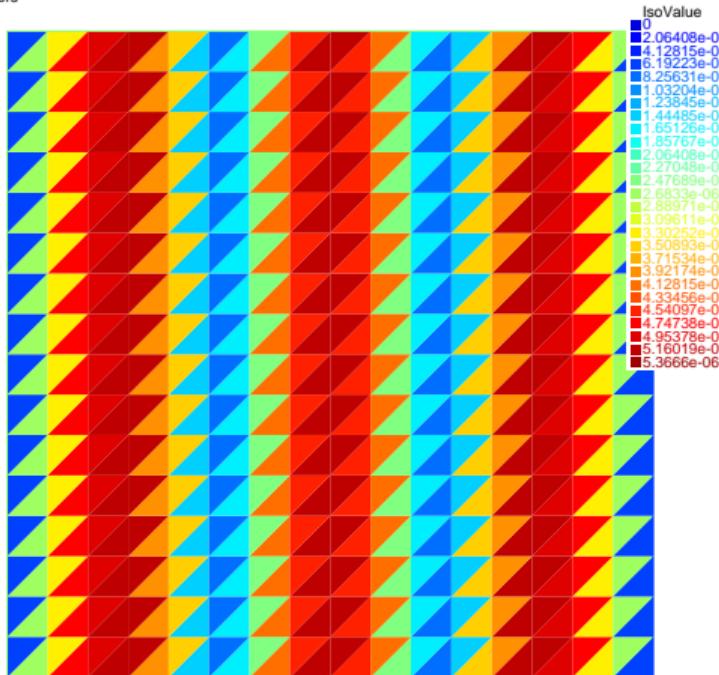


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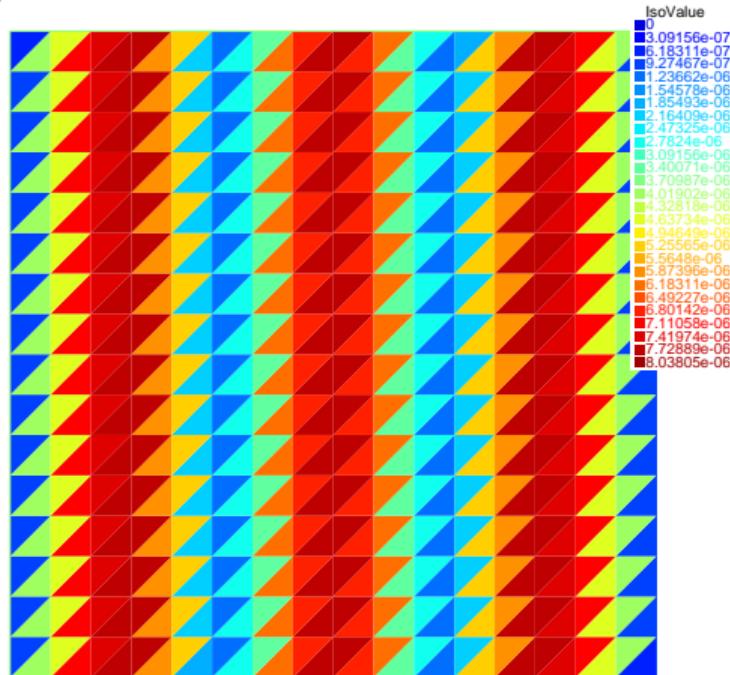
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The heat equation ($f \in L^2(0, T; L^2(\Omega))$, $u_0 \in L^2(\Omega)$)

The heat equation

$$\begin{aligned}\partial_t u - \Delta u &= f \quad \text{in } \Omega \times (0, T), \\ u &= 0 \quad \text{on } \partial\Omega \times (0, T), \\ u(0) &= u_0 \quad \text{in } \Omega\end{aligned}$$

Spaces

$$\begin{aligned}X &:= L^2(0, T; H_0^1(\Omega)), \\ \|v\|_X^2 &:= \int_0^T \|\nabla v\|^2 dt, \\ Y &:= L^2(0, T; H_0^1(\Omega)) \cap H^1(0, T; H^{-1}(\Omega)), \\ \|v\|_Y^2 &:= \int_0^T \|\partial_t v\|_{H^{-1}(\Omega)}^2 + \|\nabla v\|^2 dt + \|v(T)\|^2\end{aligned}$$

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

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Guaranteed upper bound

- $\|u - u_{h\tau}\|_{?, \Omega \times (0, T)}^2 \leq \sum_{n=1}^N \sum_{K \in \mathcal{T}^n} \eta_K^n(u_{h\tau})^2$
- no undetermined constant: **error control**

Local efficiency

- $\eta_K^n(u_{h\tau}) \leq C_{\text{eff}} \|u - u_{h\tau}\|_{?, \text{neighbors of } K \times (t^{n-1}, t^n)}$
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- **local** in **time** and in **space** error lower bound

Robustness

- C_{eff} independent of data, domain Ω , **final time** T , meshes, solutions u , $u_{h\tau}$, **polynomial degrees** of $u_{h\tau}$ in space p and in time q

Small evaluation cost

- estimators $\eta_K^n(u_{h\tau})$ can be **evaluated cheaply** (locally) **from** $u_{h\tau}$

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- $\sum_{n=1}^N \sum_{K \in \mathcal{T}^n} \eta_K^n(u_h)^2 / \|u - u_h\|_{?, \Omega \times (0, T)}^2 \rightarrow 1$
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 - ✗ constrained lower bound (h and τ strongly linked)
- Repin (2002), guaranteed upper bound
- Verfürth (2003) (cf. also Bergam, Bernardi, and Mghazli (2005)), work with the Y norm:
 - ✓ upper bound $\|u - u_{h\tau}\|_Y^2 \leq C \sum_{n=1}^N \sum_{K \in \mathcal{T}^n} \eta_K^n(u_{h\tau})^2$
 - ✓ efficiency $\sum_{K \in \mathcal{T}^n} \eta_K^n(u_{h\tau})^2 \leq C \|u - u_{h\tau}\|_{Y(I_n)}^2$
 - ✓ robustness with respect to the final time T , no link $h \leftrightarrow \tau$
 - ✗ efficiency local in time but global in space
- Makridakis and Nochetto (2006): Radau reconstruction
- Ern and Vohralík (2010): unified framework for different spatial discretizations (FEs, NCFEs, DGs, MFEs, FVs)

Outline

1 Introduction

2 The reaction–diffusion equation

- Equivalence between error and dual norm of the residual
- Guaranteed upper bound
- Local efficiency and robustness
- Numerical experiments

3 The heat equation

- Equivalence between error and dual norm of the residual
- High-order discretization & Radau reconstruction
- Guaranteed upper bound
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4 Conclusions and future directions

Equivalence between error and residual

The heat equation

$$\begin{aligned}\partial_t u - \Delta u &= f \quad \text{in } \Omega \times (0, T), \\ u &= 0 \quad \text{on } \partial\Omega \times (0, T), \\ u(0) &= u_0 \quad \text{in } \Omega\end{aligned}$$

Spaces

$$\begin{aligned}X &:= L^2(0, T; H_0^1(\Omega)), \\ \|v\|_X^2 &:= \int_0^T \|\nabla v\|^2 dt, \\ Y &:= L^2(0, T; H_0^1(\Omega)) \cap H^1(0, T; H^{-1}(\Omega)), \\ \|v\|_Y^2 &:= \int_0^T \|\partial_t v\|_{H^{-1}(\Omega)}^2 + \|\nabla v\|^2 dt + \|v(T)\|^2\end{aligned}$$

Weak solution

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

$$\int_0^T \langle \partial_t u, v \rangle + (\nabla u, \nabla v) dt = \int_0^T (f, v) dt \quad \forall v \in X$$

Equivalence between error and residual

Theorem (Parabolic inf–sup identity)

For every $\varphi \in Y$, we have

$$\|\varphi\|_Y^2 = \left[\sup_{v \in X, \|v\|_X=1} \int_0^T \langle \partial_t \varphi, v \rangle + (\nabla \varphi, \nabla v) dt \right]^2 + \|\varphi(0)\|^2.$$

Residual of $u_{h\tau} \in Y$

- $\mathcal{R}(u_{h\tau}) \in X'$, the misfit of $u_{h\tau}$ in the weak formulation:

$$\langle \mathcal{R}(u_{h\tau}), v \rangle := \int_0^T (f, v) - \langle \partial_t u_{h\tau}, v \rangle - (\nabla u_{h\tau}, \nabla v) dt \quad v \in X$$

- dual norm of the residual

$$\|\mathcal{R}(u_{h\tau})\|_{X'} := \sup_{v \in X, \|v\|_X=1} \langle \mathcal{R}(u_{h\tau}), v \rangle$$

Y norm error is the dual X norm of the residual + initial condition error

$$\|u - u_{h\tau}\|_Y^2 = \|\mathcal{R}(u_{h\tau})\|_{X'}^2 + \|u_0 - u_{h\tau}(0)\|^2$$



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Proof of the parabolic inf–sup identity: $\varphi \in Y$

Proof.

- let $w_* \in X$ be defined by, a.e. in $(0, T)$,

$$(\nabla w_*, \nabla v) = \langle \partial_t \varphi, v \rangle \quad \forall v \in H_0^1(\Omega) \Rightarrow \|\nabla w_*\|^2 = \|\partial_t \varphi\|_{H^{-1}(\Omega)}^2$$

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2 The reaction–diffusion equation

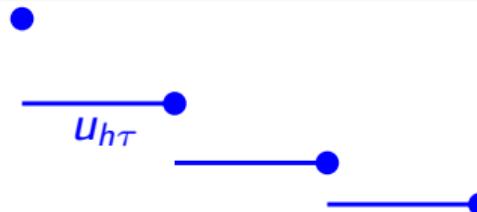
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4 Conclusions and future directions

Approximate solution and Radau reconstruction



Approximate solution

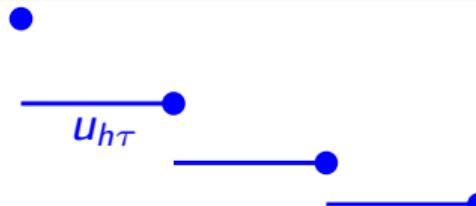
- ✓ $u_{h\tau}(t)$, $t \in I_n$, is a piecewise **continuous** polynomial in space in $V_h^n := \{v_h \in H_0^1(\Omega), v_h|_K \in \mathcal{P}_{p_K}(K) \quad \forall K \in \mathcal{T}^n\}$
- ✗ $u_{h\tau}$ is a piecewise **discontinuous** polynomial in time
- ✗ $u_{h\tau} \notin Y \Rightarrow$ impossible to estimate $\|u - u_{h\tau}\|_Y$

Radau reconstruction

- ✓ $\mathcal{I}u_{h\tau} \in Y$, $\mathcal{I}u_{h\tau}|_{I_n} \in Q_{q_n+1}(I_n; \widetilde{V}_h^n)$ (Makridakis–Nochetto)

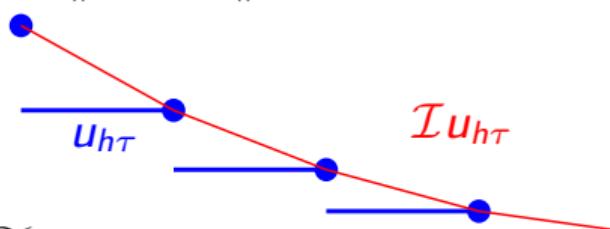
$$\int_{I_n} (\mathcal{I}u_{h\tau}, v_{h\tau}) + (\nabla u_{h\tau}, \nabla v_{h\tau}) dt = \int_{I_n} (f, v_{h\tau}) dt \quad \forall v_{h\tau} \in Q_q(I_n; V_h^n)$$

Approximate solution and Radau reconstruction



Approximate solution

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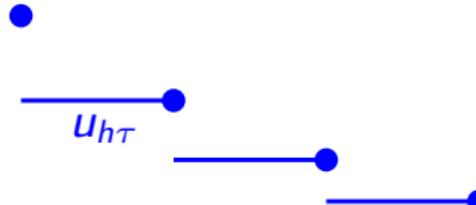


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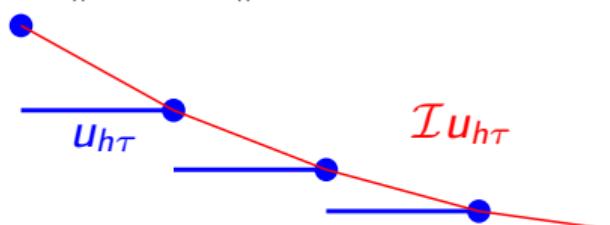
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Results in the Y norm

Theorem (Guaranteed upper bound in the Y norm)

Suppose no data oscillation (f and u_0 piecewise polynomial). Then, for any $\sigma_{h\tau} \in L^2(0, T; \mathbf{H}(\text{div}, \Omega))$ with $\nabla \cdot \sigma_{h\tau} = f - \partial_t \mathcal{I}u_{h\tau}$, there holds

$$\|u - \mathcal{I}u_{h\tau}\|_Y^2 \leq \int_0^T \|\sigma_{h\tau} + \nabla \mathcal{I}u_{h\tau}\|^2 dt.$$

Proof of the upper bound

Proof.

- equivalence error-residual (supposing no error in the initial condition):

$$\|u - \mathcal{I}u_{h\tau}\|_Y = \sup_{v \in X, \|v\|_X=1} \langle \mathcal{R}(\mathcal{I}u_{h\tau}), v \rangle$$

- Green theorem

$$\int_0^T (\sigma_{h\tau}, \nabla \mathcal{I}u_{h\tau}) + (\nabla \cdot \sigma_{h\tau}, \mathcal{I}u_{h\tau}) dt = 0$$

- residual definition, Cauchy–Schwarz inequality:

$$\begin{aligned} \langle \mathcal{R}(\mathcal{I}u_{h\tau}), v \rangle &= \int_0^T (f, v) - (\partial_t \mathcal{I}u_{h\tau}, v) - (\nabla \mathcal{I}u_{h\tau}, \nabla v) dt \\ &= \int_0^T \underbrace{(f - \partial_t \mathcal{I}u_{h\tau} - \nabla \cdot \sigma_{h\tau}, v)}_{=0} - (\nabla \mathcal{I}u_{h\tau} + \sigma_{h\tau}, \nabla v) dt \\ &\leq \left\{ \int_0^T \|\sigma_{h\tau} + \nabla \mathcal{I}u_{h\tau}\|^2 dt \right\}^{\frac{1}{2}} \|v\|_X \end{aligned}$$

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Equilibrated flux reconstruction

Definition (Equilibrated flux reconstruction)

For each time-step interval I_n and for each vertex $\mathbf{a} \in \mathcal{V}^n$, let

$$\sigma_{h\tau}^{\mathbf{a},n} := \arg \min_{\mathbf{v}_h \in \mathcal{V}_{h\tau}^{\mathbf{a},n}} \int_{I_n} \|\mathbf{v}_h + \psi_{\mathbf{a}} \nabla \mathbf{u}_{h\tau}\|_{\omega_{\mathbf{a}}}^2 dt.$$

$\nabla \cdot \mathbf{v}_h = \psi_{\mathbf{a}}(f - \partial_t \mathcal{I} \mathbf{u}_{h\tau}) - \nabla \psi_{\mathbf{a}} \cdot \nabla \mathbf{u}_{h\tau}$

Then set

$$\sigma_{h\tau} := \sum_{n=1}^N \sum_{\mathbf{a} \in \mathcal{V}^n} \sigma_{h\tau}^{\mathbf{a},n}.$$

Comments

- ✓ satisfies $\sigma_{h\tau} \in L^2(0, T; \mathbf{H}(\text{div}, \Omega))$ with $\nabla \cdot \sigma_{h\tau} = f - \partial_t \mathcal{I} \mathbf{u}_{h\tau}$
- ✗ a priori a local space-time problem, $\mathcal{V}_{h\tau}^{\mathbf{a},n} := \mathcal{Q}_{q_n}(I_n; \mathcal{V}_h^{\mathbf{a},n})$
- ✓ uncouples to q_n elliptic problems posed in $\mathcal{V}_h^{\mathbf{a},n}$

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- ✓ satisfies $\boldsymbol{\sigma}_{h\tau} \in L^2(0, T; \mathbf{H}(\text{div}, \Omega))$ with $\nabla \cdot \boldsymbol{\sigma}_{h\tau} = f - \partial_t \mathcal{I} \mathbf{u}_{h\tau}$
- ✗ a priori a local space-time problem, $\mathcal{V}_{h\tau}^{\mathbf{a},n} := \mathcal{Q}_{q_n}(I_n; \mathcal{V}_h^{\mathbf{a},n})$
- ✓ uncouples to q_n elliptic problems posed in $\mathcal{V}_h^{\mathbf{a},n}$

Equilibrated flux reconstruction

Definition (Equilibrated flux reconstruction)

For each time-step interval I_n and for each vertex $\mathbf{a} \in \mathcal{V}^n$, let

$$\sigma_{h\tau}^{\mathbf{a},n} := \arg \min_{\mathbf{v}_h \in \mathbf{V}_{h\tau}^{\mathbf{a},n}} \int_{I_n} \|\mathbf{v}_h + \psi_{\mathbf{a}} \nabla \mathbf{u}_{h\tau}\|_{\omega_{\mathbf{a}}}^2 dt.$$

$\nabla \cdot \mathbf{v}_h = \psi_{\mathbf{a}}(f - \partial_t \mathcal{I} \mathbf{u}_{h\tau}) - \nabla \psi_{\mathbf{a}} \cdot \nabla \mathbf{u}_{h\tau}$

Then set

$$\boldsymbol{\sigma}_{h\tau} := \sum_{n=1}^N \sum_{\mathbf{a} \in \mathcal{V}^n} \sigma_{h\tau}^{\mathbf{a},n}.$$

Comments

- ✓ satisfies $\boldsymbol{\sigma}_{h\tau} \in L^2(0, T; \mathbf{H}(\text{div}, \Omega))$ with $\nabla \cdot \boldsymbol{\sigma}_{h\tau} = f - \partial_t \mathcal{I} \mathbf{u}_{h\tau}$
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1 Introduction

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- Equivalence between error and dual norm of the residual
- Guaranteed upper bound
- Local efficiency and robustness
- Numerical experiments

3 The heat equation

- Equivalence between error and dual norm of the residual
- High-order discretization & Radau reconstruction
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- Local space-time efficiency and robustness**
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4 Conclusions and future directions

Global efficiency \sim missing Galerkin orthogonality

Efficiency

There holds

$$\int_{I_n} \|\sigma_{h\tau} + \nabla \mathcal{I}u_{h\tau}\|_{\Omega}^2 dt \leq C_{\text{eff}}^2 \|u - \mathcal{I}u_{h\tau}\|_{Y(I_n)} \quad \forall 1 \leq n \leq N.$$

- ✗ local-in-time but **global-in-space** only (as in Verfürth & Bergam–Bernardi–Mghazli)

Reason

- ✗ $\mathcal{I}u_{h\tau}$ misses the Galerkin orthogonality:

$$\int_{I_n} (f, v_{h\tau}) - (\partial_t \mathcal{I}u_{h\tau}, v_{h\tau}) - (\nabla \mathcal{I}u_{h\tau}, \nabla v_{h\tau}) dt$$

the LBB condition is violated

Global efficiency \sim missing Galerkin orthogonality

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$$\int_{I_n} (f, v_{h\tau}) - (\partial_t \mathcal{I}u_{h\tau}, v_{h\tau}) - (\nabla \mathcal{I}u_{h\tau}, \nabla v_{h\tau}) dt \neq 0 \quad \forall v_{h\tau} \in Q_{q_n}(I_n; V_h^n)$$

✓ the misfit is known: $u_{h\tau} - \mathcal{I}u_{h\tau}$

Global efficiency \sim missing Galerkin orthogonality

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- ✓ the misfit is known: $u_{h\tau} - \mathcal{I}u_{h\tau}$

Guaranteed upper bound

A decisive trick

Define the final norm as

$$\|u - u_{h\tau}\|_{\mathcal{E}_Y}^2 := \|u - \mathcal{I}u_{h\tau}\|_Y^2 + \underbrace{\|u_{h\tau} - \mathcal{I}u_{h\tau}\|_X^2}_{\text{known, computable}}$$

Theorem (Guaranteed upper bound)

Suppose no data oscillation (f and u_0 piecewise polynomial). Then there holds

$$\|u - u_{h\tau}\|_{\mathcal{E}_Y}^2 \leq \sum_{n=1}^N \sum_{K \in \mathcal{T}^n} \int_{I_n} \|\sigma_{h\tau} + \nabla \mathcal{I}u_{h\tau}\|_K^2 + \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|_K^2 dt.$$

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Local space-time efficiency and robustness

Local error contributions

$$\begin{aligned} |u - u_{h\tau}|_{\mathcal{E}_Y^{\mathbf{a},n}}^2 &= \int_{I_n} \|\partial_t(u - \mathcal{I}u_{h\tau})\|_{H^{-1}(\omega_{\mathbf{a}})}^2 + \|\nabla(u - \mathcal{I}u_{h\tau})\|_{\omega_{\mathbf{a}}}^2 dt \\ &\quad + \int_{I_n} \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|_{\omega_{\mathbf{a}}}^2 dt \end{aligned}$$

Theorem (Local space-time efficiency and robustness)

For each time-step interval I_n and for each element $K \in \mathcal{T}^n$, there holds, in the absence of data oscillation,

$$\int_{I_n} \|\sigma_{h\tau} + \nabla \mathcal{I}u_{h\tau}\|_K^2 + \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|_K^2 dt \leq C_{\text{eff}}^2 \sum_{\mathbf{a} \in \mathcal{V}_K} |u - u_{h\tau}|_{\mathcal{E}_Y^{\mathbf{a},n}}^2.$$

Comments

- ✓ local in space and in time
- ✓ C_{eff} only depends on shape regularity \Rightarrow robustness w.r.t the final time T and the polynomial degrees p and q

Local space-time efficiency and robustness

Local error contributions

$$\begin{aligned} |u - u_{h\tau}|_{\mathcal{E}_Y^{\mathbf{a}, \textcolor{brown}{n}}}^2 &= \int_{I_n} \|\partial_t(u - \mathcal{I}u_{h\tau})\|_{H^{-1}(\omega_{\mathbf{a}})}^2 + \|\nabla(u - \mathcal{I}u_{h\tau})\|_{\omega_{\mathbf{a}}}^2 dt \\ &\quad + \int_{I_n} \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|_{\omega_{\mathbf{a}}}^2 dt \end{aligned}$$

recall

$$\begin{aligned} \|u - u_{h\tau}\|_{\mathcal{E}_Y}^2 &= \int_0^T \|\partial_t(u - \mathcal{I}u_{h\tau})\|_{H^{-1}(\Omega)}^2 dt + \int_0^T \|\nabla(u - \mathcal{I}u_{h\tau})\|^2 dt \\ &\quad + \int_0^T \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|^2 dt + \|(u - \mathcal{I}u_{h\tau})(T)\|^2 \end{aligned}$$

Local space-time efficiency and robustness

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Local space-time efficiency and robustness

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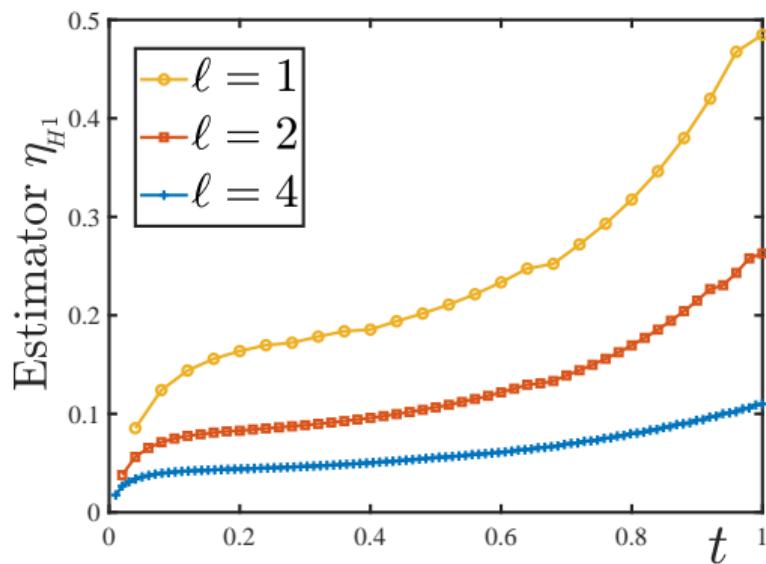
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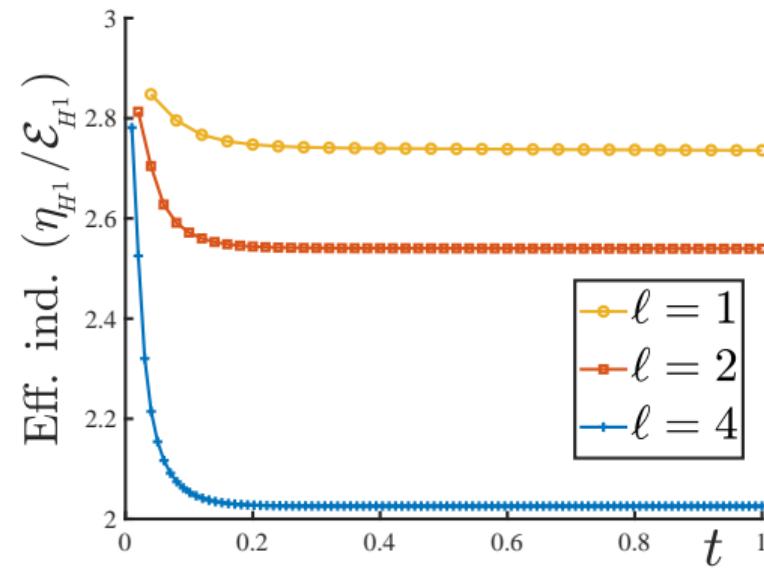
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4 Conclusions and future directions

Richards equation $\partial_t S(u) - \nabla \cdot [\kappa(S(u)) (\nabla u + g)] = f$ (results by K. Mitra, three levels of uniform space-time mesh refinement)

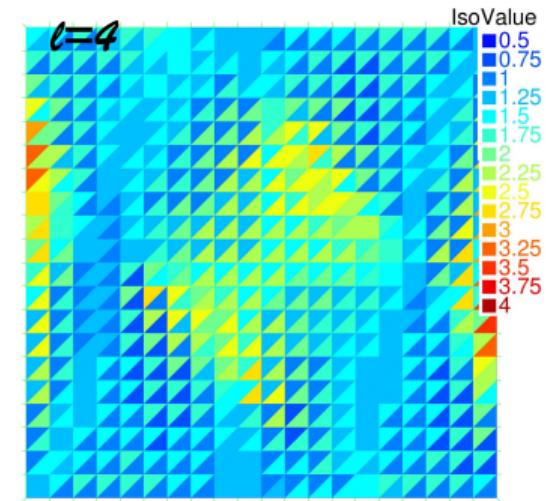
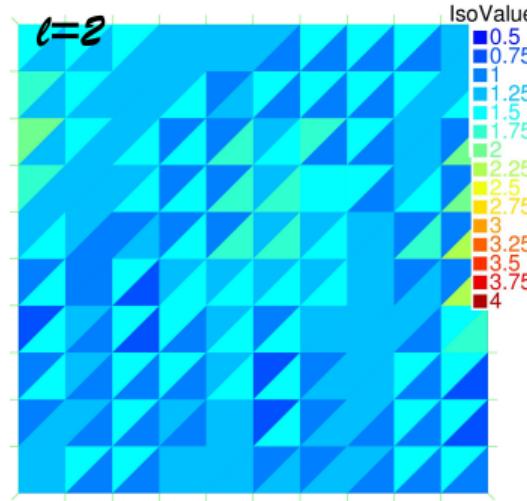
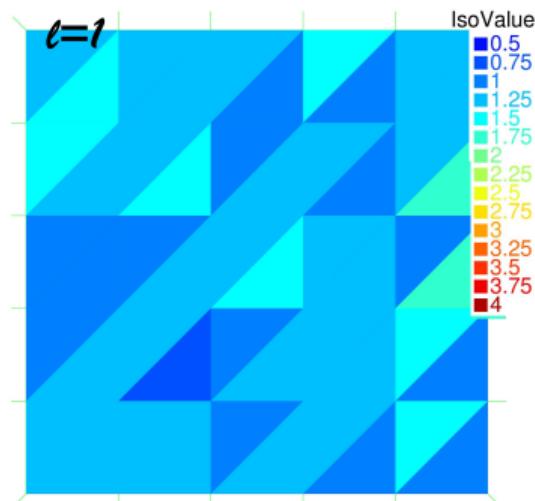


Estimators $\eta(u_{h\tau})$ as a function of T



Effectivity indices $\eta(u_{h\tau})/\|u - u_{h\tau}\|_{\mathcal{E}_Y}$

Richards equation, local space-time effectivity indices



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Conclusions (reaction-diffusion equation)

- ✓ **guaranteed** upper bound
- ✓ local efficiency and **robustness** with respect to reaction and diffusion parameters
- ✓ simple form (no submesh), local estimators minimization, any polynomial degree

Conclusions (heat equation)

- ✓ **guaranteed** upper bound
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Future directions

- nonlinear and coupled problems

Conclusions and future directions

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- ERN A., SMEARS, I., VOHRALÍK M., Guaranteed, locally space-time efficient, and polynomial-degree robust a posteriori error estimates for high-order discretizations of parabolic problems, *SIAM J. Numer. Anal.* **55** (2017), 2811–2834.
- ERN A., SMEARS, I., VOHRALÍK M., Discrete p -robust $\mathbf{H}(\text{div})$ -liftings and a posteriori estimates for elliptic problems with H^{-1} source terms, *Calcolo* **54** (2017), 1009–1025.

Thank you for your attention!

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Thank you for your attention!

Fundamental results on a reference tetrahedron

Bounded right inverse of the divergence operator

- polynomial volume data
- Costabel & McIntosh (2010):

Let $K \in \mathcal{T}$ and $r \in \mathcal{P}_p(K)$. Then there exists $\xi_h \in \mathcal{RTN}_p(K)$ s.t. $\nabla \cdot \xi_h = r$ and

$$\|\xi_h\|_K \leq C \|r\|_{H^{-1}(K)} = \sup_{v \in H_0^1(K), \|\nabla v\|_K=1} (r, v)_K.$$

Polynomial extensions in $H(\text{div})$

- polynomial boundary data
- Demkowicz, Gopalakrishnan, Schöberl (2012):

Let $K \in \mathcal{T}$ and $r \in \mathcal{P}_p(\mathcal{F}_K)$ satisfying $(r, 1)_{\partial K} = 0$. Then there exists $\xi_h \in \mathcal{RTN}_p(K)$ s.t. $\xi_h \cdot \mathbf{n}_K = r$ on ∂K , $\nabla \cdot \xi_h = 0$ in K , and

$$\|\xi_h\|_K \leq C \|r\|_{H^{-1/2}(\partial K)} = \sup_{v \in H^1(K), \|\nabla v\|_K=1} (r, v)_{\partial K}.$$

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General result on a physical tetrahedron

Lemma ($\mathbf{H}(\text{div})$ polynomial extension on a tetrahedron)

Let $K \in \mathcal{T}$, $\mathcal{F}_K^N \subset \mathcal{F}_K$. Let $r \in \mathcal{P}_p(\mathcal{F}_K^N) \times \mathcal{P}_p(K)$, satisfying

$\sum_{F \in \mathcal{F}_K} (r_F, 1)_F = (r_K, 1)_K$ if $\mathcal{F}_K^N = \mathcal{F}_K$. Then for $C = C(\kappa_K) > 0$,

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Context

- $-\Delta \zeta_K = r_K$ in K ,
- $-\nabla \zeta_K \cdot \mathbf{n}_K = r_F$ on all $F \in \mathcal{F}_K^N$,
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Set $\xi_K := -\nabla \zeta_K$.

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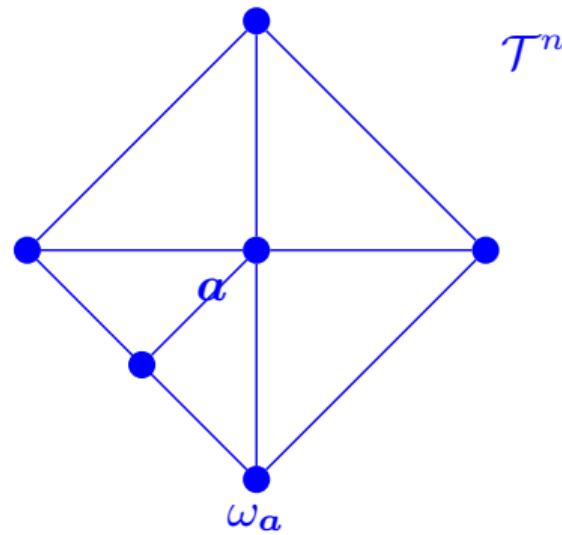
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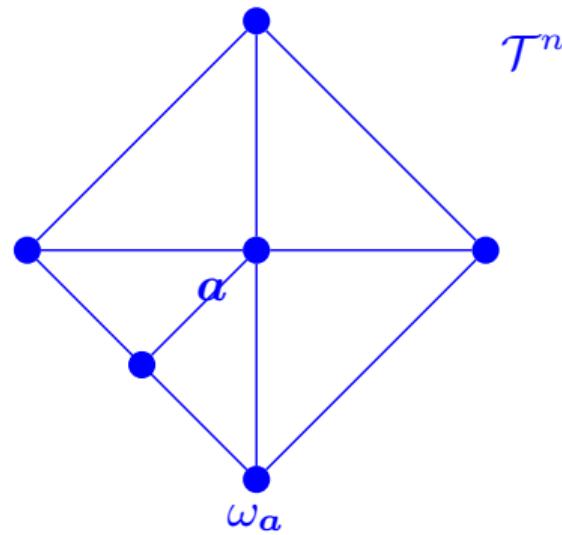
Stable broken $\mathbf{H}(\text{div})$ polynomial extension on a patch

- Braess, Pillwein, & Schöberl (2009), 2D
- Ern & V. (2016), 3D
- Ern, Smears, & V. (2017), 2-3D, patches with subrefinement



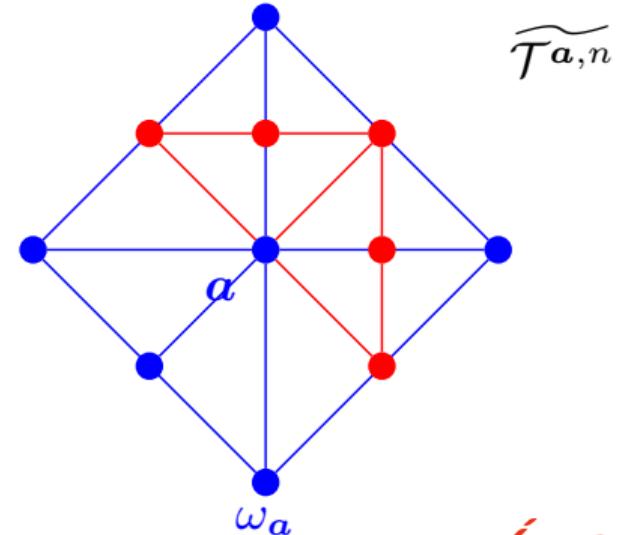
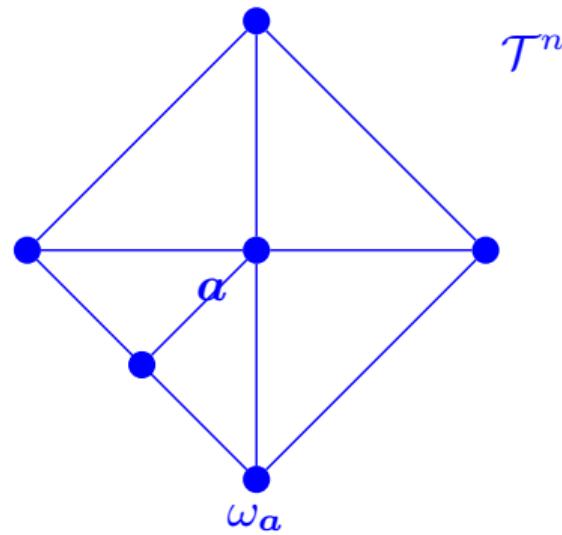
Stable broken $\mathbf{H}(\text{div})$ polynomial extension on a patch

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High-order space-time discretization

CG in space & DG in time

- p -degree **continuous** piecewise polynomials in **space**

$$V_h^n := \{ v_h \in H_0^1(\Omega), \quad v_h|_K \in \mathcal{P}_{p_K}(K) \quad \forall K \in \mathcal{T}^n \}$$

- q -degree **discontinuous** piecewise polynomials in **time**

$$\mathcal{Q}_{q_n}(I_n; V) := \{ V\text{-valued pols of degree at most } q_n \text{ over } I_n \}$$

High-order discretization

Find $u_{h\tau}|_{I_n} \in \mathcal{Q}_{q_n}(I_n; V_h^n)$ with $u_{h\tau}(0) = \Pi_h u_0$ such that

$$\begin{aligned} & \int_{I_n} (\partial_t u_{h\tau}, v_{h\tau}) + (\nabla u_{h\tau}, \nabla v_{h\tau}) dt - ((u_{h\tau})_{n-1}, v_{h\tau}(t_{n-1}^+)) \\ &= \int_{I_n} (f, v_{h\tau}) dt \quad \forall v_{h\tau} \in \mathcal{Q}_{q_n}(I_n; V_h^n) \quad \forall 1 \leq n \leq N. \end{aligned}$$

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Augmented norm

- augment the norm: $\|v\|_{\mathcal{E}_Y}^2 := \|\mathcal{I}v\|_Y^2 + \|v - \mathcal{I}v\|_X^2$, $v \in Y + V_{h\tau}$
- $\mathcal{I}u = u \Rightarrow$

$$\|u - u_{h\tau}\|_{\mathcal{E}_Y}^2 = \|u - \mathcal{I}u_{h\tau}\|_Y^2 + \underbrace{\|u_{h\tau} - \mathcal{I}u_{h\tau}\|_X^2}_{\text{known, computable}}$$

- we are adding to Y norm the time jumps in X norm (Schötzau–Wihler):

$$\begin{aligned} \|u_{h\tau} - \mathcal{I}u_{h\tau}\|_{X(I_n)}^2 &= \int_{I_n} \|\nabla(u_{h\tau} - \mathcal{I}u_{h\tau})\|^2 dt \\ &= \frac{\tau_n(q_n+1)}{(2q_n+1)(2q_n+3)} \|\nabla(u_{h\tau})_{n-1}\|^2 \end{aligned}$$

Equivalence between the γ and \mathcal{E}_γ norms

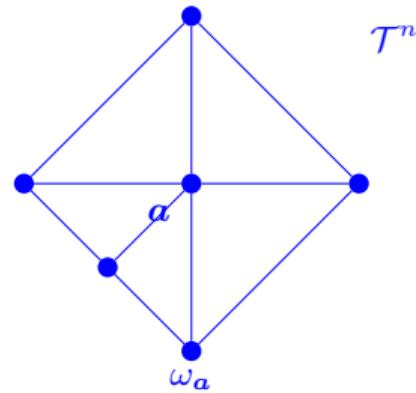
Theorem (Global equivalence)

Suppose *no source term oscillation or no coarsening*. Then there holds

$$\|u - \mathcal{I}u_{h\tau}\|_\gamma \leq \|u - u_{h\tau}\|_{\mathcal{E}_\gamma} \leq 3\|u - \mathcal{I}u_{h\tau}\|_\gamma$$

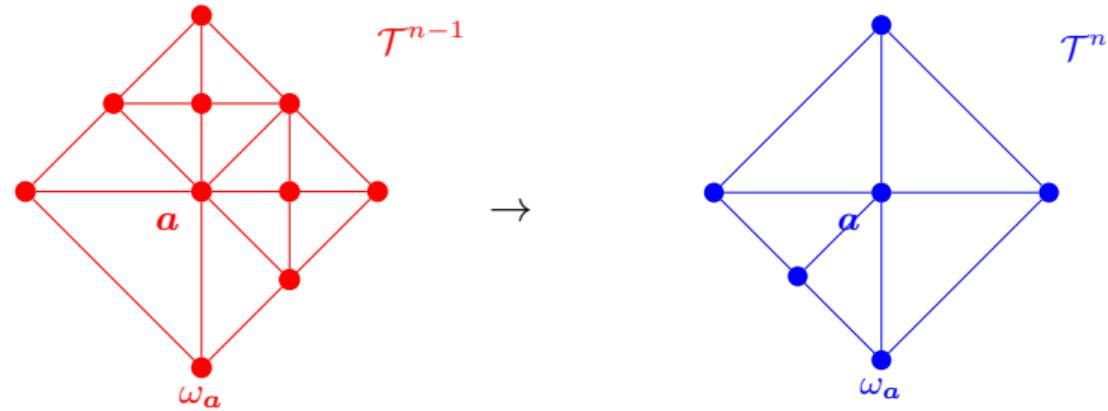
- the two norms $\|\cdot\|_\gamma$ and $\|\cdot\|_{\mathcal{E}_\gamma}$ still may differ locally
- in general, an additional source term oscillation or coarsening term appears

Handling mesh adaptivity



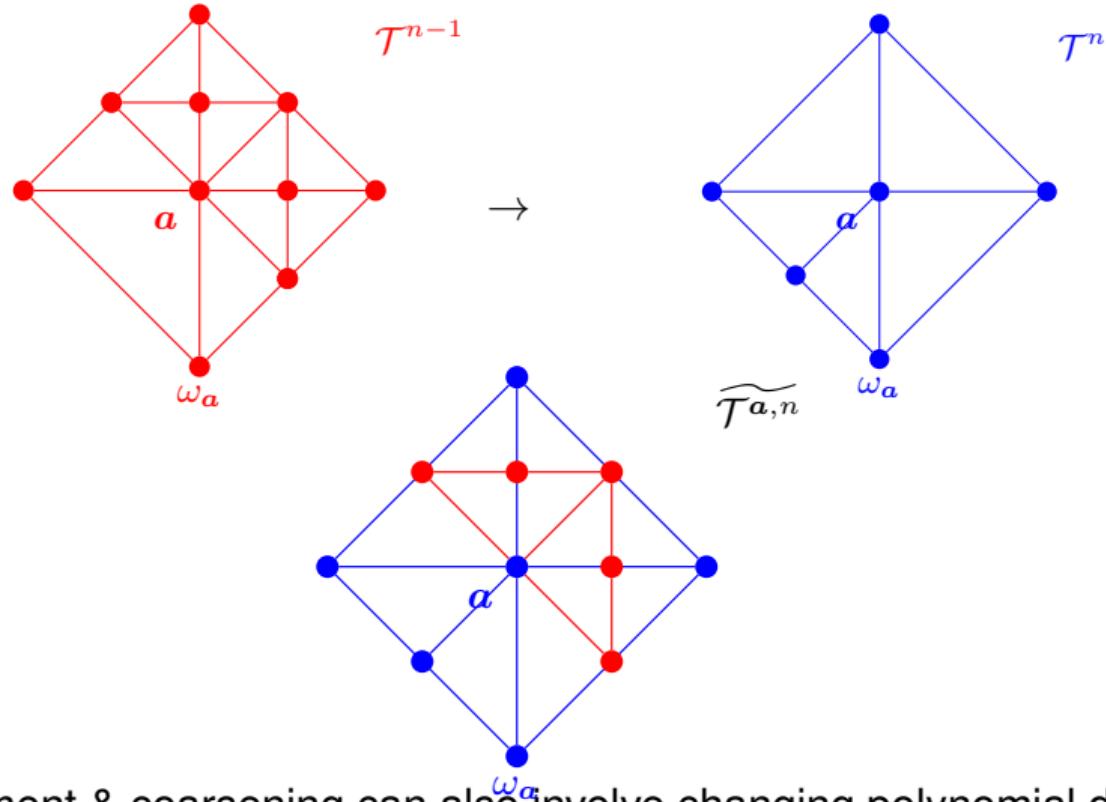
- refinement & coarsening can also involve changing polynomial degrees

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