

Adaptive iterative approximation in nonlinear PDEs I

Martin Vohralík

in collaboration with

Patrik Daniel, Vít Dolejší, Alexandre Ern, Alexander Haberl, Koondanibha Mitra, Jan Papež, Dirk Praetorius,
Ulrich Rüde, Stefan Schimanko, & Barbara Wohlmuth

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INSTITUT
POLYTECHNIQUE
DE PARIS

Outline

- 1 Introduction: adaptive iterative approximation
 - A posteriori error estimates and adaptivity
 - Achievements and example results
 - Real life comparison
- 2 Linear diffusion: discretization error, mesh and polynomial degree adaptivity
 - A posteriori error estimates
 - Potential reconstruction
 - Flux reconstruction
 - A posteriori error control
 - Balancing error components: mesh adaptivity
 - Balancing error components: polynomial-degree adaptivity
- 3 Nonlinear diffusion: overall error and solvers adaptivity
 - A posteriori error estimates (overall and components)
 - Balancing error components: solvers adaptivity
- 4 Conclusions

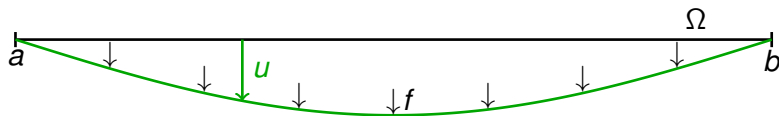
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Numerical methods for PDEs: example $-\Delta u = f$ in Ω , $u = 0$ on $\partial\Omega$

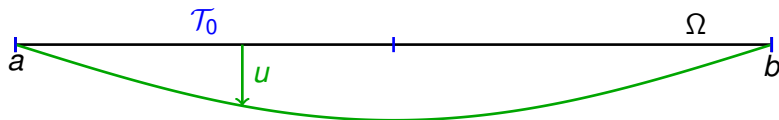
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- unknown exact solution u



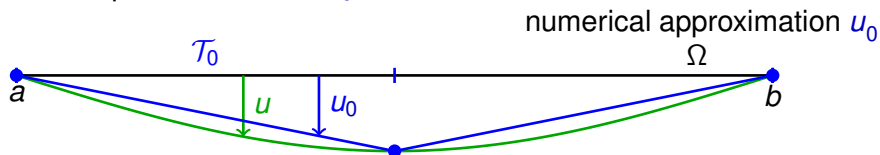
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- computational mesh $\mathcal{T}_0$



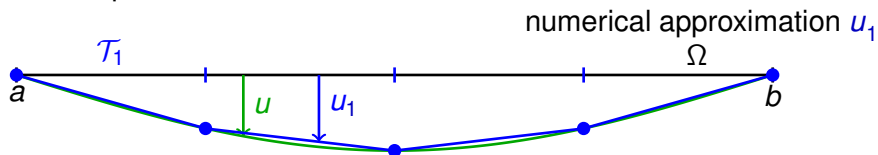
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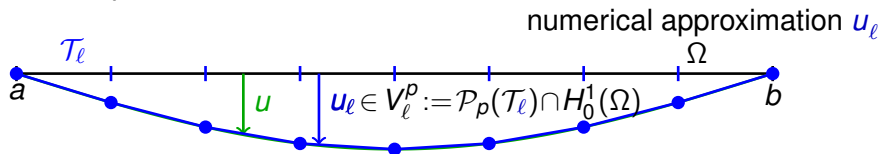
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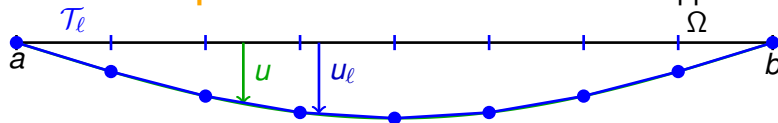
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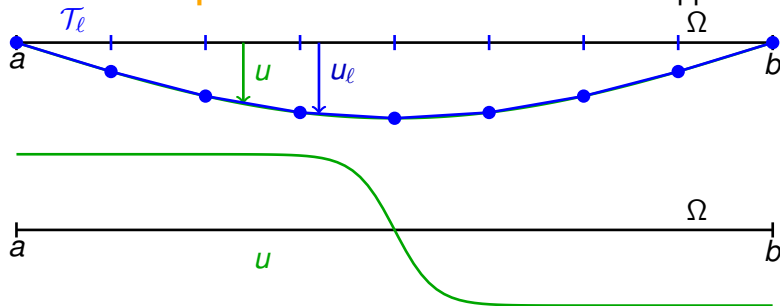
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- unknown exact solution u
- computational mesh $\mathcal{T}_\ell$
- **more computational resources** $\Rightarrow$ numerical approximation u_ℓ **closer** to u



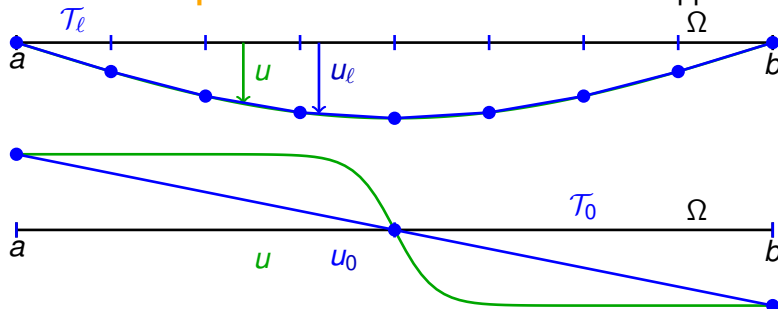
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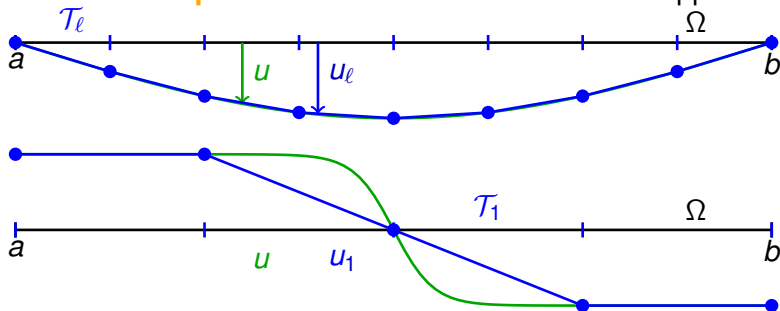
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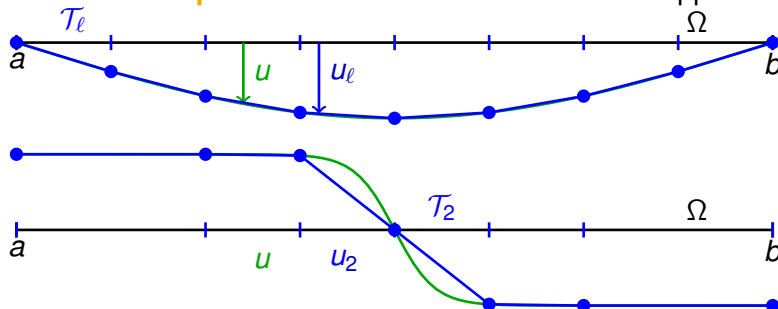
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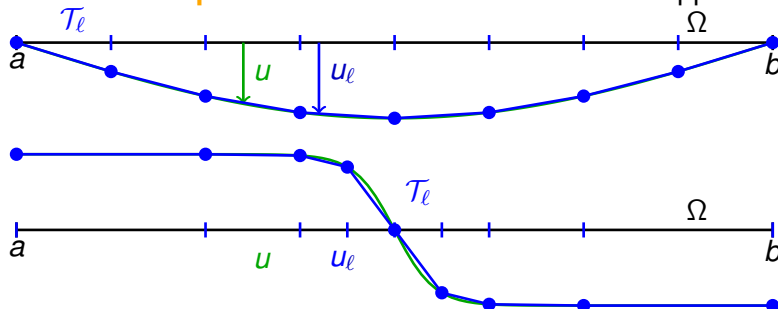
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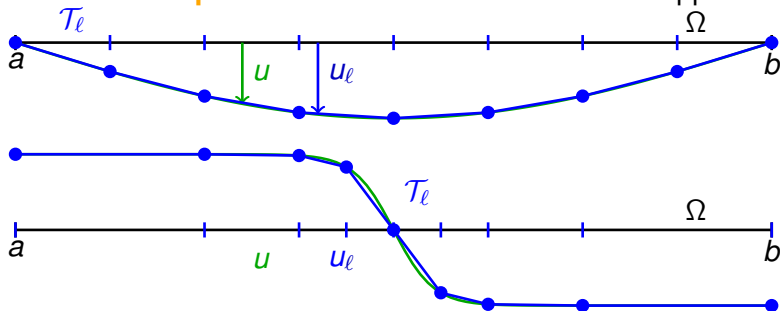
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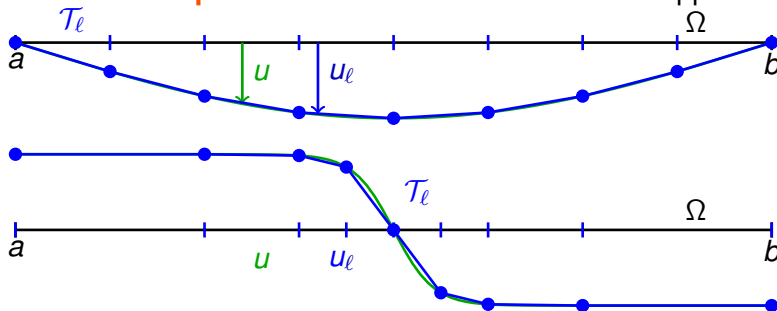
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- unknown exact solution u
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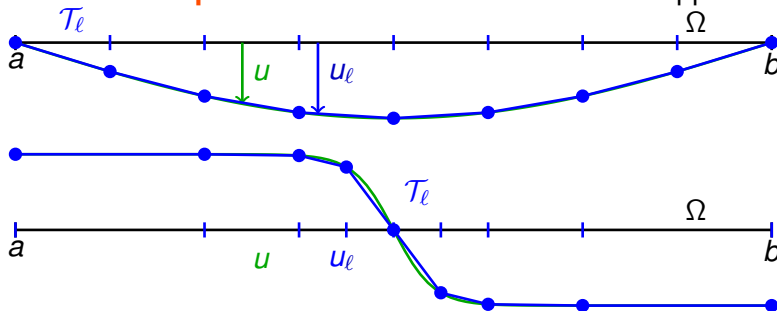


Need to solve

$$\mathbb{A}_\ell \mathbf{U}_\ell = \mathbf{F}_\ell$$

Numerical methods for PDEs: example $-\Delta u = f$ in Ω , $u = 0$ on $\partial\Omega$

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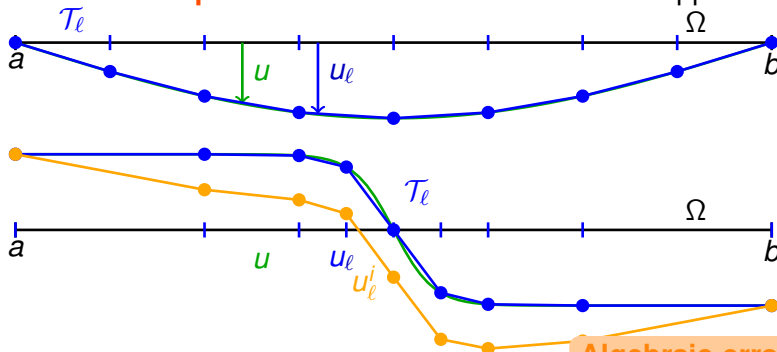
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Algebraic error (on iteration $i \geq 0$)

$$\|\nabla(u_\ell - u_\ell^i)\| = \left\{ \int_a^b |(u_\ell - u_\ell^i)'|^2 \right\}^{\frac{1}{2}}$$

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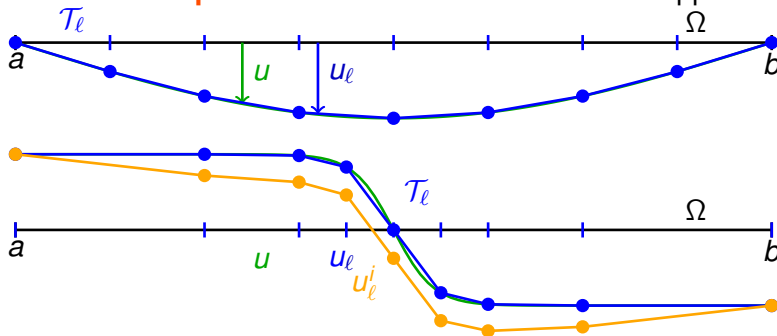
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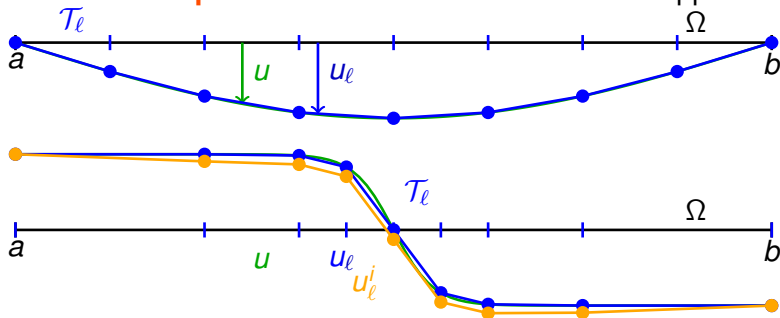
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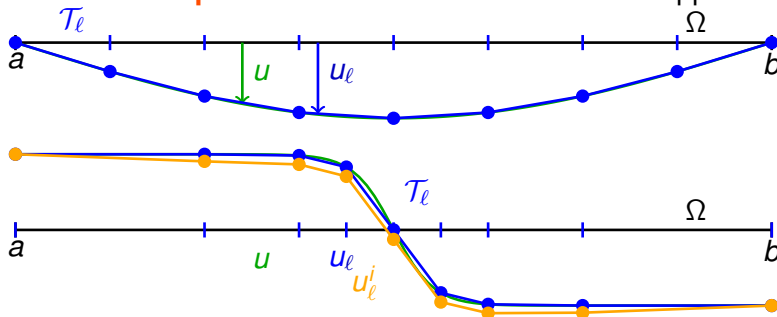
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Need to solve

$$\mathbb{A}_\ell \mathbf{U}_\ell = \mathbf{F}_\ell$$

Total error

$$\|\nabla(u - u_\ell^i)\| = \left\{ \int_a^b |(u - u_\ell^i)'|^2 \right\}^{\frac{1}{2}}$$

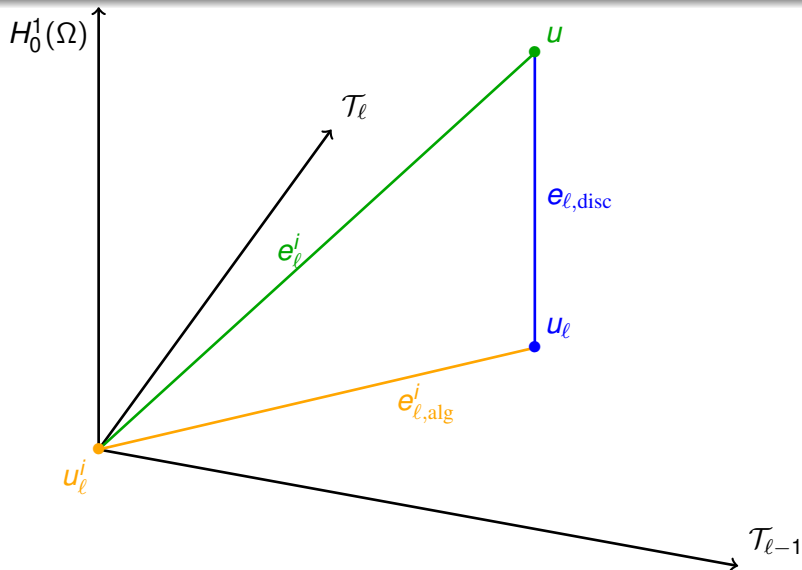
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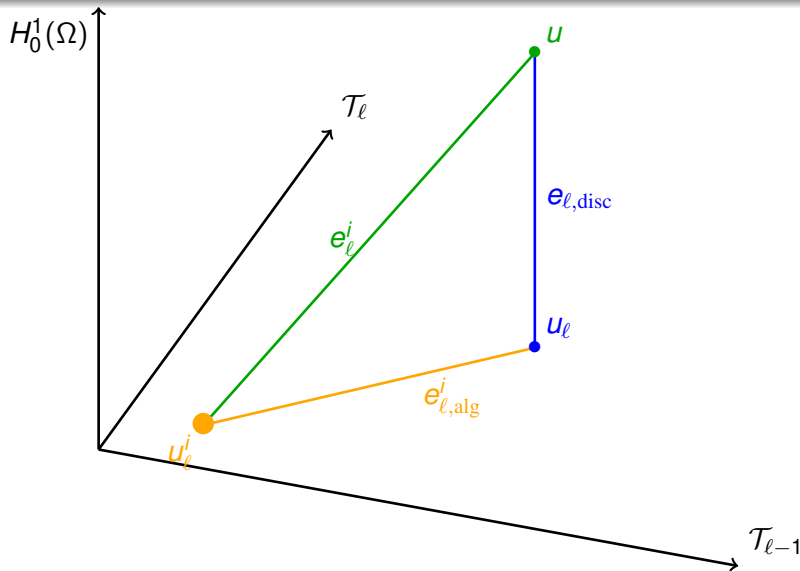
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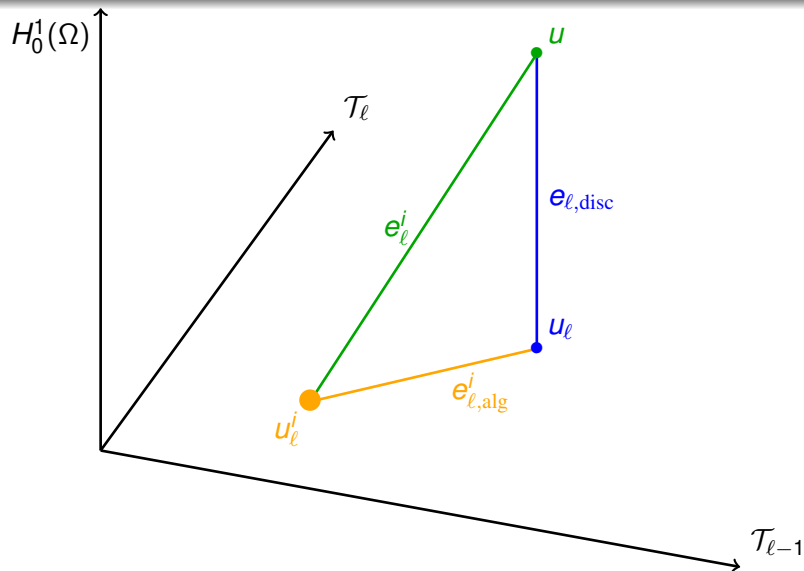
Adaptive iterative approximation



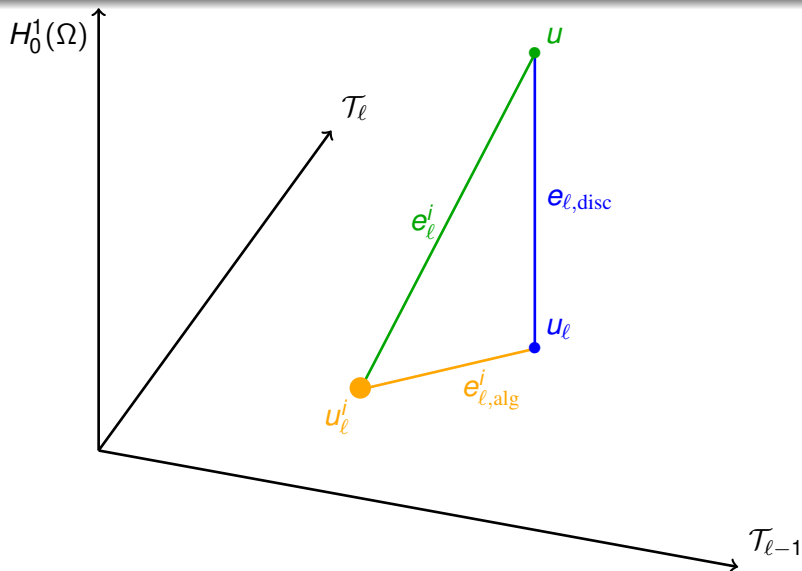
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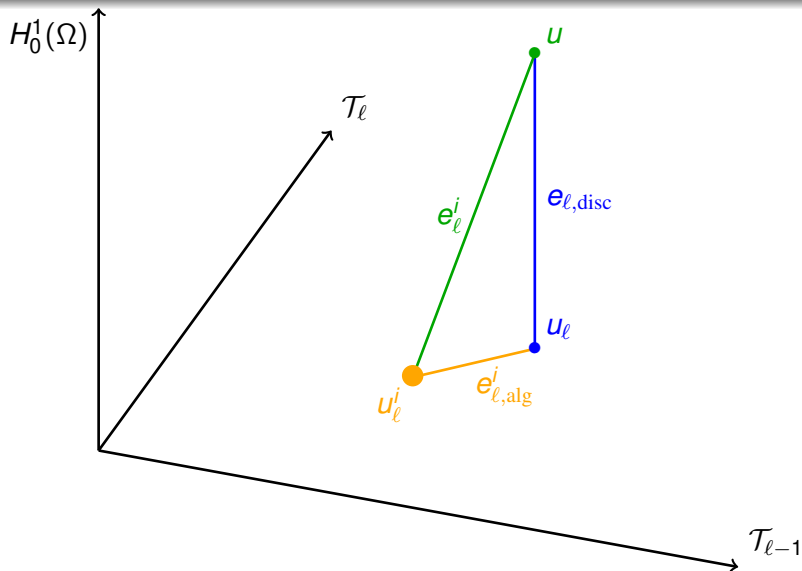
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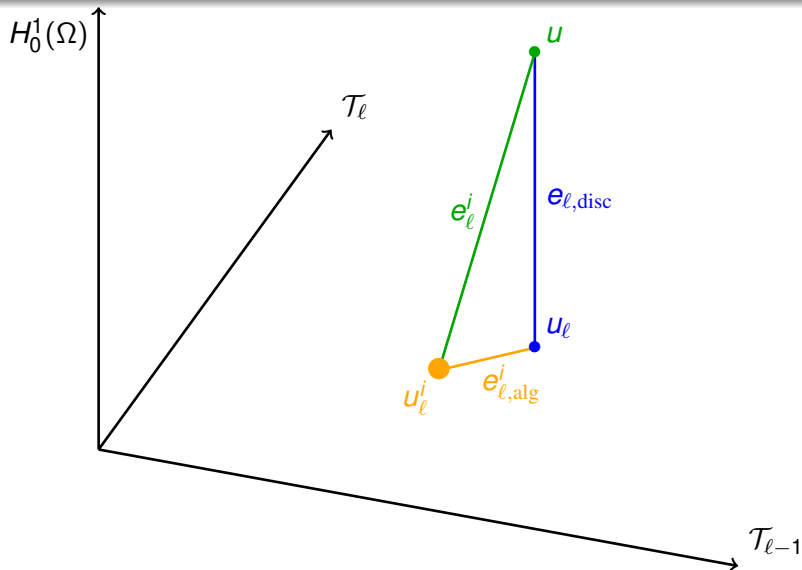
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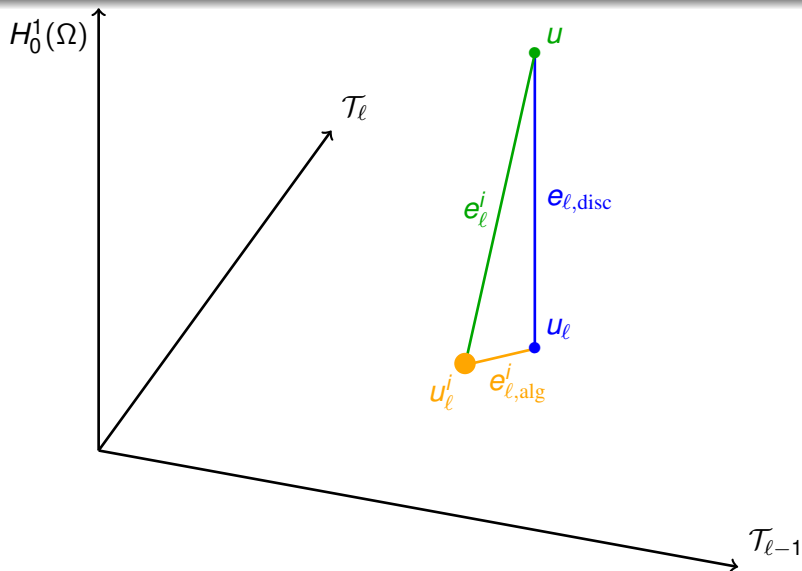
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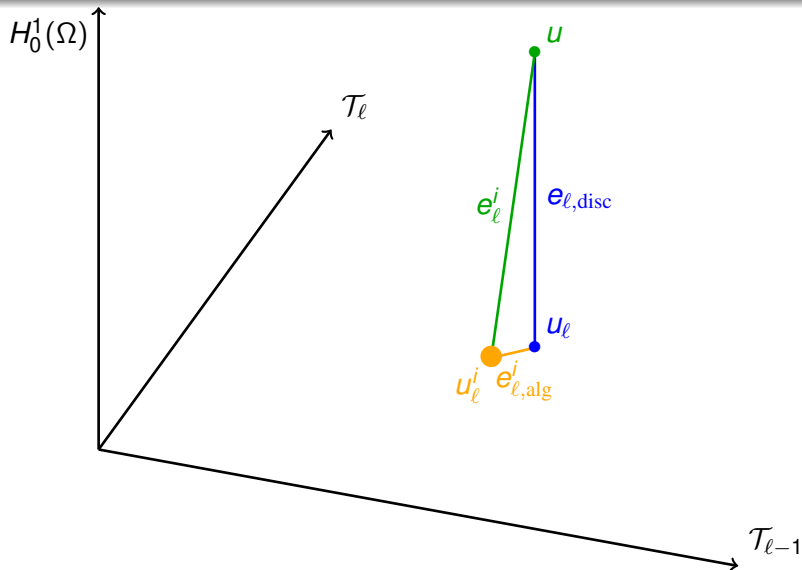
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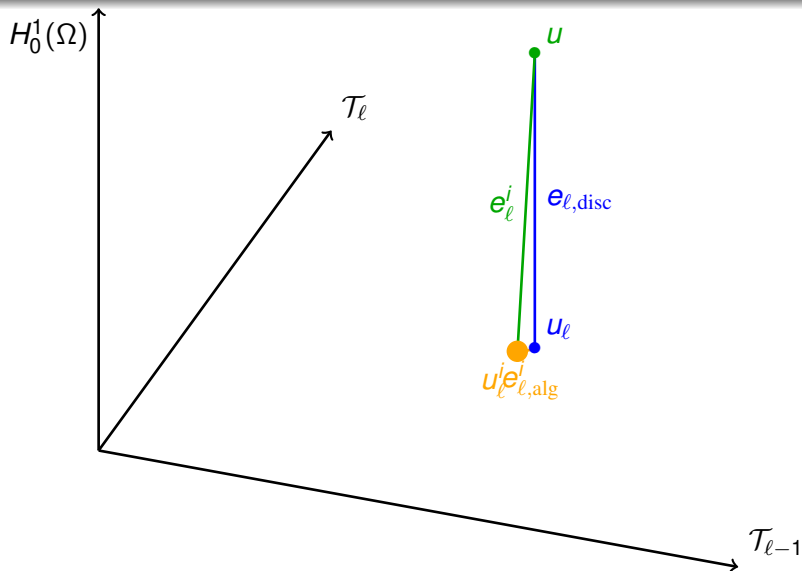
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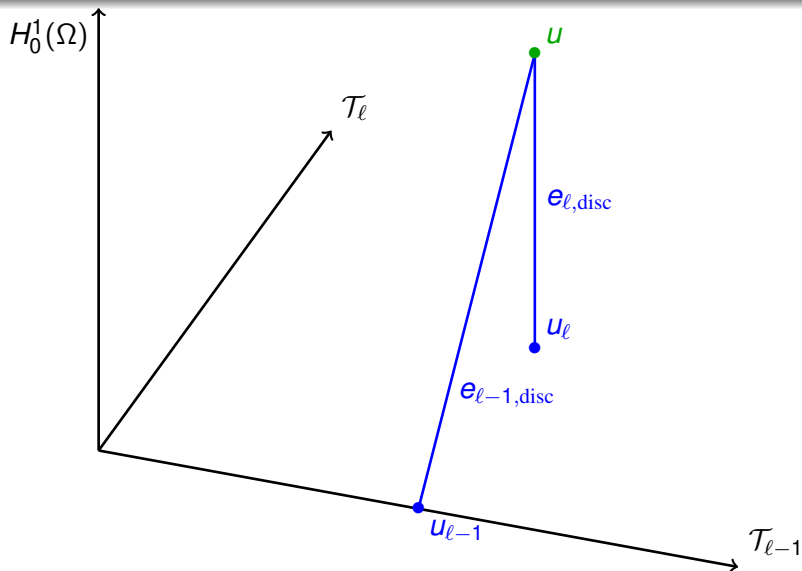
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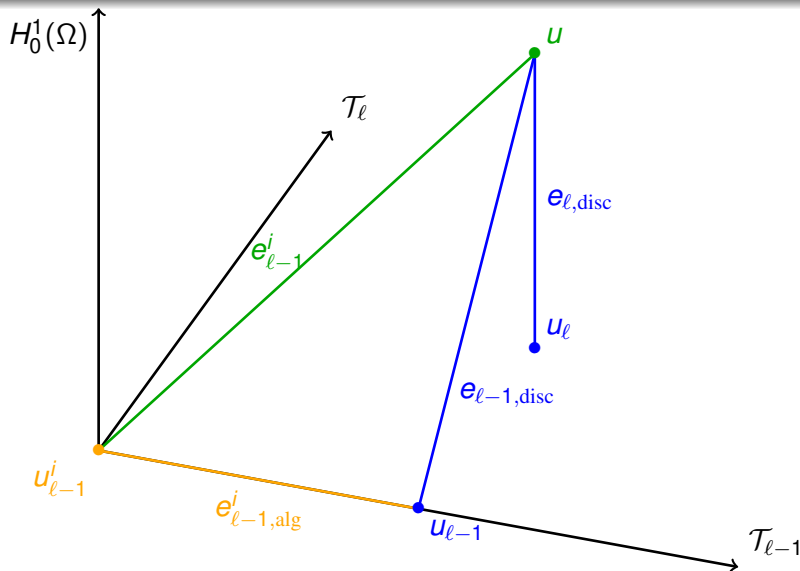
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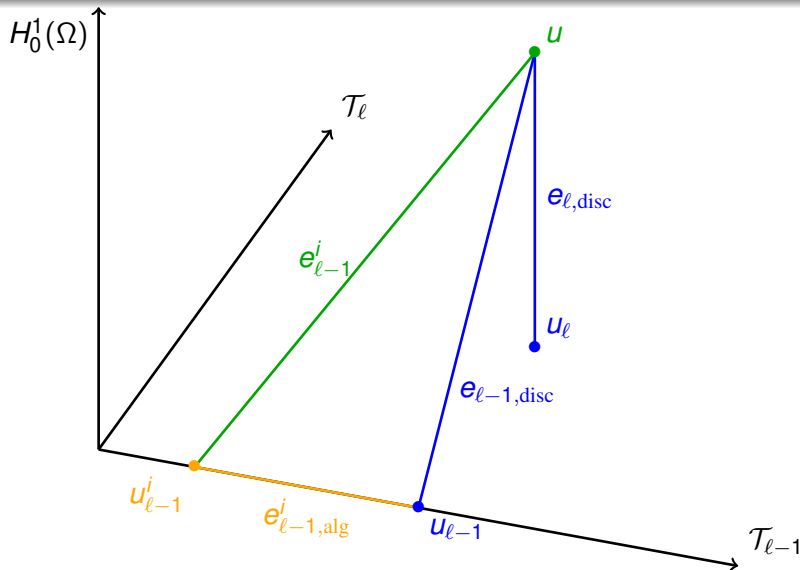
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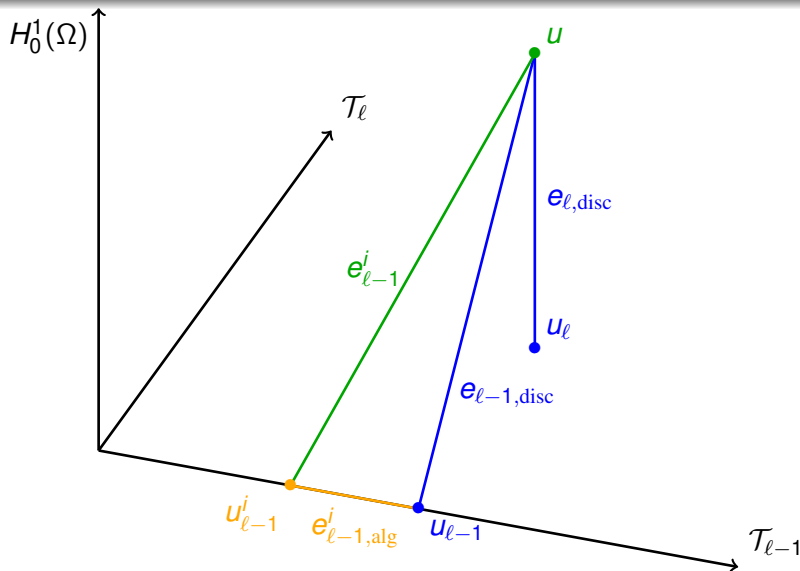
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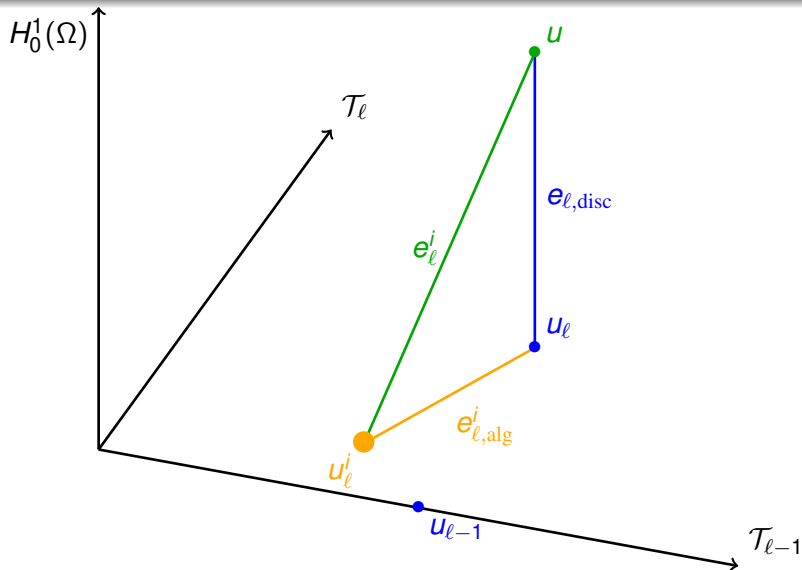
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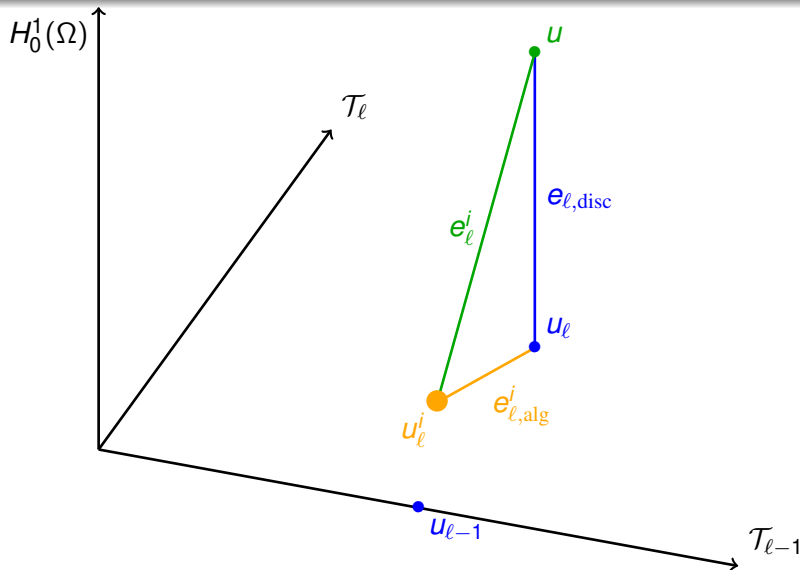
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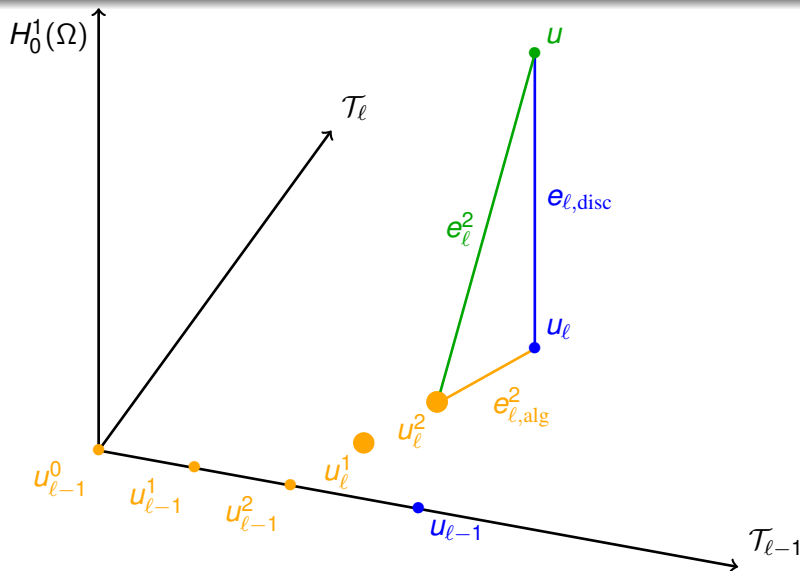
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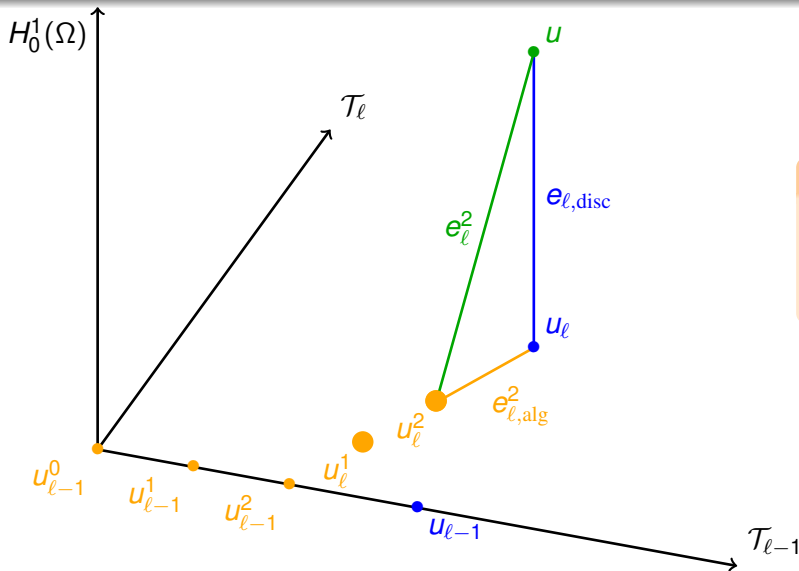
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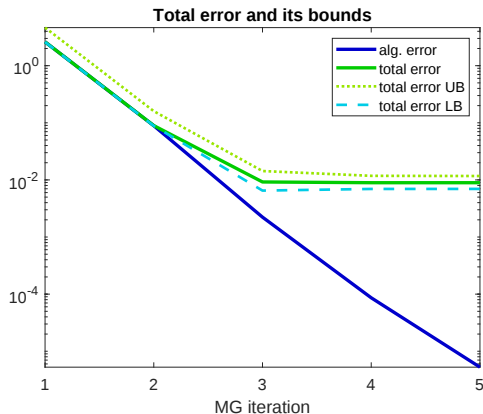
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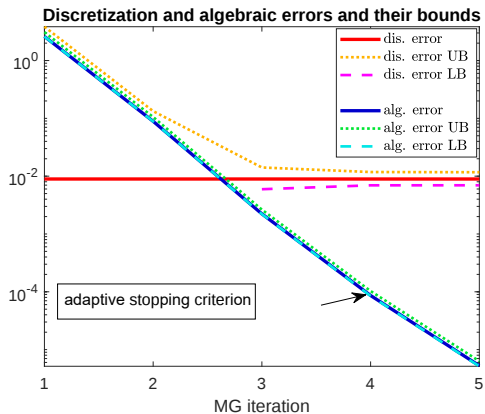
Cost

$$\sum_{\ell=0}^{\bar{\ell}} \sum_{i(\ell)=0}^{\bar{i}(\ell)} |V_{\ell}^p|$$

A posteriori estimates of total and algebraic errors: $\mathbb{A}_\ell U_\ell^i \neq F_\ell$



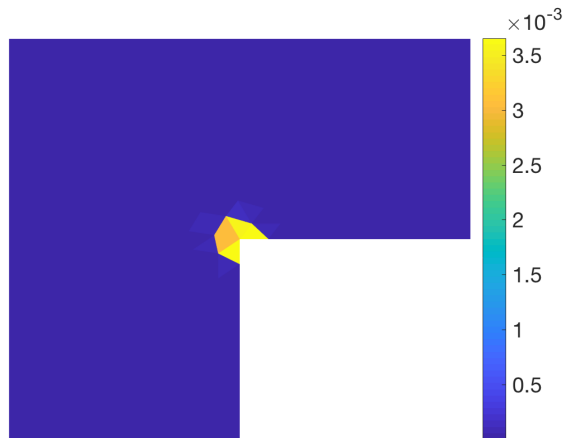
Total error



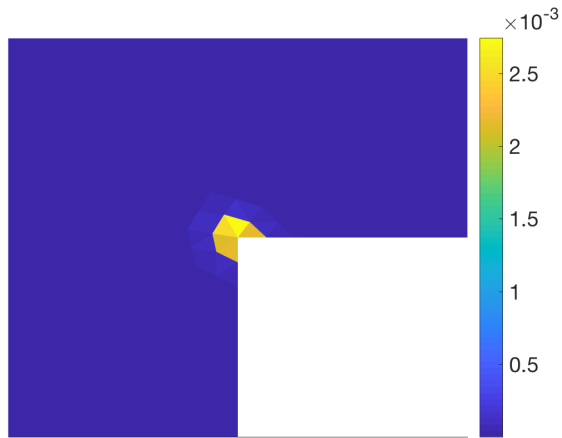
Error components and adaptive st. crit.

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

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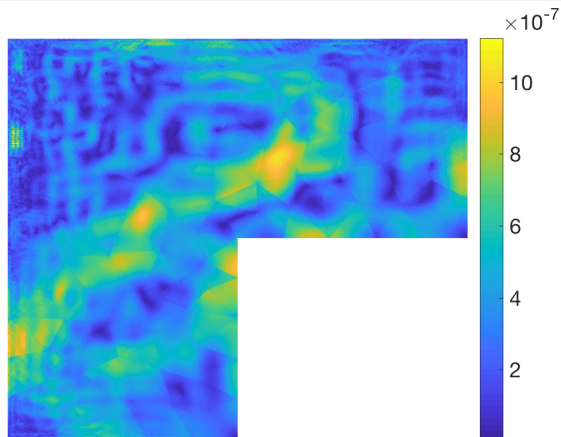
Estimated total errors $\eta_K(u_\ell^i)$



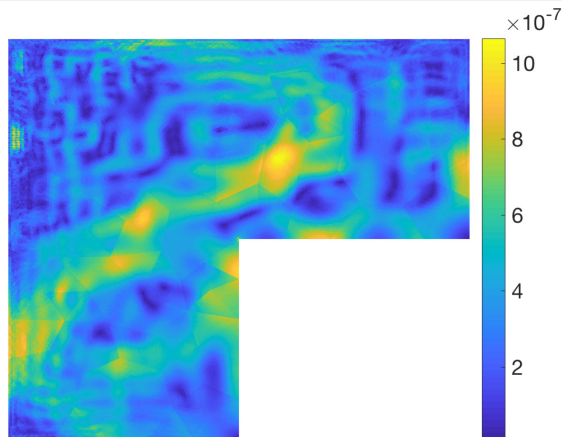
Exact total errors $\|\nabla(u - u_\ell^i)\|_K$

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Estimated algebraic errors $\eta_{\text{alg},K}(u_\ell^i)$



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Numerical approximations of nonlinear PDEs:

Setting

- u : unknown exact PDE solution
- u_ℓ : known numerical approximation on mesh $\mathcal{T}_\ell$

Numerical approximations of nonlinear PDEs:

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- u : unknown exact PDE solution
- $u_\ell^{k,i}$: known numerical approximation on mesh $\mathcal{T}_\ell$, linearization step k , and linear solver step i

Numerical approximations of nonlinear PDEs: 3 crucial questions

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Crucial questions

- 1 How **large** is the overall **error** between u and $u_\ell^{k,i}$?

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- 3 Can we **decrease** it **efficiently**?

Numerical approximations of nonlinear PDEs: 3 crucial questions & suggested answers

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Suggested answers

- 1 Computable **a posteriori** error **estimates**.

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- 1 Computable **a posteriori error estimates**.
- 2 Identification of **error components**.

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Suggested answers

- 1 Computable **a posteriori** error **estimates**.
- 2 Identification of **error components**.
- 3 **Balancing** error components, **adaptivity** (working where needed).

Main achievements

A posteriori error estimates

a posteriori error estimates

$$|||u - u_\ell^{k,i}||| \leq \eta(u_\ell^{k,i})$$

Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates

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Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates

efficient

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

Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates
respect to **strength of nonlinearities**

efficient, robust with re-

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

Main achievements

A posteriori error estimates

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

Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates, **locally efficient**, **robust** with respect to **strength of nonlinearities**

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

$$\eta_K (u_\ell^{k,i}) \leq C_{\text{eff}} \|u - u_\ell^{k,i}\|_{\omega_K} \text{ for all } K \in \mathcal{T}_\ell,$$

Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates, **locally efficient**, **robust** with respect to **strength of nonlinearities**, and identifying **error components**.

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

$$\eta_K(u_\ell^{k,i}) \leq C_{\text{eff}} \|u - u_\ell^{k,i}\|_{\omega_K} \text{ for all } K \in \mathcal{T}_\ell,$$

$$\eta_K(u_\ell^{k,i}) = \eta_{\text{disc},K}(u_\ell^{k,i}) + \eta_{\text{lin},K}(u_\ell^{k,i}) + \eta_{\text{alg},K}(u_\ell^{k,i}).$$

Main achievements

A posteriori error estimates

Guaranteed a posteriori error estimates, **locally efficient**, **robust** with respect to **strength of nonlinearities**, and identifying **error components**.

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

$$\eta_K(u_\ell^{k,i}) \leq C_{\text{eff}} \|u - u_\ell^{k,i}\|_{\omega_K} \text{ for all } K \in \mathcal{T}_\ell,$$

$$\eta_K(u_\ell^{k,i}) = \eta_{\text{disc},K}(u_\ell^{k,i}) + \eta_{\text{lin},K}(u_\ell^{k,i}) + \eta_{\text{alg},K}(u_\ell^{k,i}).$$

Optimal decay rate of error wrt computational cost

Adaptive iterative approximation allows to achieve

$$\|u - u_\ell^{k,i}\| \lesssim \left\{ \sum_{\ell=0}^{\bar{\ell}} \sum_{k(\ell)=0}^{\bar{k}(\ell)} \sum_{i(\ell,k)=0}^{\bar{i}(\ell,k)} |V_\ell^p| \right\}^{-p/d}.$$

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Holistic approach: interplay PDE–numerics–linearization–algebra

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Algorithm

Algorithm 1 Adaptive iterative approximation

```

while  $\frac{\eta(u_\ell^{k,i})}{|||u_\ell^{k,i}|||} > \gamma_{\text{tot}}$  do                                ▷ while user-specified relative precision is not reached
    assemble the finite element space  $V_\ell^p$ 
    assemble the nonlinear problem
    while  $\eta_{\text{lin},K}(u_\ell^{k,i}) > \gamma_{\text{lin}} \eta_{\text{disc},K}(u_\ell^{k,i}) \ \forall K \in \mathcal{T}_\ell$  do    ▷ while linearization dominates discretization
        assemble the linearized problem
        while  $\eta_{\text{alg},K}(u_\ell^{k,i}) > \gamma_{\text{alg}} \eta_{\text{lin},K}(u_\ell^{k,i}) \ \forall K \in \mathcal{T}_\ell$  do    ▷ while algebra dominates linearization
            one step of iterative algebraic solver,  $i++$ 
        end while (iterative algebraic solver)
         $k++$ 
    end while (iterative linearization)
    refine  $K \in \mathcal{T}_\ell$  with big  $\eta_{\text{disc},K}(u_\ell^{k,i})$ ,  $\ell++$ 
end while (adaptive mesh refinement)
  
```

Implementation and observations

Error components

- $\eta_{\text{disc},K}(u_\ell^{k,i})$: discretization
- $\eta_{\text{lin},K}(u_\ell^{k,i})$: linearization
- $\eta_{\text{alg},K}(u_\ell^{k,i})$: algebraic solver

Implementation and observations

Error control

- at **any moment** during the simulation
- price: local finite element quadrature/sparse **matrix-vector** multiplication

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- **same physical units** of all component estimators
- **balance** all component estimators
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- **easy implementation** into existing codes

Optimal decay rate wrt degrees of freedom & computational cost

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classical

estimated error

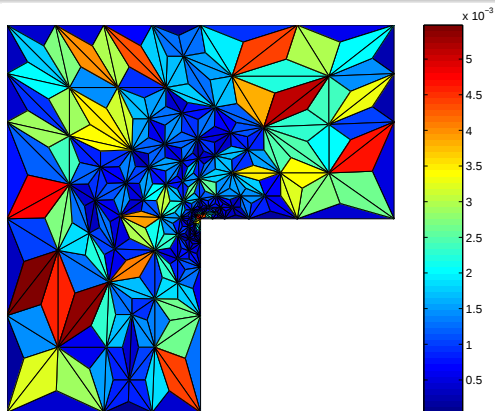
4.6%

Optimal decay rate wrt degrees of freedom & computational cost

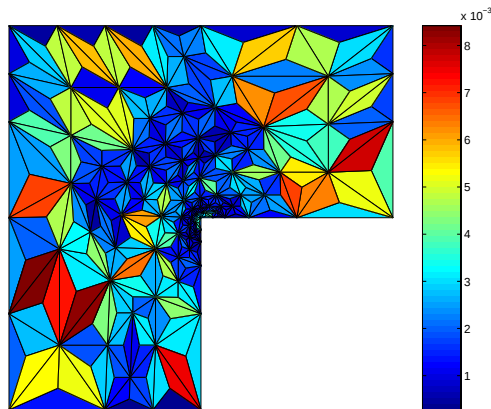
classical

tot. algebraic solver iter.	10890
estimated error	4.6%

Optimal decay rate wrt degrees of freedom & computational cost



Estimated local total errors $\eta_K(u_\ell^{k,i})$

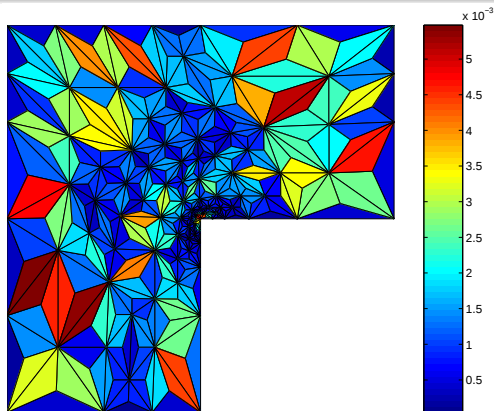


Exact local total errors

classical

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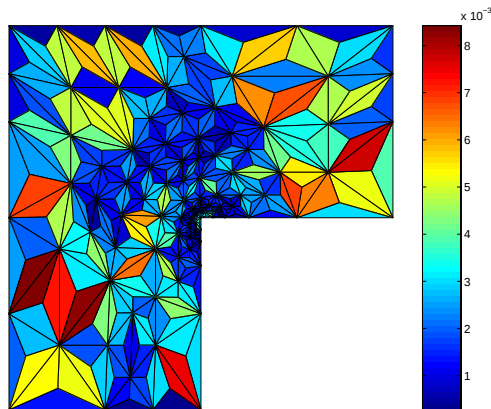
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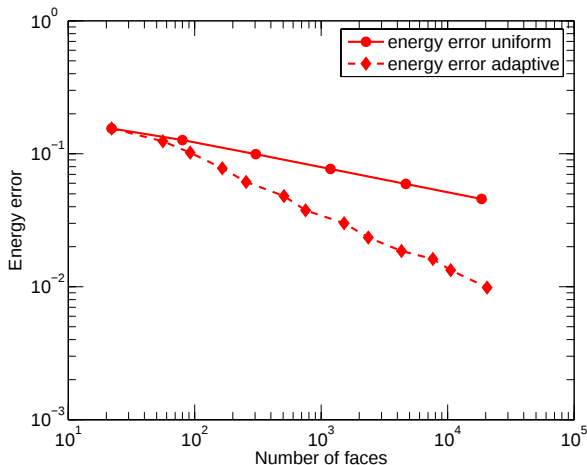


Exact local total errors

adaptive

tot. algebraic solver iter.	242
estimated error	1.1%

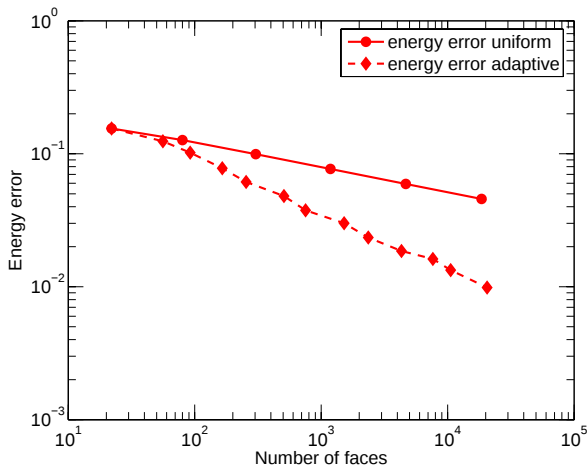
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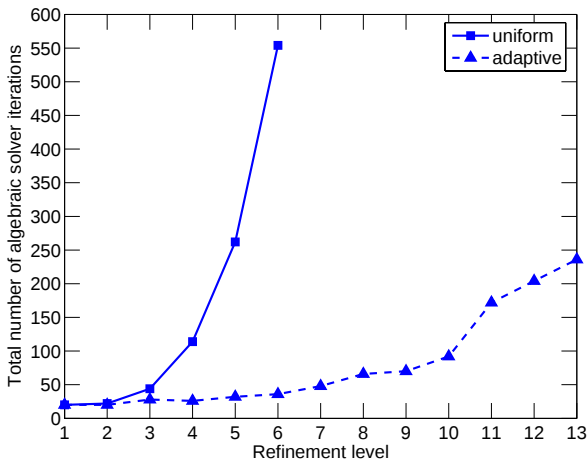
Optimal decay rate wrt DoFs

A. Ern, M. Vohralík, SIAM Journal on Scientific Computing (2013)

Optimal decay rate wrt degrees of freedom & computational cost



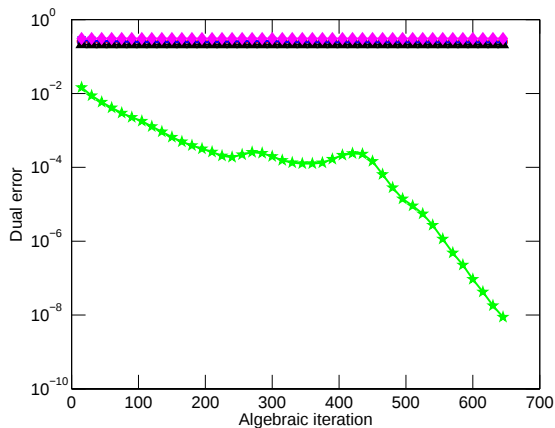
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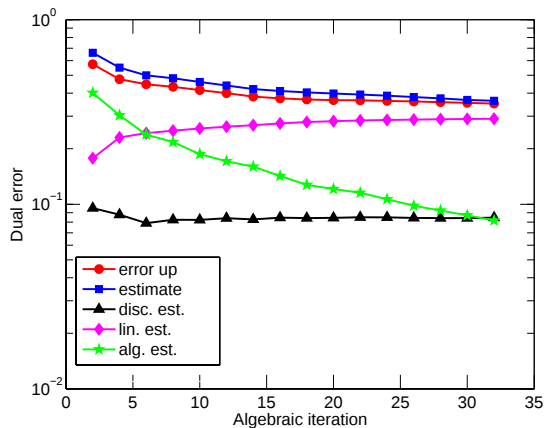
Optimal computational cost

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Total, linearization, & alg. errors: $(\mathcal{A}_\ell(U_\ell^{k,i}) \neq F_\ell, \mathbb{A}_\ell^{k-1}U_\ell^{k,i} \neq F_\ell^{k-1})$

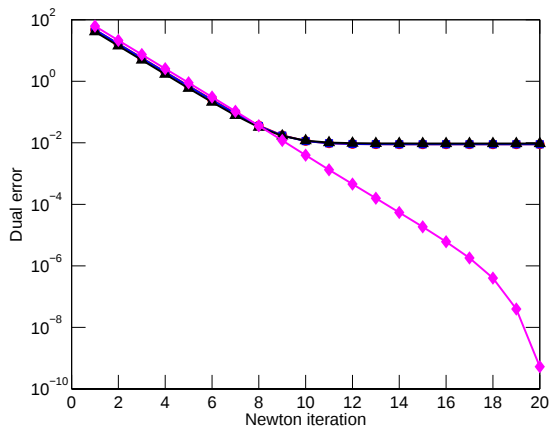


Newton

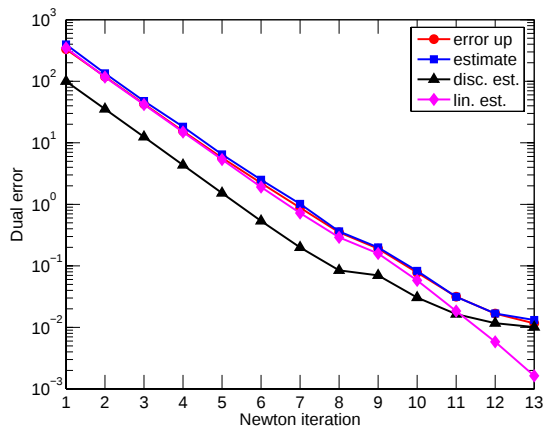


adaptive inexact Newton

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adaptive inexact Newton

Outline

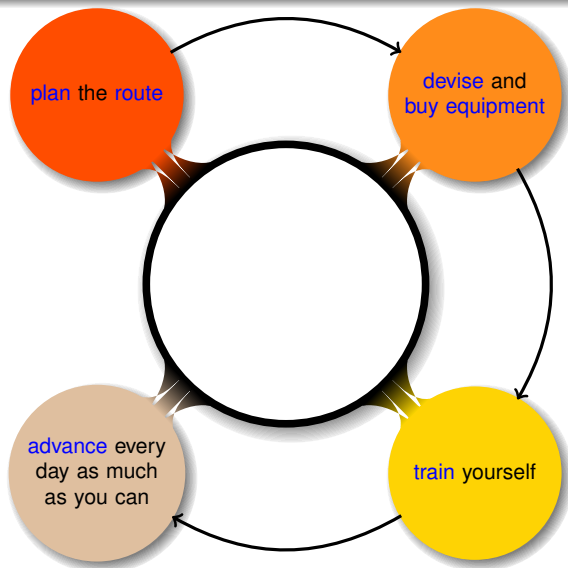
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Control the error and act adaptively: real life

Control the error and act adaptively: real life

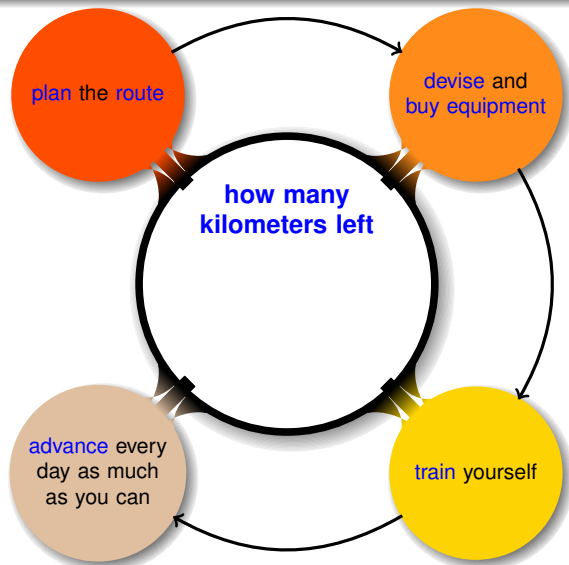
wandering Paris–Santiago
de Compostela

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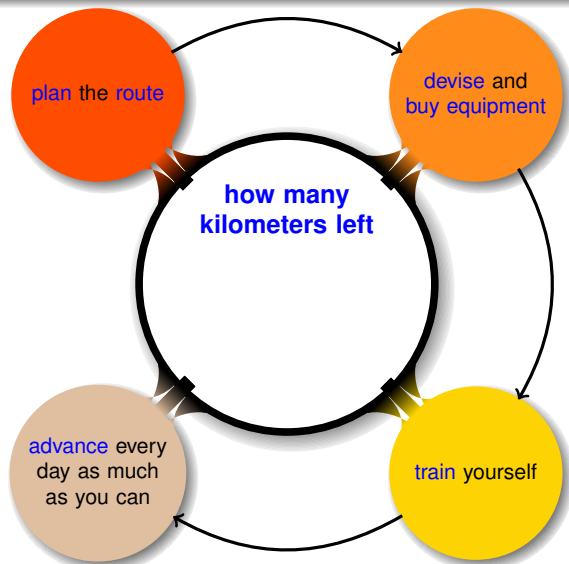


wandering Paris–Santiago
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control the error

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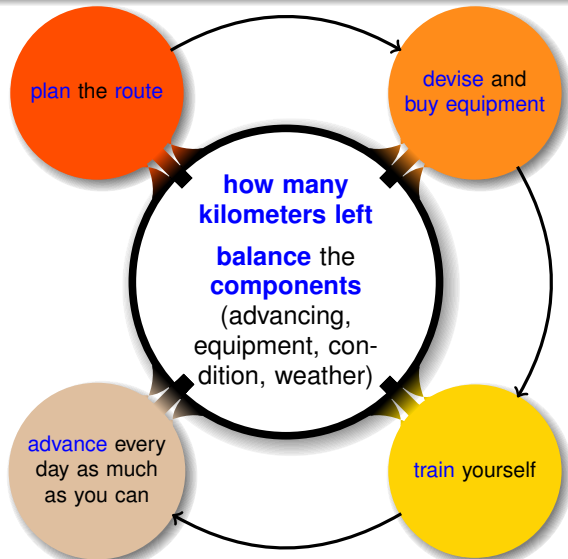


control the error

possible since

- target known
-

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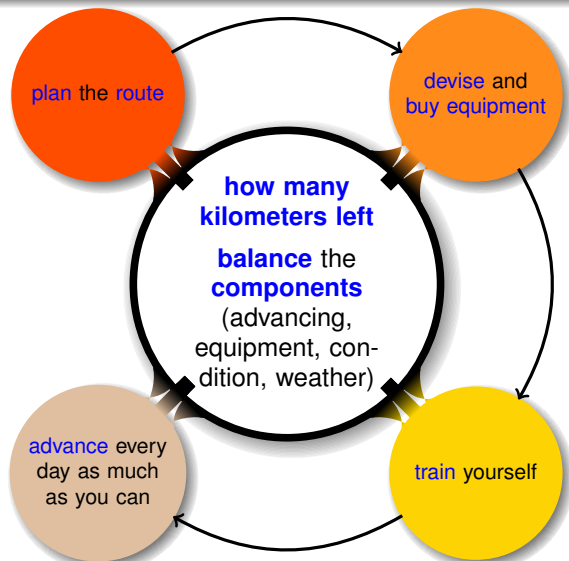


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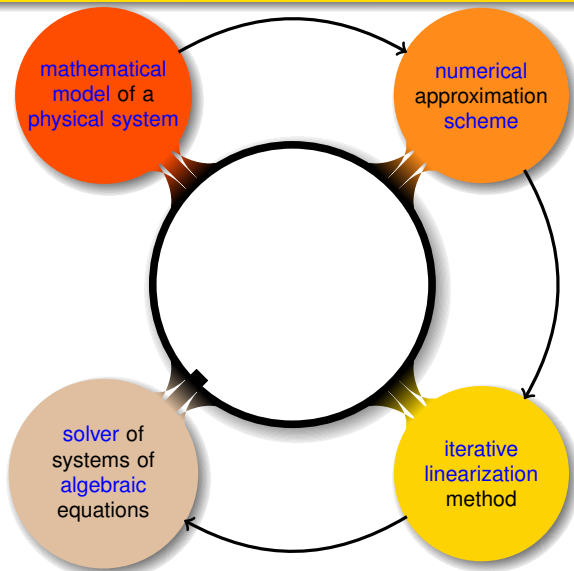
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Control the error and act adaptively: numerical simulations

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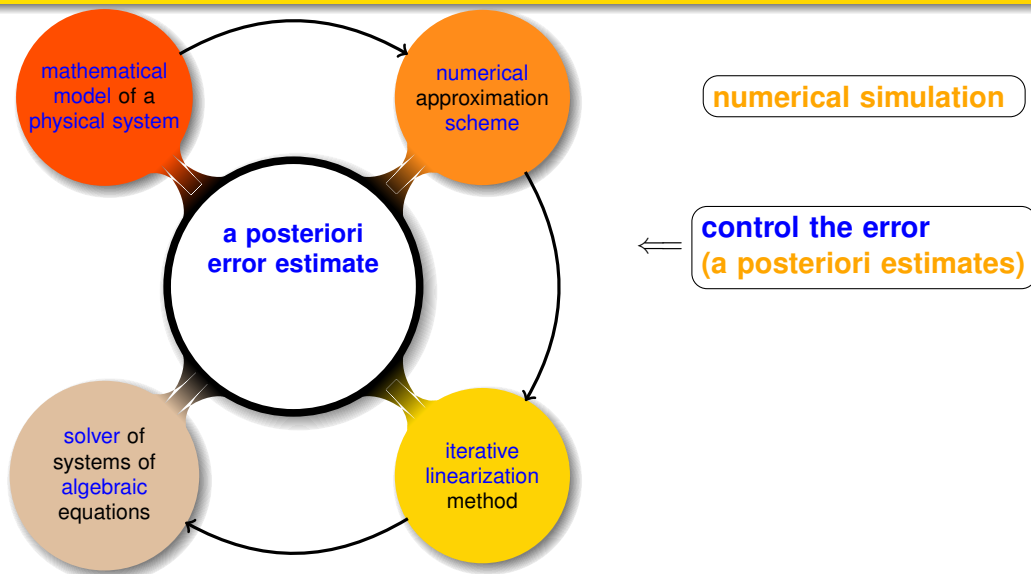
numerical simulation

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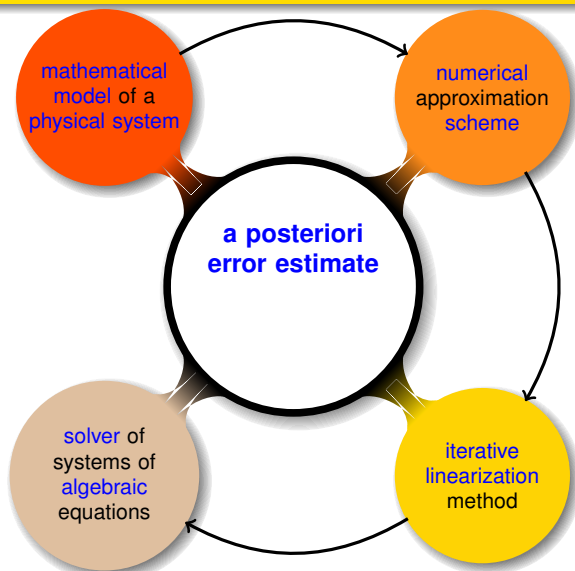


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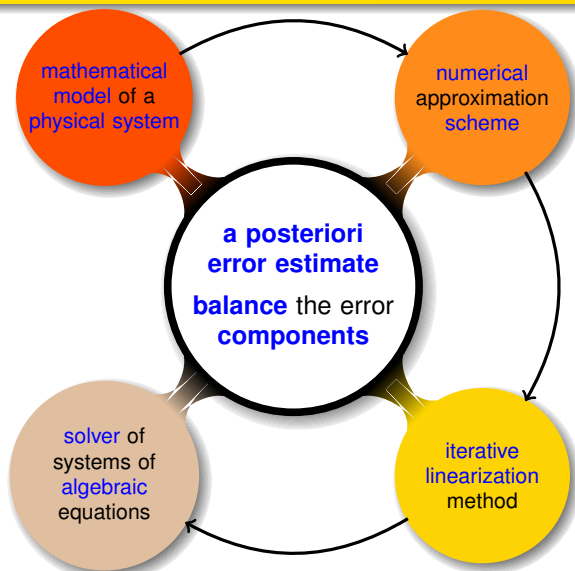
numerical simulation

control the error
(a posteriori estimates)

hard since

- **target unknown**
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Control the error and act adaptively: numerical simulations



numerical simulation

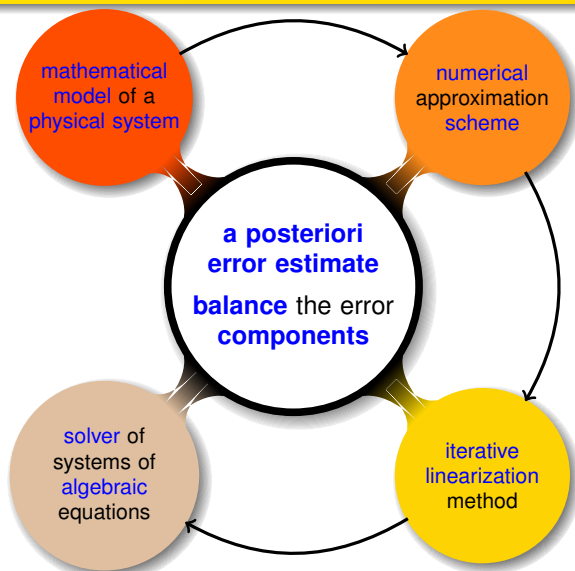
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A posteriori error estimates: discretization error control

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$$-\Delta u = f \quad \text{in } \Omega,$$

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

Guaranteed error upper bound (reliability) ($u_\ell \in \mathcal{P}_p(\mathcal{T}_\ell) \cap H_0^1(\Omega)$, $p \geq 1$, FEs)

$$\underbrace{\|\nabla(u - u_\ell)\|}_{\text{unknown error}} \quad \underbrace{\eta(u_\ell)}_{\text{computable estimator}}$$

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- Prager and Synge (1947), Ladevèze (1975), Babuška & Rheinboldt (1987), Verfürth (1989), Ainsworth & Oden (1993), Destuynder & Métivet (1999), Vejchodský (2006), Braess, Pillwein, & Schöberl (2009), Ern & Vohralík (2015)

Error characterization

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Let $u \in H_0^1(\Omega)$ be the weak solution and let $u_\ell \in H^1(\mathcal{T}_\ell)$ be *arbitrary*. Then

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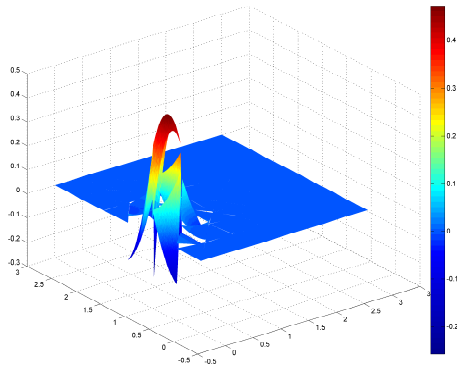
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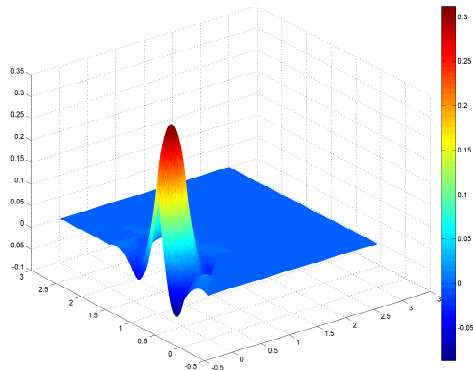
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Potential reconstruction



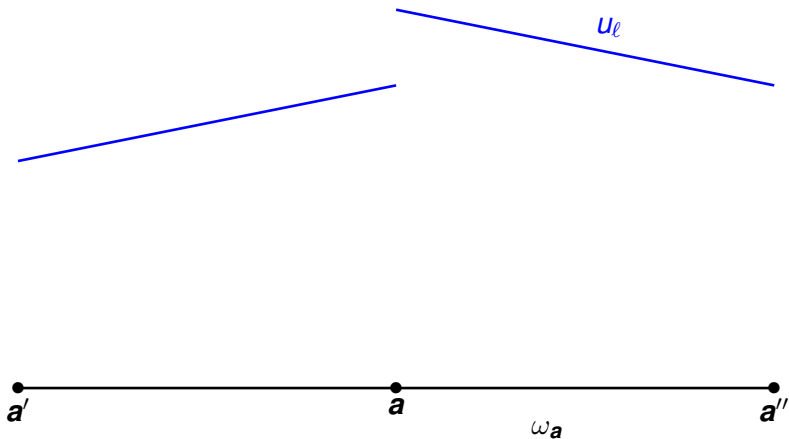
Potential u_ℓ



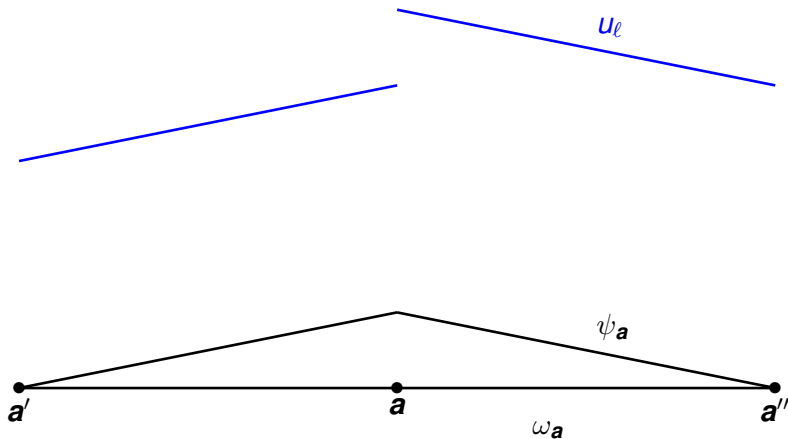
Potential reconstruction s_ℓ

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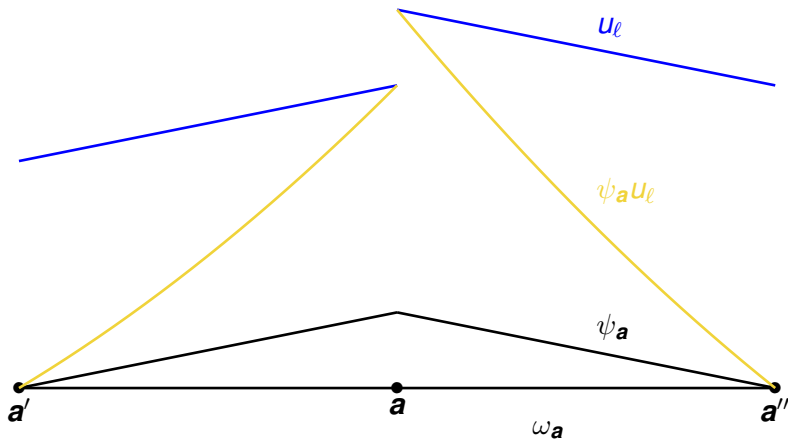
Potential reconstruction in 1D, $p = 1$



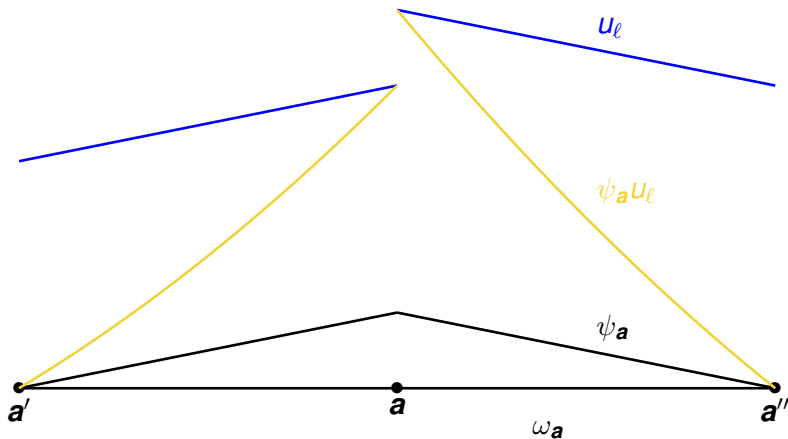
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Potential reconstruction: datum $u_\ell \in \mathcal{P}_p(\mathcal{T}_\ell)$, $p \geq 1$

Definition (**Nonconformity residual lifting** S_ℓ Ern & V. (2015), $\approx$ Carstensen and Mardon (2013))

For each vertex $\mathbf{a} \in \mathcal{V}_\ell$, solve the **local minimization problem**

$$s_\ell^{\mathbf{a}} := \arg \min_{v_\ell \in V_\ell^{\mathbf{a}} = \mathcal{P}_{p+1}(\mathcal{T}^{\mathbf{a}}) \cap H_0^1(\omega_{\mathbf{a}})} \|\nabla(\psi_{\mathbf{a}} u_\ell - v_\ell)\|_{\omega_{\mathbf{a}}}$$

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Equivalent form: **conforming FEs**

Find $s_\ell^{\mathbf{a}} \in V_\ell^{\mathbf{a}}$ such that

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- localization to patches $\mathcal{T}^{\mathbf{a}}$
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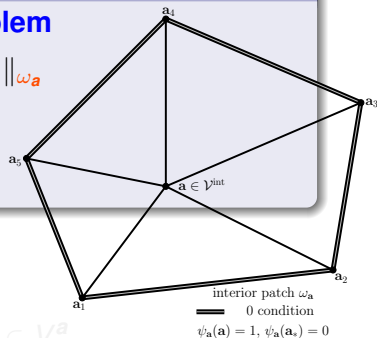
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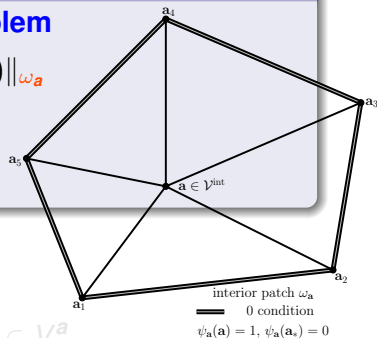
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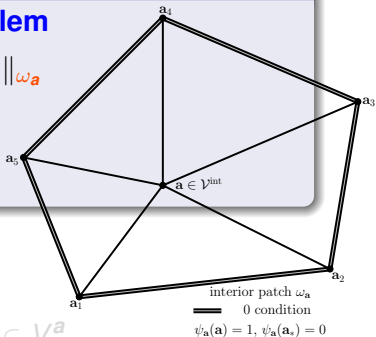
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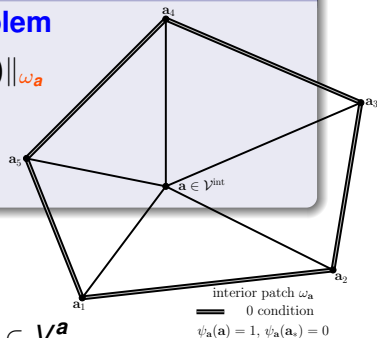
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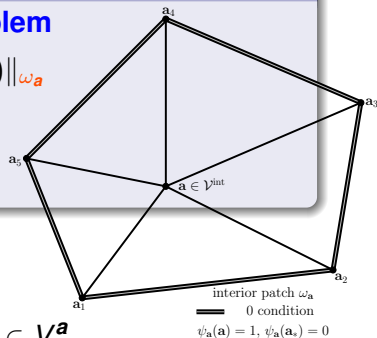
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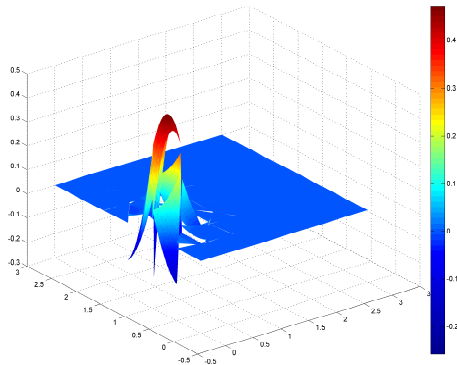
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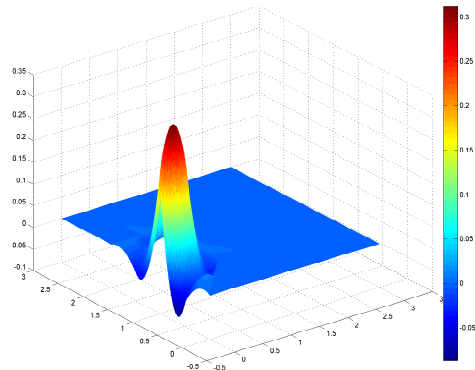
Key points

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- **projection** of the discontinuous $\psi_{\mathbf{a}} u_\ell$ to a conforming space
- homogeneous **Dirichlet** BC on $\partial\omega_{\mathbf{a}}$: $s_\ell \in \mathcal{P}_{p+1}(\mathcal{T}_\ell) \cap H_0^1(\Omega)$

Potential reconstruction



Potential u_ℓ



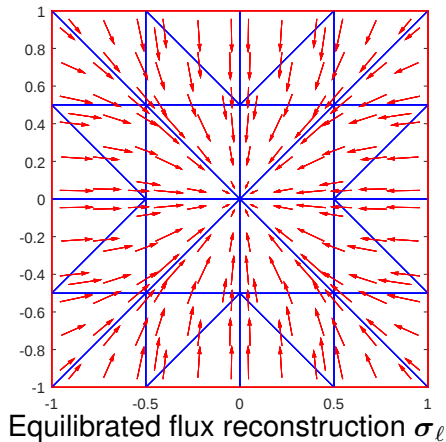
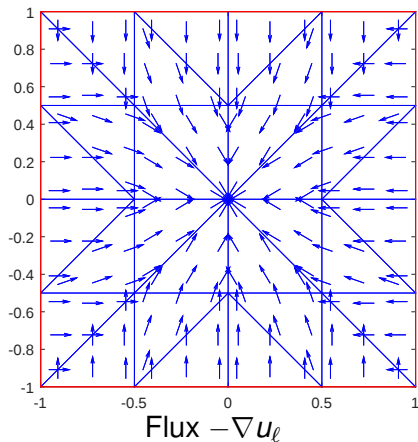
Potential reconstruction s_ℓ

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Outline

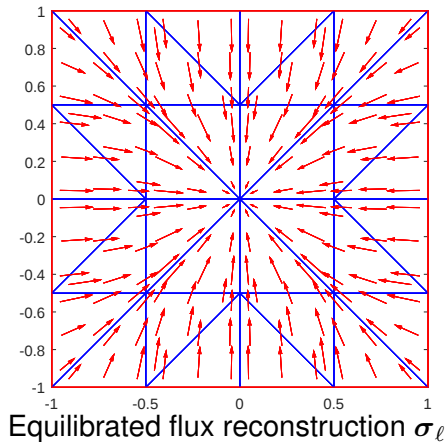
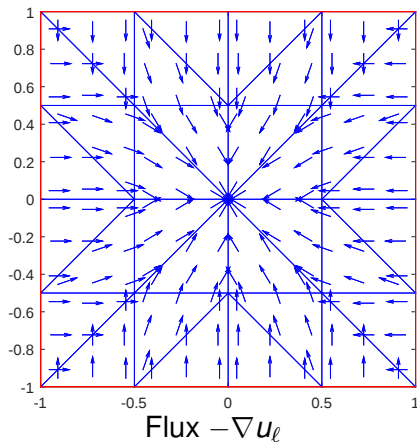
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Equilibrated flux reconstruction



$$\underbrace{-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell), f \in L^2(\Omega)}_{(f, \psi_a)_{\omega_a} - (\nabla u_\ell, \nabla \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}} \rightarrow \sigma_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}, \Omega), \nabla \cdot \sigma_\ell = \Pi_p f$$

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Equilibrated flux reconstruction: $-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell)$, $p \geq 1$, $f \in L^2(\Omega)$

Assumption (Orthogonality wrt hat functions)

There holds $(f, \psi_a)_{\omega_a} - (\nabla u_\ell, \nabla \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}.$

Definition (Dual PDE residual lifting σ_ℓ , Destuynder and Métivet (1999) & Braess and Schöberl (2008))

For each $a \in \mathcal{V}_\ell$, solve the **local constrained minimization pb**

$$\sigma_\ell^a := \arg \min_{\substack{\mathbf{v}_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell) \\ \nabla \cdot \mathbf{v}_\ell = f \psi_a}} \|\psi_a \nabla u_\ell + \mathbf{v}_\ell\|_{\omega_a}$$

and combine $\sigma_\ell = \sum_{a \in \mathcal{V}_\ell} \sigma_\ell^a$

Key points

- homogeneous Neumann BC on $\partial\omega_a$: $\sigma_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell) \cap H(\text{div}, \Omega)$
- equilibrium** $\nabla \cdot \sigma_\ell = \sum_{a \in \mathcal{V}_\ell} \nabla \cdot \sigma_\ell^a = \sum_{a \in \mathcal{V}_\ell} \Pi_p(f \psi_a - \nabla u_\ell \cdot \nabla \psi_a) = \Pi_p f$

Equilibrated flux reconstruction: $-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell)$, $p \geq 1$, $f \in L^2(\Omega)$

Assumption (Orthogonality wrt hat functions)

There holds $(f, \psi_a)_{\omega_a} - (\nabla u_\ell, \nabla \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}.$

Definition (Dual PDE residual lifting σ_ℓ , Destuynder and Métivet (1999) & Braess and Schöberl (2008))

For each $a \in \mathcal{V}_\ell$, solve the **local constrained minimization pb**

$$\sigma_\ell^a := \arg \min_{\substack{\mathbf{v}_\ell \in \mathcal{RT}_p(\mathcal{T}^a) \cap H_0(\text{div}, \omega_a) \\ \nabla \cdot \mathbf{v}_\ell = \Pi_p(f\psi_a - \nabla u_\ell \nabla \psi_a)}} \|\psi_a \nabla u_\ell + \mathbf{v}_\ell\|_{\omega_a}$$

and combine $\sigma_\ell = \sum_{a \in \mathcal{V}_\ell} \sigma_\ell^a$

Key points

- homogeneous Neumann BC on $\partial\omega_a$: $\sigma_\ell \in \mathcal{RT}_p(\mathcal{T}_\ell) \cap H(\text{div}, \Omega)$
- equilibrium** $\nabla \cdot \sigma_\ell = \sum_{a \in \mathcal{V}_\ell} \nabla \cdot \sigma_\ell^a = \sum_{a \in \mathcal{V}_\ell} \Pi_p(f\psi_a - \nabla u_\ell \nabla \psi_a) = \Pi_p f$

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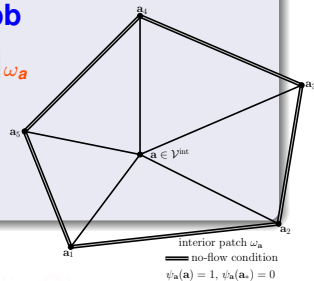
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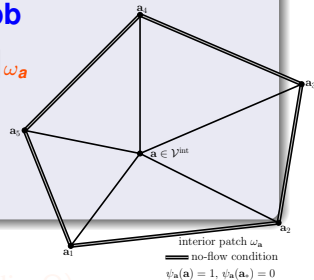
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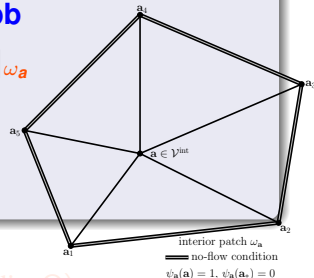
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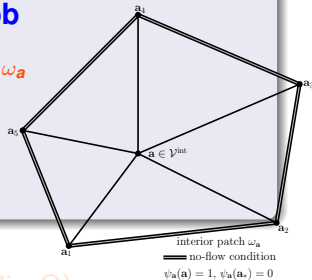
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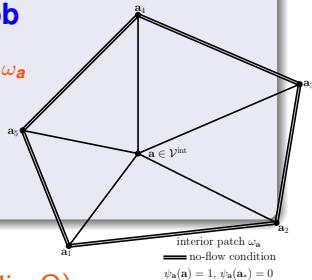
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and combine

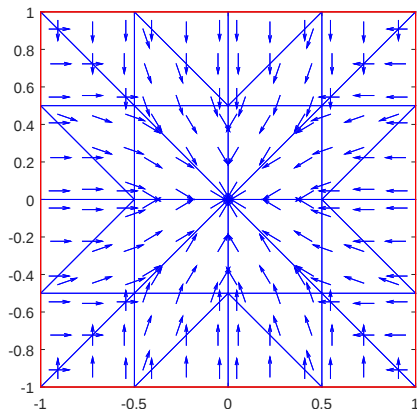
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Key points

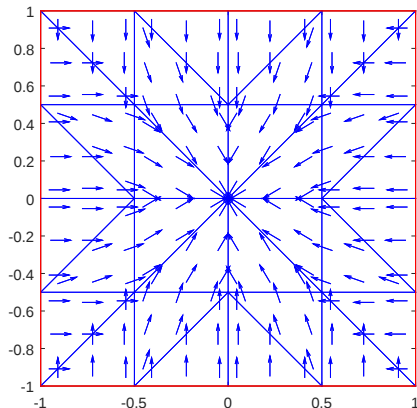
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Flux $-\nabla u_\ell \notin \mathbf{H}(\text{div}, \Omega)$, $\nabla \cdot (-\nabla u_\ell) \neq f$

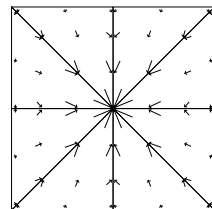
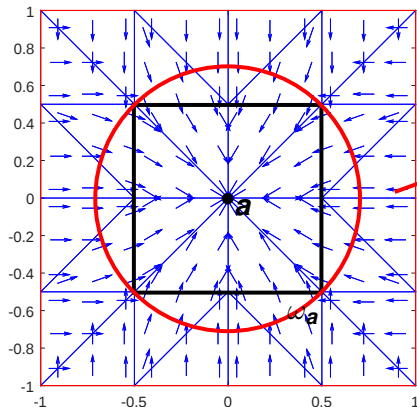
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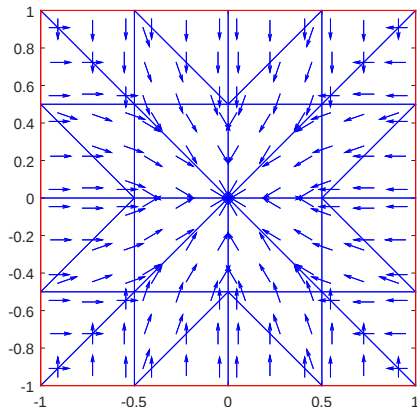
$$-\psi_a \nabla u_\ell$$

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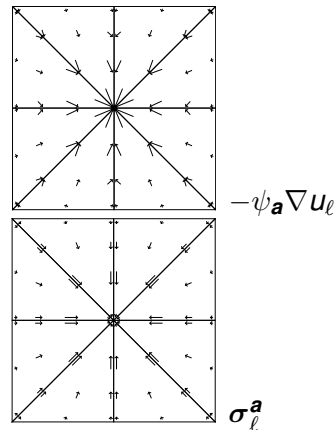
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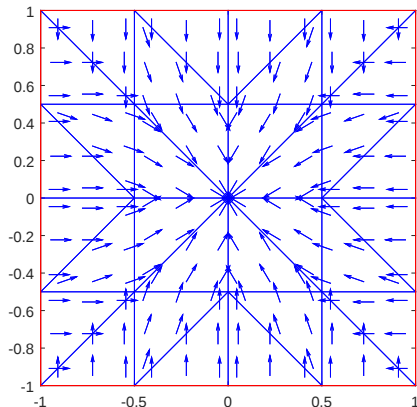


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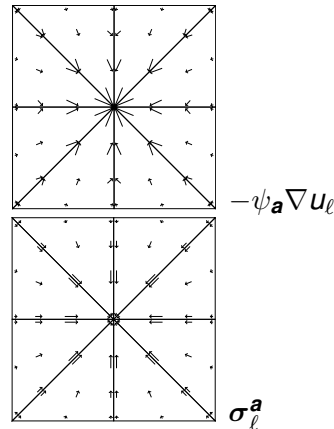


$$\underbrace{-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell), f \in L^2(\Omega)}_{(f, \psi_a)_{\omega_a} - (\nabla u_\ell, \nabla \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}}$$

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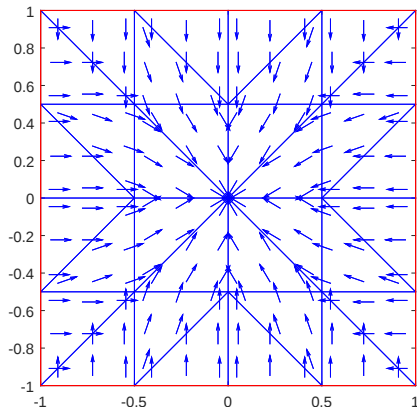
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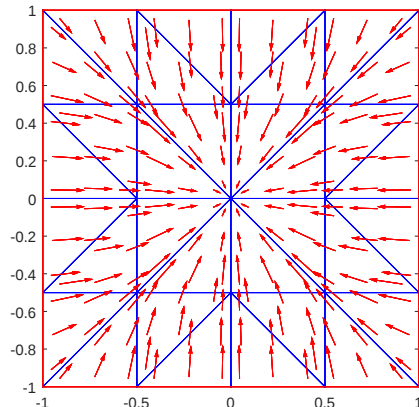
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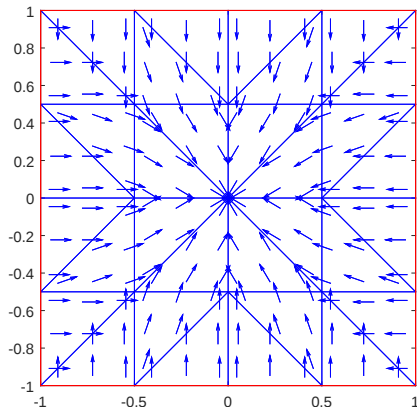
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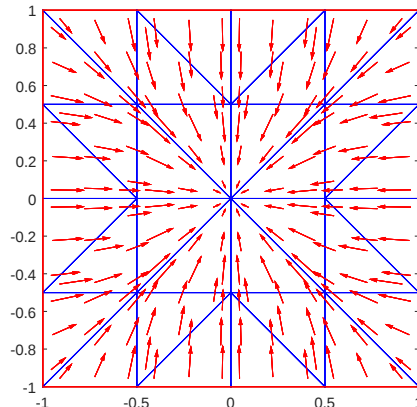
Equilibrated flux rec. σ_ℓ

$$\underbrace{-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell), f \in L^2(\Omega)}_{(f, \psi_a)_{\omega_a} - (\nabla u_\ell, \nabla \psi_a)_{\omega_a} = 0 \quad \forall a \in \mathcal{V}_\ell^{\text{int}}} \rightarrow \sigma_\ell := \sum_{a \in \mathcal{V}_\ell} \sigma_\ell^a \in \mathcal{RT}_p(\mathcal{T}_\ell) \cap \mathbf{H}(\text{div}, \Omega), \nabla \cdot \sigma_\ell = \Pi_p f$$

Equilibrated flux reconstruction: $-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell)$, $p \geq 1$, $f \in L^2(\Omega)$



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Equilibrated flux rec. $\sigma_\ell \in \mathbf{H}(\text{div}, \Omega)$, $\nabla \cdot \sigma_\ell = f$

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Outline

- 1 Introduction: adaptive iterative approximation
 - A posteriori error estimates and adaptivity
 - Achievements and example results
 - Real life comparison
- 2 Linear diffusion: discretization error, mesh and polynomial degree adaptivity
 - A posteriori error estimates
 - Potential reconstruction
 - Flux reconstruction
 - **A posteriori error control**
 - Balancing error components: mesh adaptivity
 - Balancing error components: polynomial-degree adaptivity
- 3 Nonlinear diffusion: overall error and solvers adaptivity
 - A posteriori error estimates (overall and components)
 - Balancing error components: solvers adaptivity
- 4 Conclusions

How **large** is the discretization error? (model pb, known smooth solution)

ℓ	p	$\eta(u_\ell)$	rel. error estimate $\frac{\eta(u_\ell)}{\ \nabla u_\ell\ }$	$\ \nabla(u - u_\ell)\ $	rel. error $\frac{\ \nabla(u - u_\ell)\ }{\ \nabla u_\ell\ }$	$\ell^p = \frac{\eta(u_\ell)}{\ \nabla(u - u_\ell)\ }$
0	1	1.25	28%	1.07	24%	1.17
1	2	6.07×10^{-2}				
2	3	3.10×10^{-3}				
3	4	1.45×10^{-4}				
1	2					
2	3					
3	4					

A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

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

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0	1	1.25	28%	1.07	24%	1.17
1	1	6.07×10^{-1}	25%	5.58×10^{-1}	24%	1.17
2	1	3.10×10^{-1}	25%	2.99×10^{-1}	24%	1.17
3	1	1.45×10^{-1}	25%	1.39×10^{-1}	24%	1.17
1	2	4.23×10^{-2}	25%	4.00×10^{-2}	24%	1.17
2	3	2.62×10^{-4}	25%	2.50×10^{-4}	24%	1.17
3	4	2.60×10^{-7}	25%	2.50×10^{-7}	24%	1.17

A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

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

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ℓ	p	$\eta(u_\ell)$	rel. error estimate $\frac{\eta(u_\ell)}{\ \nabla u_\ell\ }$	$\ \nabla(u - u_\ell)\ $	rel. error $\frac{\ \nabla(u - u_\ell)\ }{\ \nabla u_\ell\ }$	$\rho^{\text{eff}} = \frac{\eta(u_\ell)}{\ \nabla(u - u_\ell)\ }$
0	1	1.25	28%	1.07	24%	1.17
1	2	6.07×10^{-1}	14%	5.58×10^{-1}	13%	1.17
2	3	3.10×10^{-1}	7.0%	2.99×10^{-1}	6.5%	1.17
3	4	1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	1.17
1	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.00×10^{-2}	$8.8 \times 10^{-1}\%$	1.17
2	3	2.62×10^{-3}	$5.9 \times 10^{-2}\%$	2.50×10^{-3}	$5.5 \times 10^{-2}\%$	1.17
3	4	2.60×10^{-4}	$5.9 \times 10^{-3}\%$	2.50×10^{-4}	$5.5 \times 10^{-3}\%$	1.17

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0	1	1.25	28%	1.07	24%	1.17
1		6.07×10^{-1}	14%	5.56×10^{-1}	13%	1.22
2		3.10×10^{-1}	7.0%	2.92×10^{-1}	6.5%	
3		1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	
1	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.07×10^{-2}		
2	3	2.62×10^{-3}	$5.9 \times 10^{-3}\%$	2.60×10^{-3}		
3	4	2.60×10^{-7}	$5.9 \times 10^{-6}\%$	2.58×10^{-7}		

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0	1	1.25	28%	1.07	24%	1.17
1		6.07×10^{-1}	14%	5.56×10^{-1}	13%	1.09
2		3.10×10^{-1}	7.0%	2.92×10^{-1}	6.6%	1.05
3		1.45×10^{-1}	3.3%	1.39×10^{-1}	3.1%	1.03
1	2	4.23×10^{-2}	$9.5 \times 10^{-1}\%$	4.07×10^{-2}	$9.2 \times 10^{-1}\%$	
2	3	2.62×10^{-3}	$5.9 \times 10^{-3}\%$	2.60×10^{-3}	$5.9 \times 10^{-3}\%$	
3	4	2.60×10^{-7}	$5.9 \times 10^{-6}\%$	2.58×10^{-7}	$5.8 \times 10^{-6}\%$	

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

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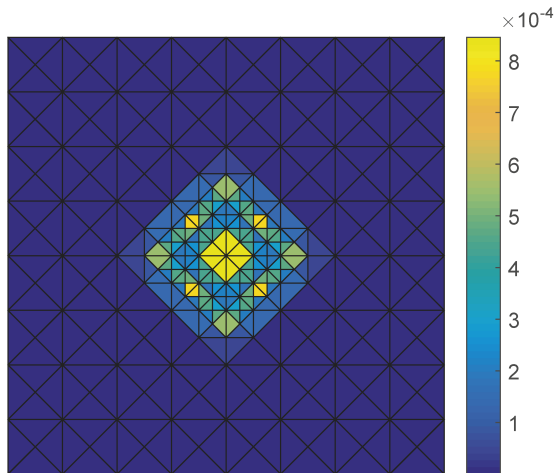
A. Ern, M. Vohralík, SIAM Journal on Numerical Analysis (2015)

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

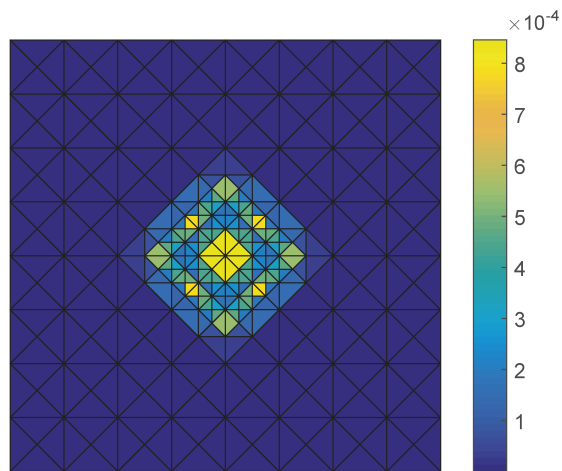
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Where (in space) is the error **localized**? (known smooth solution)

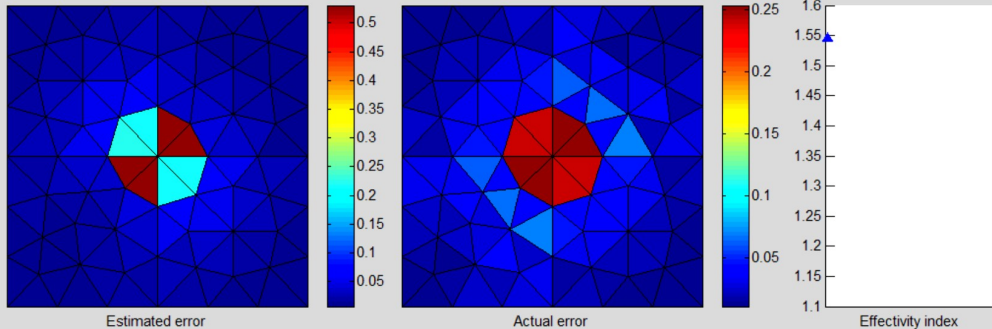


Estimated error distribution $\eta_K(u_\ell)$



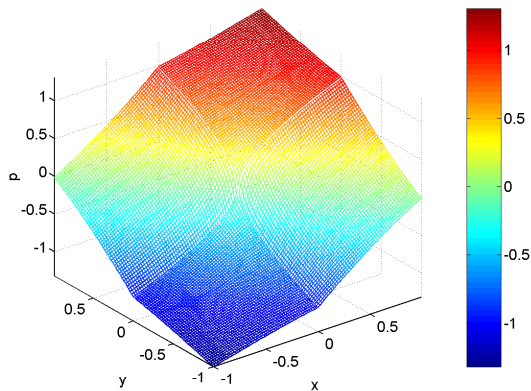
Exact error distribution $\|\nabla(u - u_\ell)\|_K$

Can we decrease the error efficiently? (adaptive mesh refinement)

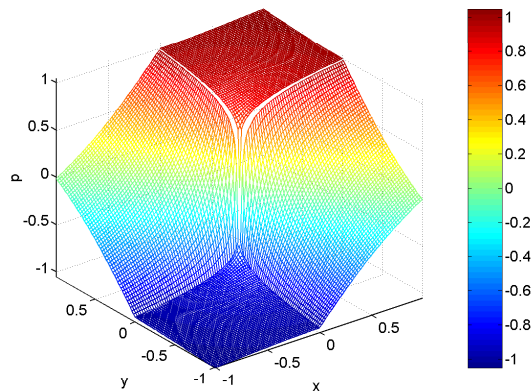


M. Vohralík, SIAM Journal on Numerical Analysis (2007)

Singular solutions

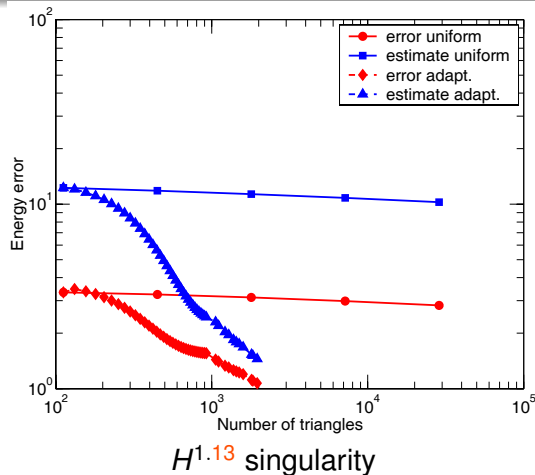
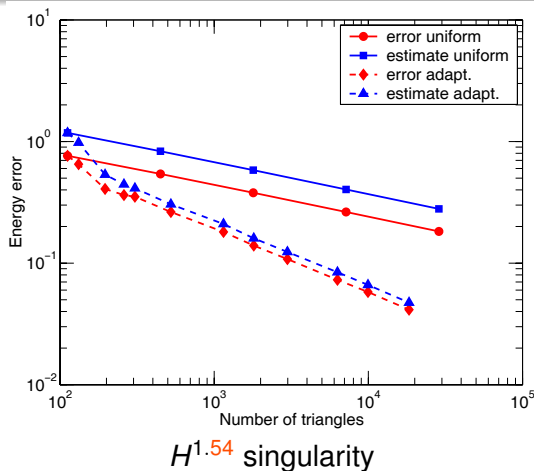


$H^{1.54}$ singularity



$H^{1.13}$ singularity

Estimated and actual error against the number of elements in uniformly/adaptively refined meshes (singular solutions)



Adaptive mesh refinement

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$$\sum_{K \in \mathcal{T}_\ell} \eta_K(u_\ell)^2 = \eta(u_\ell)^2$$

Adaptive mesh refinement

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- Dörfler marking: subset $\mathcal{M}_\ell$ containing θ -fraction of the estimates

$$\sum_{K \in \mathcal{M}_\ell} \eta_K(u_\ell)^2 \geq \theta^2 \sum_{K \in \mathcal{T}_\ell} \eta_K(u_\ell)^2 = \theta^2 \eta(u_\ell)^2$$

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Convergence on a sequence of adaptively refined meshes

- $\|\nabla(u - u_\ell)\| \rightarrow 0$
- some mesh elements may not be refined at all: $h \not\rightarrow 0$
- Babuška & Miller (1987), Dörfler (1996)

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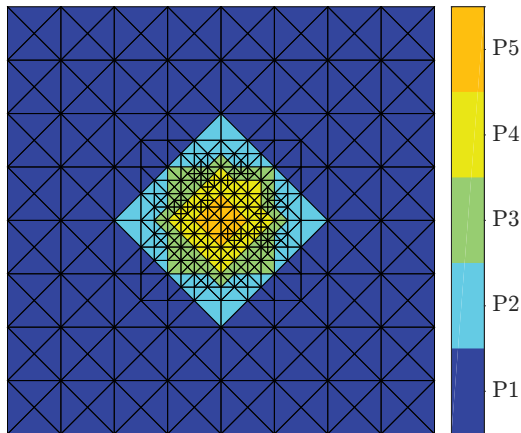
Optimal error decay rate wrt degrees of freedom

- $\|\nabla(u - u_\ell)\| \lesssim |\text{DoF}_\ell|^{-p/d}$ (replaces h^p)
- same for smooth & singular solutions: ~~higher-order only pay-off for sm. sol.~~
- decays to zero as fast as on a best-possible sequence of meshes
- Morin, Nochetto, Siebert (2000), Stevenson (2005, 2007), Cascón, Kreuzer, Nochetto, Siebert (2008), Canuto, Nochetto, Stevenson, Verani (2017)

Outline

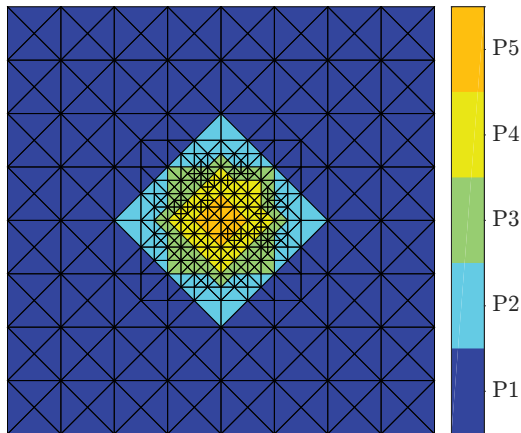
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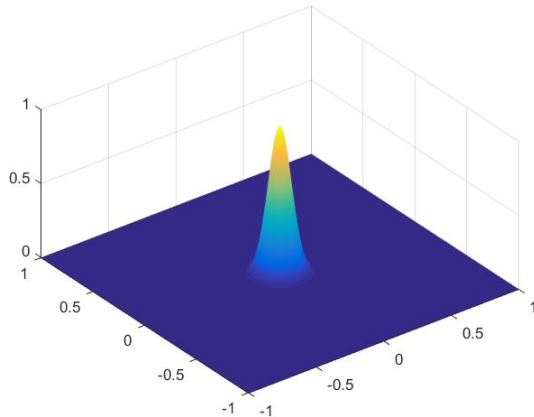


Mesh $\mathcal{T}_\ell$ and pol. degrees p_K

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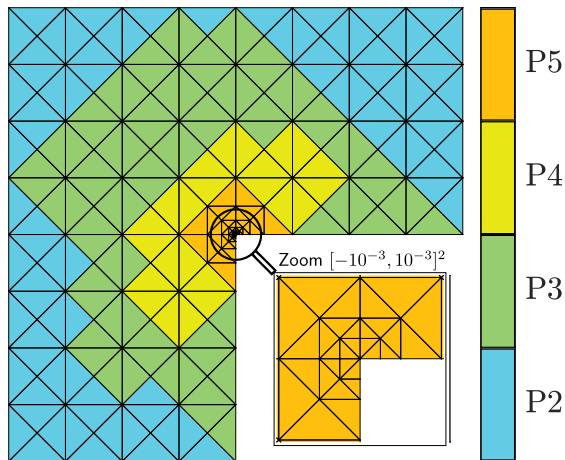
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Exact solution

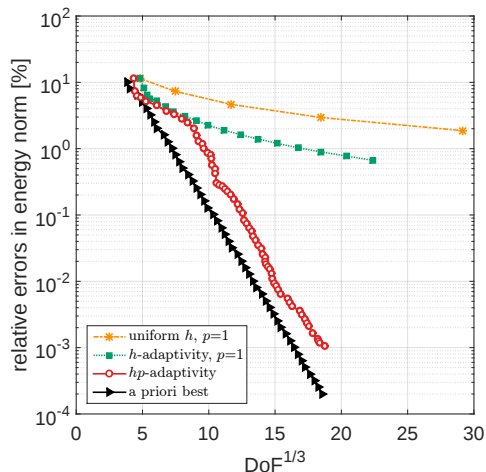
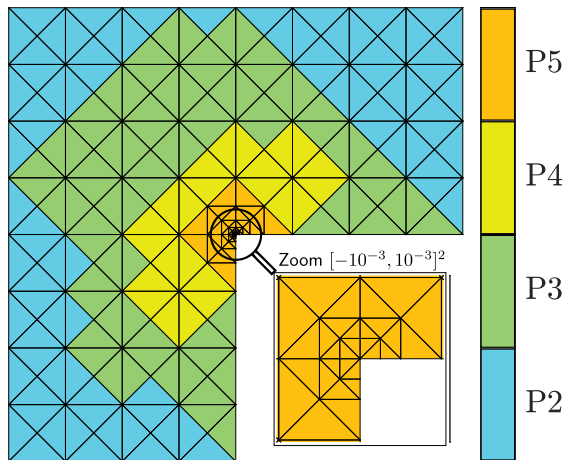
P. Daniel, A. Ern, I. Smears, M. Vohralík, Computers & Mathematics with Applications (2018)

Best-possible error decrease: *hp* adaptivity, (singular solution)



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Relative error as a function of DoF

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A model nonlinear problem

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

$$\underbrace{(a(|\nabla u|)\nabla u, \nabla v)}_{\langle \mathcal{A}(u), v \rangle: \text{nonlinear operator}} = (f, v) \quad \forall v \in H_0^1(\Omega).$$

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Assumption (Gradient-dependent diffusivity)

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

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- $a_m \leq a(r) \leq a_c, \quad a_m \leq (a(r)r)' \leq a_c$

Example of the nonlinear function a

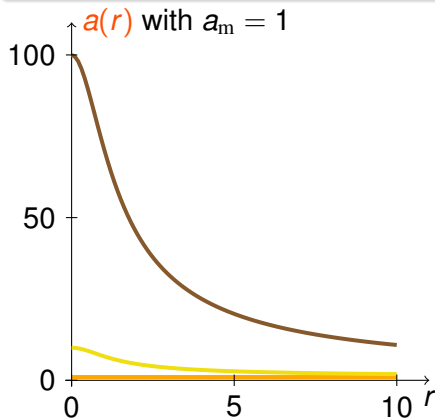
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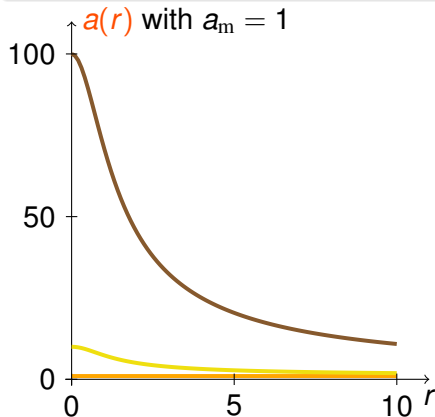


$$\begin{aligned} a_c &= 100 \\ a_c &= 10 \\ a_c &= 1 \end{aligned}$$

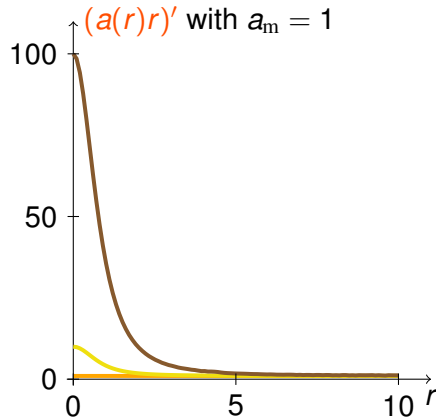
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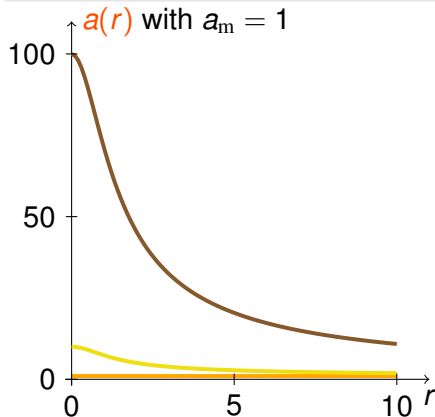
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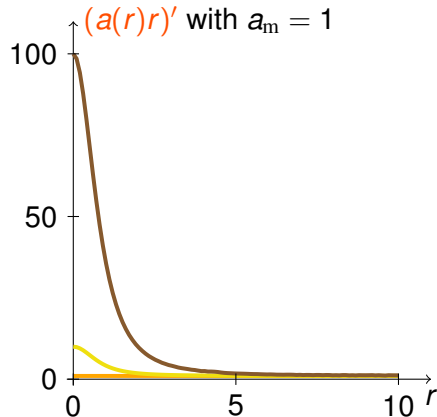
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Strength of the nonlinearity

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



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- system of **nonlinear algebraic equations**

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- $\mathbf{A}_\ell^{k-1}, \mathbf{b}_\ell^{k-1}$: matrix- and vector-valued functions constructed from u_ℓ^{k-1}

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- iterative linearization index $k \geq 1$
- $u_\ell^0 \in V_\ell$ a given initial guess
- $\mathbf{A}_\ell^{k-1}, \mathbf{b}_\ell^{k-1}$: matrix- and vector-valued functions constructed from u_ℓ^{k-1}
- system of **linear algebraic equations**

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Iterative algebraic solver

Definition (Iterative algebraic solver)

Find $u_\ell^{k,i} \in V_\ell$ such that

$$(\mathbf{A}_\ell^{k-1} \nabla u_\ell^{k,i} - \mathbf{b}_\ell^{k-1}, \nabla v_\ell) = (f, v_\ell) - \underbrace{(r_\ell^{k,i}, v_\ell)}_{\langle \mathcal{R}^{\text{alg}}(u_\ell^{k,i}), v_\ell \rangle: \text{algebraic residual}} \quad \forall v_\ell \in V_\ell.$$

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Outline

- 1 Introduction: adaptive iterative approximation
 - A posteriori error estimates and adaptivity
 - Achievements and example results
 - Real life comparison
- 2 Linear diffusion: discretization error, mesh and polynomial degree adaptivity
 - A posteriori error estimates
 - Potential reconstruction
 - Flux reconstruction
 - A posteriori error control
 - Balancing error components: mesh adaptivity
 - Balancing error components: polynomial-degree adaptivity
- 3 Nonlinear diffusion: overall error and solvers adaptivity
 - A posteriori error estimates (overall and components)
 - Balancing error components: solvers adaptivity
- 4 Conclusions

Error measure, error components

Definition (Intrinsic measure of error)

$$|||u - u_{\ell}^{k,i}||| := \max_{\substack{v \in H_0^1(\Omega) \\ \|\nabla v\|=1}} \left\{ \underbrace{(f, v) - (a(|\nabla u_{\ell}^{k,i}|) \nabla u_{\ell}^{k,i}, \nabla v)} \right\}$$

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Error components

$$\begin{aligned} \underbrace{(f, v) - \langle \mathcal{A}(u_\ell^{k,i}), v \rangle}_{\langle \mathcal{R}(u_\ell^{k,i}), v \rangle: \text{total residual}} &= \underbrace{(f, v) - \langle \mathcal{A}^{k-1}(u_\ell^{k,i}), v \rangle - \langle \mathcal{R}^{\text{alg}}(u_\ell^{k,i}), v \rangle}_{\langle \mathcal{R}^{\text{disc}}(u_\ell^{k,i}), v \rangle: \text{discretization residual}} \\ &+ \underbrace{\langle \mathcal{A}^{k-1}(u_\ell^{k,i}), v \rangle - \langle \mathcal{A}(u_\ell^{k,i}), v \rangle}_{\langle \mathcal{R}^{\text{lin}}(u_\ell^{k,i}), v \rangle: \text{linearization residual}} \\ &+ \underbrace{\langle \mathcal{R}^{\text{alg}}(u_\ell^{k,i}), v \rangle}_{\text{algebraic residual}} \end{aligned}$$

Reminder from the linear case

Theorem (Error characterization)

Let $u \in H_0^1(\Omega)$ be the weak solution and let $u_\ell \in H_0^1(\Omega)$ be *arbitrary*. Then

$$\|\nabla(u - u_\ell)\|^2 = \underbrace{\min_{\substack{\mathbf{v} \in \mathbf{H}(\text{div}, \Omega) \\ \nabla \cdot \mathbf{v} = f}} \|\nabla u_\ell + \mathbf{v}\|^2}_{\substack{\text{constrained distance to } \mathbf{H}(\text{div}, \Omega) \\ = \max_{\substack{\mathbf{v} \in H_0^1(\Omega) \\ \|\nabla \mathbf{v}\|=1}} \{(f, \mathbf{v}) - (\nabla u_\ell, \nabla \mathbf{v})\}^2}} \quad .$$

dual norm of the PDE residual

Comments

- It is enough to choose suitable (discrete, piecewise polynomial) $\sigma_\ell \in \mathbf{H}(\text{div}, \Omega)$ with $\nabla \cdot \sigma_\ell = f$ to get the guaranteed upper bound $\|\nabla(u - u_\ell)\| \leq \|\nabla u_\ell + \sigma_\ell\|$.

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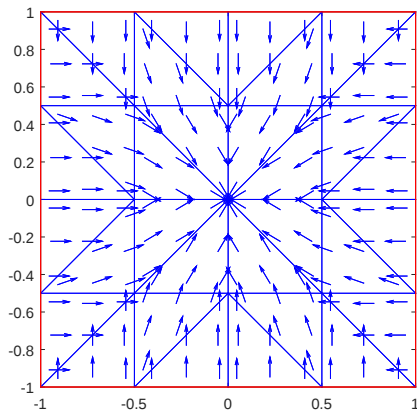
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dual norm of the PDE residual

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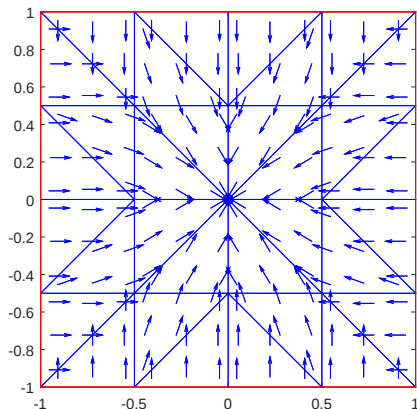
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Equilibrated flux reconstruction: $-\nabla u_\ell \in \mathcal{RT}_{p-1}(\mathcal{T}_\ell)$, $p \geq 1$, $f \in L^2(\Omega)$



Flux $-\nabla u_\ell \notin \mathbf{H}(\text{div}, \Omega)$, $\nabla \cdot (-\nabla u_\ell) \neq f$

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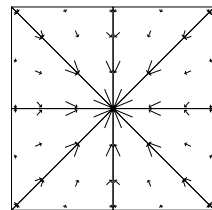
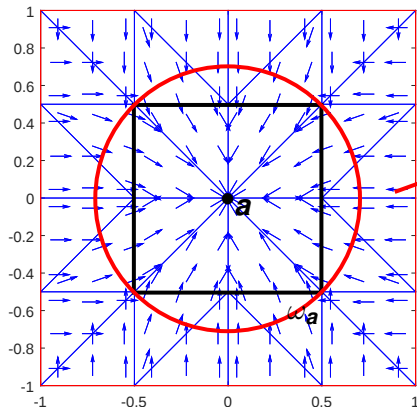


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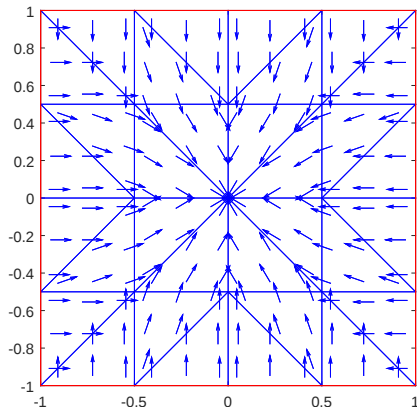
$$-\psi_a \nabla u_\ell$$

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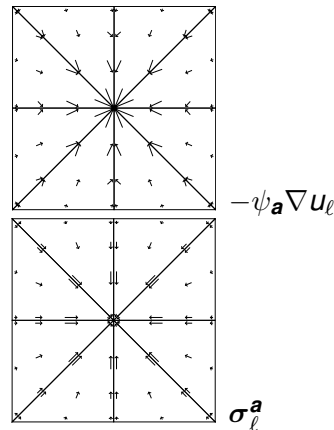
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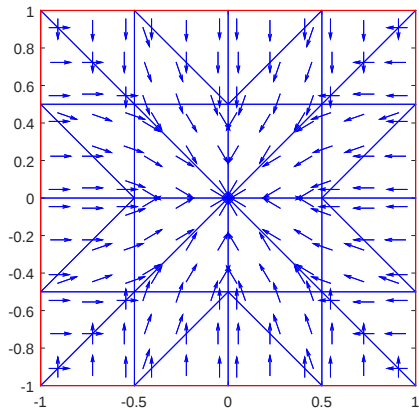


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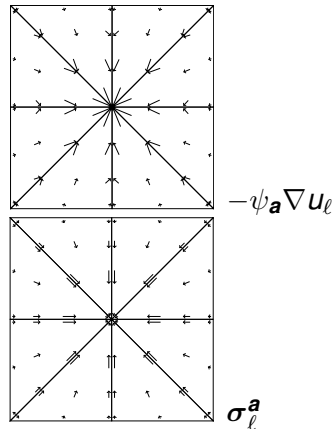


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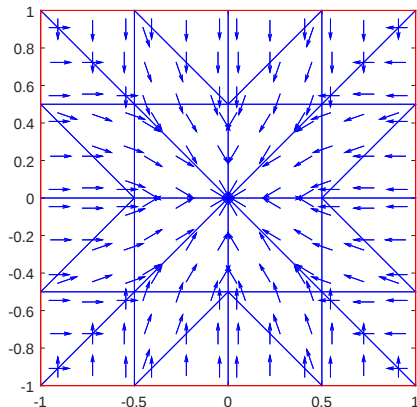
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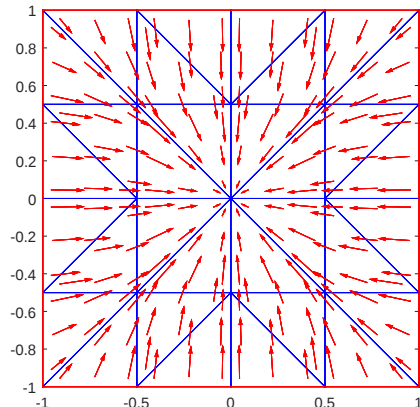
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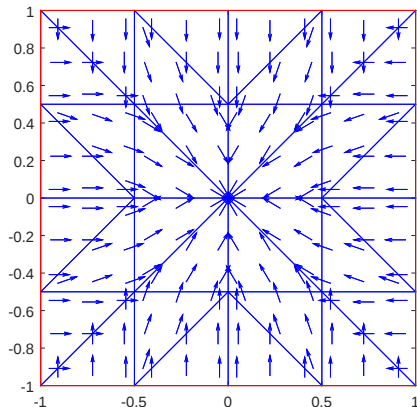
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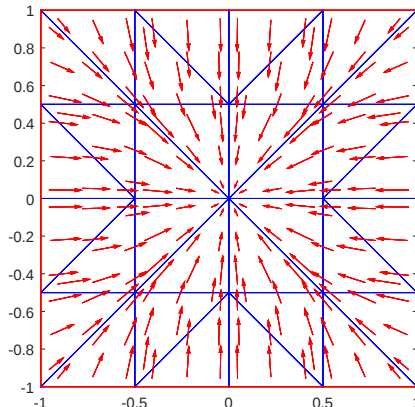
Equilibrated flux rec. σ_ℓ

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Flux reconstructions (residual liftings)

Definition (Discretization residual lifting $\sigma_{\text{disc},\ell}^{k,i}$)

For each $\mathbf{a} \in \mathcal{V}_\ell$, solve the local constrained minimization pb

$$\sigma_{\text{disc},\ell}^{k,i,\mathbf{a}} := \arg \min_{\substack{\mathbf{v}_\ell \in \mathcal{RT}_p(\mathcal{T}^{\mathbf{a}}) \cap \mathbf{H}_0(\text{div}, \omega_{\mathbf{a}}) \\ \nabla \cdot \mathbf{v}_\ell = \Pi_p((f - r_\ell^{k,i})\psi_{\mathbf{a}} - (\mathbf{A}_\ell^{k-1} \nabla u_\ell^{k,i} - \mathbf{b}_\ell^{k-1}) \cdot \nabla \psi_{\mathbf{a}})}} \|\psi_{\mathbf{a}}(\mathbf{A}_\ell^{k-1} \nabla u_\ell^{k,i} - \mathbf{b}_\ell^{k-1}) + \mathbf{v}_\ell\|_{\omega_{\mathbf{a}}}$$

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Definition (Linearization residual lifting $\sigma_{\text{lin},\ell}^{k,i}$)

For each $\mathbf{a} \in \mathcal{V}_\ell$, solve the local constrained minimization pb

$$\sigma_{\text{lin},\ell}^{k,i,\mathbf{a}} := \arg \min_{\substack{\mathbf{v}_\ell \in \mathcal{RT}_p(\mathcal{T}^{\mathbf{a}}) \cap \mathbf{H}_0(\text{div}, \omega_{\mathbf{a}}) \\ \nabla \cdot \mathbf{v}_\ell = 0}} \|\psi_{\mathbf{a}}(a(|\nabla u_\ell^{k,i}|) \nabla u_\ell^{k,i} - (\mathbf{A}_\ell^{k-1} \nabla u_\ell^{k,i} - \mathbf{b}_\ell^{k-1})) + \mathbf{v}_\ell\|_{\omega_{\mathbf{a}}}$$

and combine

$$\sigma_{\text{lin},\ell}^{k,i} := \sum_{\mathbf{a} \in \mathcal{V}_\ell} \sigma_{\text{lin},\ell}^{k,i,\mathbf{a}}.$$

Algebraic error flux reconstruction, two-level setting

Definition (Coarse grid solve)

Find $\rho_{\text{alg},0}^{k,i} \in \mathcal{P}_1(\mathcal{T}_0) \cap H_0^1(\Omega)$ s.t.

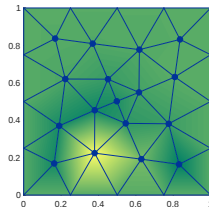
$$(\nabla \rho_{\text{alg},0}^{k,i}, \nabla \psi_{\mathbf{a}})_{\omega_{\mathbf{a}}} = (r_{\ell}^{k,i}, \psi_{\mathbf{a}})_{\omega_{\mathbf{a}}} \quad \forall \mathbf{a} \in \mathcal{V}_0.$$

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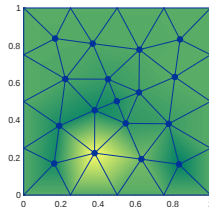
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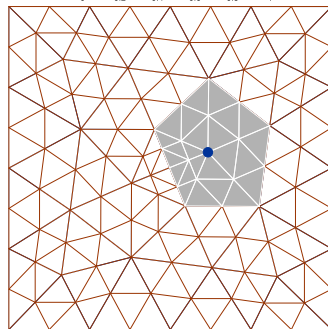
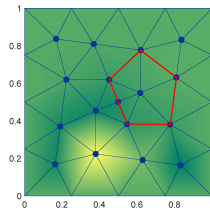
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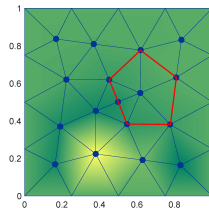
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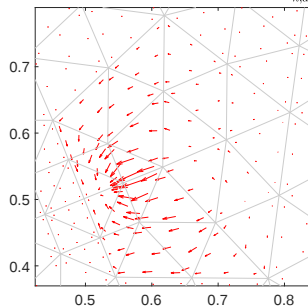
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local alg. error flux reconstruction, $\sigma_{h,\text{alg}}^{a,i}$



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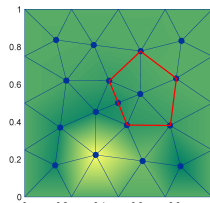
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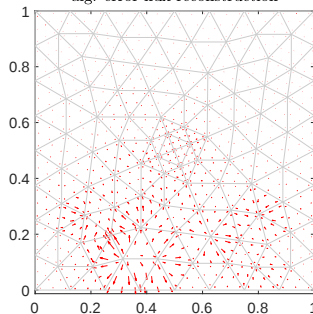
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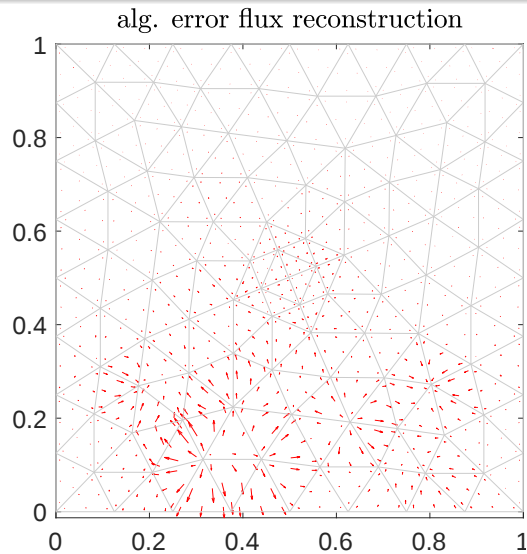
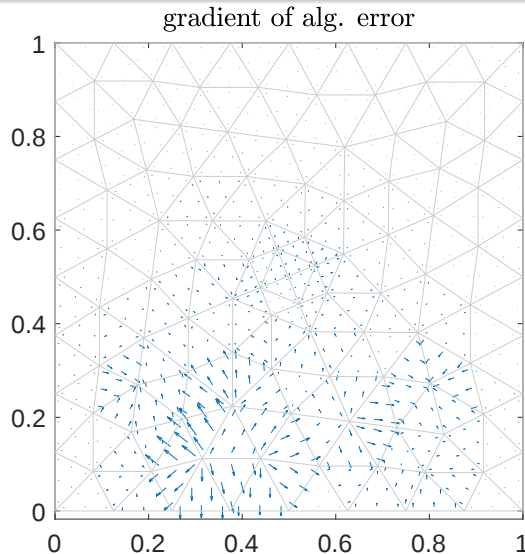
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alg. error flux reconstruction



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A posteriori error estimate distinguishing the error components

Theorem (A posteriori error estimate)

Let $u_{\ell}^{k,i} \in V_{\ell}$ be given.

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$$\begin{aligned} |||u - u_\ell^{k,i}||| &\leq \underbrace{\|a(|\nabla u_\ell^{k,i}|) \nabla u_\ell^{k,i} + \sigma_\ell^{k,i}\|}_{\eta(u_\ell^{k,i})} \\ &\leq \underbrace{\|a(|\nabla u_\ell^{k,i}|) \nabla u_\ell^{k,i} + \sigma_{\text{disc},\ell}^{k,i}\|}_{\eta_{\text{disc}}(u_\ell^{k,i})} + \underbrace{\|\sigma_{\text{lin},\ell}^{k,i}\|}_{\eta_{\text{lin}}(u_\ell^{k,i})} + \underbrace{\|\sigma_{\text{alg},\ell}^{k,i}\|}_{\eta_{\text{alg}}(u_\ell^{k,i})}. \end{aligned}$$

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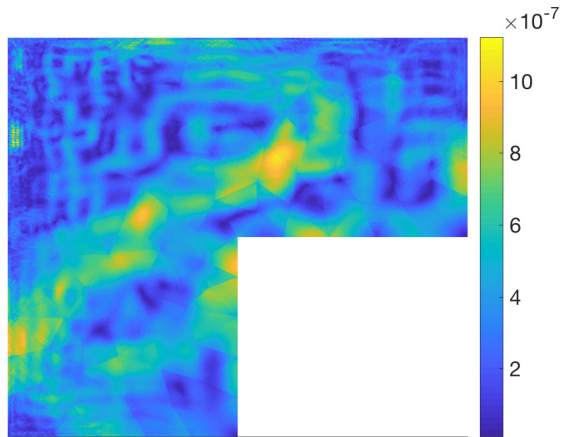
Moreover, the estimates are **locally efficient** and **robust** with respect to the **strength of nonlinearities**

$$\eta_K(u_\ell^{k,i}) \leq C_{\text{eff}} |||u - u_\ell^{k,i}|||_{\omega_K} \quad \forall K \in \mathcal{T}_\ell, \quad C_{\text{eff}} \text{ independent of } \frac{a_c}{a_m}.$$

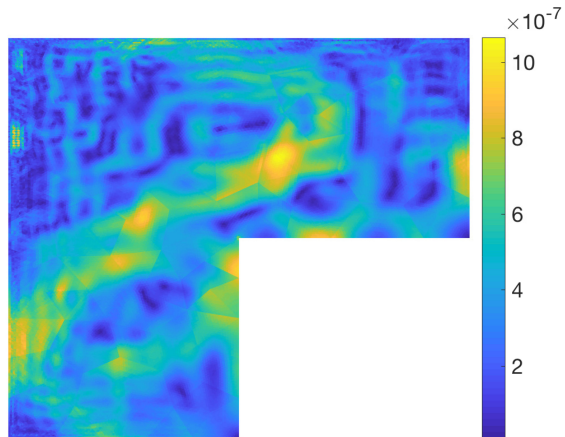
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Including algebraic error: $\mathbb{A}_\ell \mathbf{U}_\ell^i \neq \mathbf{F}_\ell$



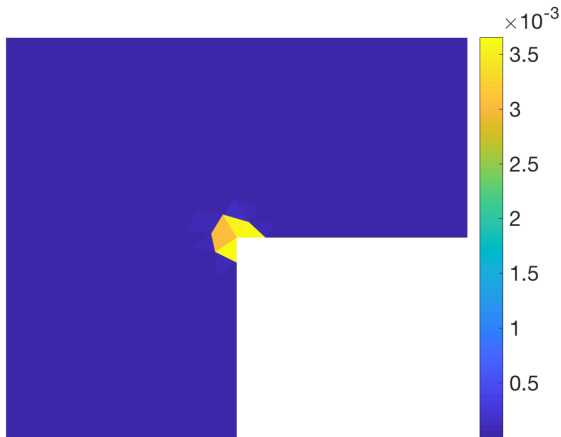
Estimated algebraic errors $\eta_{\text{alg},K}(\mathbf{u}_\ell^i)$



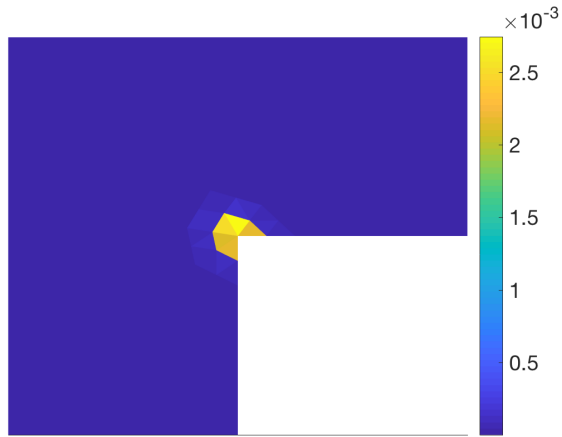
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J. Papež, U. Rüde, M. Vohralík, B. Wohlmuth, Computer Methods in Applied Mechanics and Engineering (2020)

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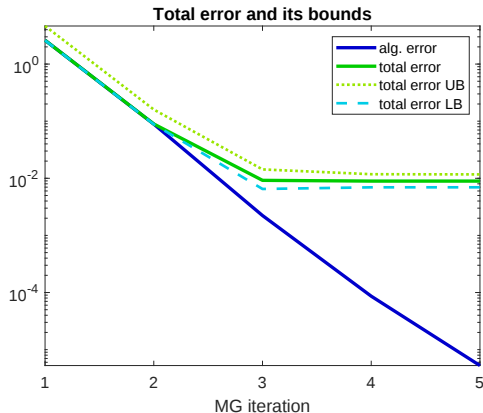
Estimated total errors $\eta_K(u_\ell^i)$



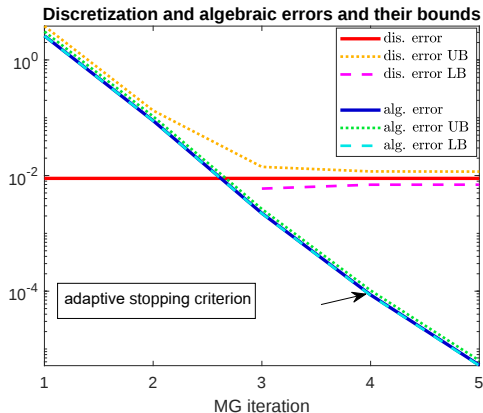
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Total error



Error components and adaptive st. crit.

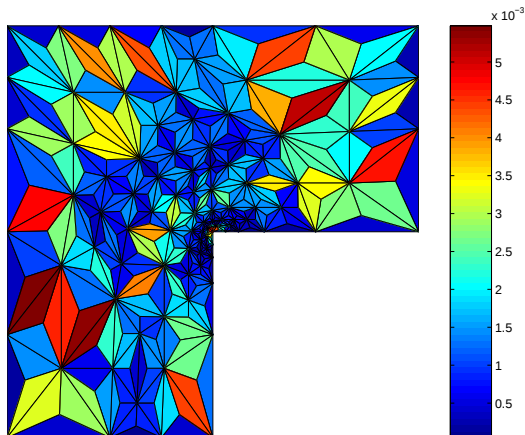
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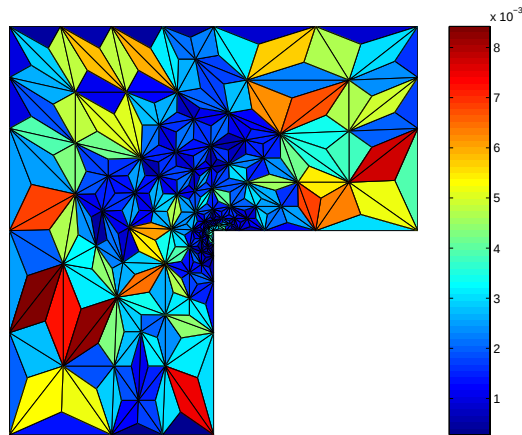
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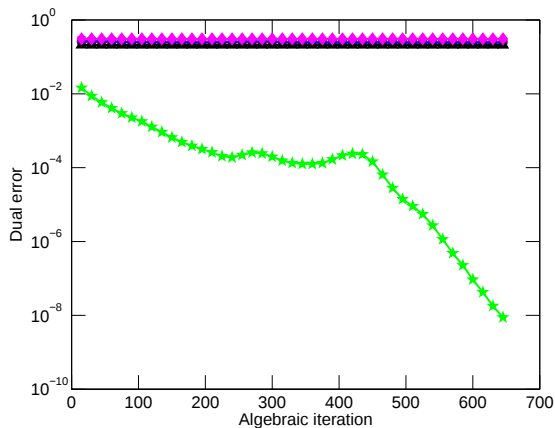
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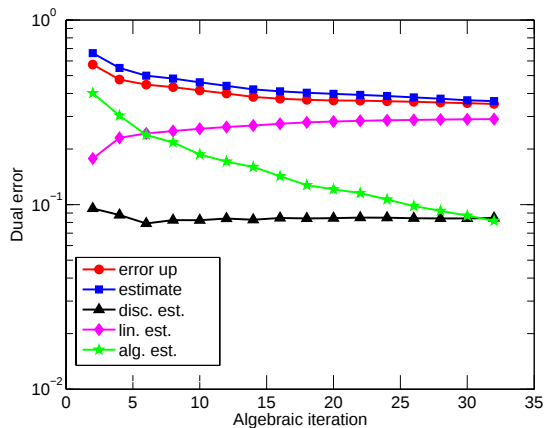
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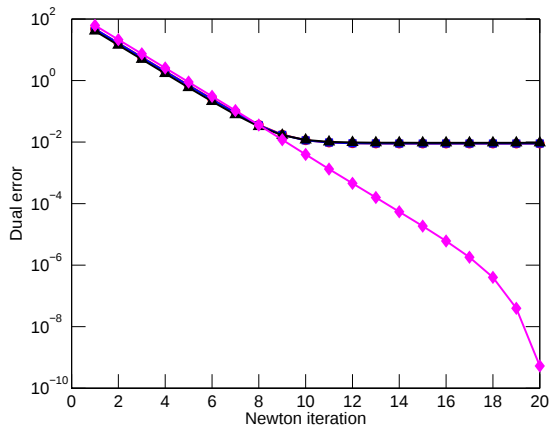
Newton



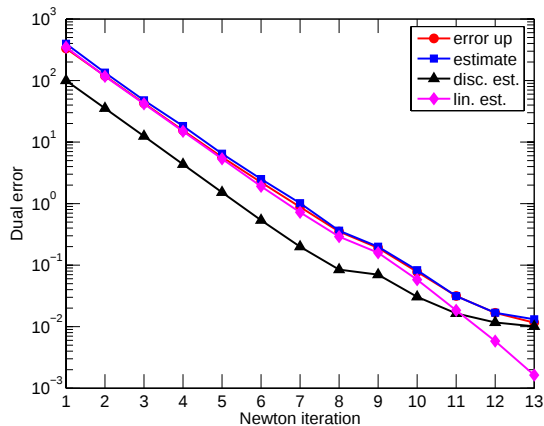
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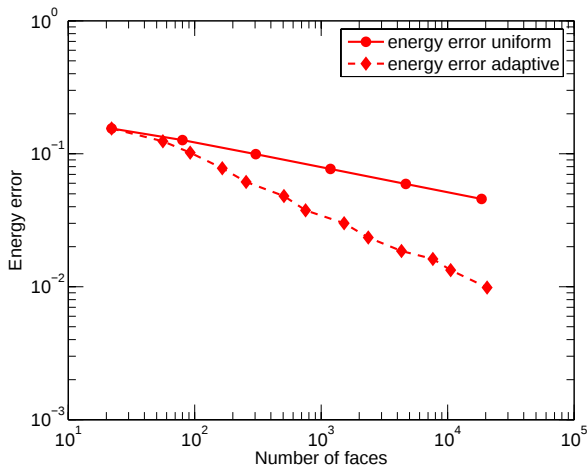
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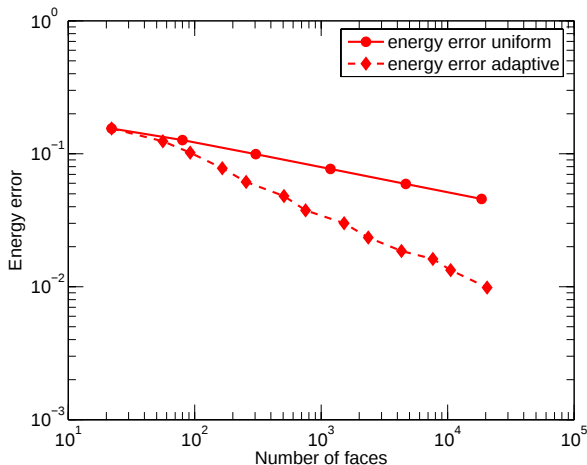
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Optimal decay rate wrt degrees of freedom & computational cost

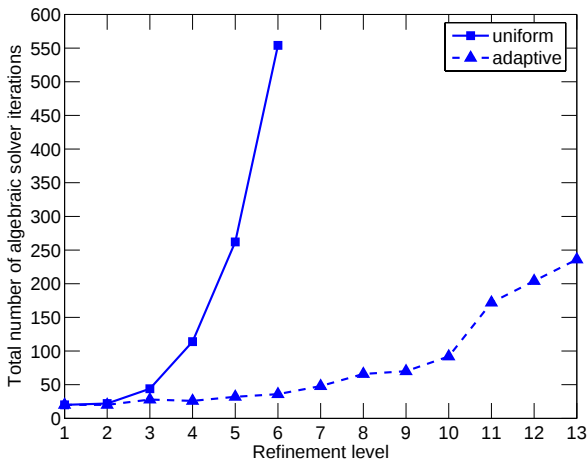


Optimal decay rate wrt DoFs

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Optimal computational cost

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Thank you for your attention!

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- probably **numerical simulations done with insufficient precision**,

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Reliability study and simulation of the progressive collapse of
Roissy Charles de Gaulle Airport

Y. El Kamari^a, W. Raphael^{a,*}, A. Chateaufort^{b,c}

^aÉcole Supérieure d'Ingénieurs de Bayreuth (ESIB), Université Saint-Joseph, CSF Mer Roubaix, PO Box 11-514, Road El-Saïd Beirut 11072050, Lebanon



CDG Terminal 2E collapse in 2004 (opened in 2003)



- no earthquake, flooding, tsunami, heavy rain, extreme temperature
- deterministic, steady problem, PDE known, data known, implementation OK

probably **numerical simulations done with insufficient precision**,
I believe **without error control**

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