



A Convergence Proof for Adaptive Finite Elements without Lower Bound

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Motivation

Most **convergence results** for adaptive finite elements rely on

- **Energy minimization**
 - symmetric elliptic operators
 - p -Laplacian
 - obstacle problems
 - convex minimization

Can be relaxed to disturbed Galerkin Orthogonality.

- **Special properties** of the estimators
 - Discrete local lower bound

- **Dörfler marking**: Given $\theta \in (0, 1]$

$$\text{Select } \mathcal{M} \subset \mathcal{T} : \quad \theta \mathcal{E}_{\mathcal{T}}(\mathcal{T}) \leq \mathcal{E}_{\mathcal{T}}(\mathcal{M})$$

- **Special refinement** of selected elements.

Optimality up to now only for **symmetric elliptic operators**.




Motivation

Convergence and optimality of adaptive finite elements is observed for

- **A larger class of problems**
 - convection-diffusion,
 - saddle point problems,
 - ...
- **Efficient estimators**, where only a continuous lower bound is available.
- **Other marking strategies**
 - Maximum strategy
 - Equidistribution Strategy
 - ...
- **Minimal refinement**.

Convergence in a **rather general setting** by Morin, S., Veeseer '08.

Optimality in this general setting **completely open**.



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Setting of the Basic Convergence Result

Formulation of only **few and basic assumptions** that lead to convergence. These assumptions should be “**necessary**” – at least **reasonable** – and “**easy to verify**” for many problems.


Main Focus in this Talk: Discrete Lower Bound

Previous convergence proofs rely on a **discrete local lower bound**:

- 1 Discrete lower bounds may be **more difficult** to obtain than continuous ones;
- 2 For more complex problems estimators **may not be efficient**, but still we may want to prove convergence.

Reliability of an estimator should be the key property for convergence. **Overestimation** should not forestall convergence:

- 1 Overestimation is a problem for efficiently stopping;
- 2 Overestimation is a problem for optimal complexity.



Problem

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Problem

Variational formulation of a linear, elliptic PDE in a domain $\Omega \subset \mathbb{R}^d$:

$$u \in \mathbb{V} : \quad \mathcal{B}[u, v] = \langle f, v \rangle \quad \forall v \in \mathbb{V}, \quad (\text{P})$$

where


- \mathbb{V} is a real **Hilbert space** with inner product $\langle \cdot, \cdot \rangle_{\mathbb{V}}$, induced norm $\| \cdot \|_{\mathbb{V}}$;
- $\mathcal{B} : \mathbb{V} \times \mathbb{V} \rightarrow \mathbb{R}$ is a **continuous bilinear form**;
- $f \in \mathbb{V}^*$.

Theorem (Niremberg, Nečas, Babuška, Brezzi)

Problem (P) admits for any $f \in \mathbb{V}^$ a unique solution, if and only if \mathcal{B} fulfills an **inf-sup condition**.*

- Coercive forms \mathcal{B} satisfy the inf-sup condition:

$$\mathcal{B}[v, v] \geq c_{\mathcal{B}} \|v\|_{\mathbb{V}}^2 \quad \forall v \in \mathbb{V}.$$



Examples

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Example (Poisson Problem in \mathbb{R}^d)

$$-\Delta u = f \quad \text{in } \Omega, \quad u = 0 \quad \text{on } \partial\Omega.$$


Variational formulation in $\mathbb{V} = H_0^1(\Omega)$:

$$\mathcal{B}[u, v] = \int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx.$$

- \mathcal{B} is continuous and coercive.
- Discretization with continuous Lagrange elements of order $p \geq 1$.
- Global upper bound for the residual estimator build from

$$\mathcal{E}_T^2(T) := h_T^2 \| -\Delta U_T - f \|_{2;T}^2 + h_T \| [U_T] \|_{2;\partial T \cap \Omega}^2.$$

- Continuous and discrete local lower bounds.



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Continuous Local Lower Bound

$$\mathcal{E}_T(T) \lesssim \|U_T - u\|_{\mathbb{V}(\omega(T))} + \text{osc}_T(\omega(T))$$

with $\text{osc}_T(T) = h_T \|f - f_T\|_{2;T}$.

Principal idea by Verfürth: Construct $\phi_T \in \mathbb{V}$ with $\|\phi_T\|_{\mathbb{V}} = 1$, $\text{supp } \phi_T \subset \omega(T)$ such that

$$\mathcal{E}_T(T) \lesssim \langle \mathcal{R}(U_T), \phi_T \rangle := \mathcal{B}[U_T - u, \phi_T] \leq \|\mathcal{B}\| \|U_T - u\|_{\mathbb{V}(\omega(T))}$$

Construction of ϕ_T

- 1 Changing to a computable error indicator leads to potential overestimation.
 - Projection to a finite dimensional space; leads to oscillation.
- 2 Localization by a suitable continuous cut-off function λ_T .



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Discrete Local Lower Bound

Let \mathcal{T}' be a refinement of \mathcal{T} with sufficient refinement around $T \in \mathcal{T}$

$$\mathcal{E}_{\mathcal{T}}(T) \lesssim \|U_{\mathcal{T}} - U_{\mathcal{T}'}\|_{V(\omega(T))} + \text{osc}_{\mathcal{T}}(\omega(T))$$

with $\text{osc}_{\mathcal{T}}(T) = h_T \|f - f_{\mathcal{T}}\|_{2;T}$.

Principal idea by Dörfler and Morin, Nochetto, S.: Construct $\Phi_T \in \mathbb{V}(T')$ with $\|\Phi_T\|_{\mathbb{V}} = 1$, $\text{supp } \Phi_T \subset \omega(T)$ such that

$$\mathcal{E}_{\mathcal{T}}(T) \lesssim \langle \mathcal{R}(U_{\mathcal{T}}), \Phi_T \rangle = \mathcal{B}[U_{\mathcal{T}} - U_{\mathcal{T}'}, \Phi_T] \leq \|\mathcal{B}\| \|U_{\mathcal{T}} - U_{\mathcal{T}'}\|_{V(\omega(T))}$$

Construction of Φ_T

- 1 Projection to a finite dimensional space; leads to oscillation.
 - 2 Localization by a suitable discrete cut-off function Λ_T .
- Projection is limited by the degree of the FE space and the discrete cut-off function.
 - Utilizing a discrete cut-off function is not always possible: A localized function has to be constructed explicitly.



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Example (Eddy Current Equations in \mathbb{R}^3)

$$\text{curl curl } \mathbf{u} + \mathbf{u} = \mathbf{f} \quad \text{in } \Omega, \quad \mathbf{u} \wedge \mathbf{n} = 0 \quad \text{on } \partial\Omega.$$

Variational formulation in $\mathbb{V} = H_0(\text{curl}; \Omega)$:

$$\mathcal{B}[\mathbf{u}, \mathbf{v}] := \int_{\Omega} \text{curl } \mathbf{u} \text{ curl } \mathbf{v} + \mathbf{u} \cdot \mathbf{v} \, dx = \langle \mathbf{f}, \mathbf{v} \rangle \quad \forall \mathbf{v} \in \mathbb{V}.$$

- \mathcal{B} is continuous and coercive;
- Discretization by Nedelec Elements of any order p ;
- Global upper bound for any order;
- Continuous local lower bound for any order;
- Discrete local lower bound available only for lowest order, i. e., for the Whitney Elements.



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Example ($H(\text{div}; \Omega)$ Elliptic Operator in \mathbb{R}^d , $d = 2, 3$)

$$-\nabla \text{div } \mathbf{u} + \mathbf{u} = \mathbf{f} \quad \text{in } \Omega, \quad \mathbf{u} \cdot \mathbf{n} = 0 \quad \text{on } \partial\Omega.$$

Variational formulation in $\mathbb{V} = H_0(\text{div}; \Omega)$:

$$\mathcal{B}[\mathbf{u}, \mathbf{v}] := \int_{\Omega} \text{div } \mathbf{u} \text{ div } \mathbf{v} + \mathbf{u} \cdot \mathbf{v} \, dx = \langle \mathbf{f}, \mathbf{v} \rangle \quad \forall \mathbf{v} \in \mathbb{V}.$$

- \mathcal{B} is continuous and coercive;
- Discretization by Raviart-Thomas or Brezzi-Douglas-Marini Elements of any order p ;
- Global upper bound for any order;
- Continuous and discrete local lower bound for any order:
 - the projection in the discrete lower bound for Raviart-Thomas Elements of order $p \geq 2$ is sub-optimal.



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Example (The Stokes Problem)

Variational formulation in $\mathbb{V} = H_0^1(\Omega; \mathbb{R}^d) \times L_0^2(\Omega)$:

$$\mathcal{B}[(\mathbf{u}, p), (\mathbf{v}, q)] := \int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v} \, dx - \int_{\Omega} p \nabla \cdot \mathbf{v} \, dx - \int_{\Omega} \nabla \cdot \mathbf{u} \, q \, dx = \langle \mathbf{f}, \mathbf{v} \rangle$$

for all $(\mathbf{v}, q) \in \mathbb{V}$.

- \mathcal{B} is continuous and fulfills the inf-sup condition.
- Discretization by the Taylor-Hood Elements of order $p \geq 2$.
- Global upper bound for

$$\mathcal{E}_{\mathcal{T}}^2(T) := h_T^2 \| -\Delta \mathbf{U}_{\mathcal{T}} + \nabla P_{\mathcal{T}} - \mathbf{f} \|_{2;T}^2 + h_T \| [\mathbf{U}_{\mathcal{T}}] \|_{2;\partial T \cap \Omega}^2 + \| \text{div } \mathbf{U}_{\mathcal{T}} \|_{2;T}^2$$

and

$$\mathcal{E}_{\mathcal{T}}^2(T) := h_T^2 \| -\Delta \mathbf{U}_{\mathcal{T}} + \nabla P_{\mathcal{T}} - \mathbf{f} \|_{2;T}^2 + h_T \| [\mathbf{U}_{\mathcal{T}}] \|_{2;\partial T \cap \Omega}^2.$$

- Continuous local lower bound for both variants.
- Discrete local lower bound available only for the second variant.



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Example (The Biharmonic Equation in \mathbb{R}^2)

$$\Delta^2 u \text{ in } \Omega, \quad u = \nabla u \cdot \mathbf{n} = 0 \text{ on } \partial\Omega.$$

Variational formulation in $\mathbb{V} = H_0^2(\Omega)$:

$$\mathcal{B}[u, v] := \int_{\Omega} \Delta u \Delta v \, dx = \langle f, v \rangle \quad \forall v \in \mathbb{V}.$$

- \mathcal{B} is continuous and coercive.
- Discretization by the Argyris Triangle: piecewise \mathbb{P}_5 and H^2 conforming.
- Global upper bound.
- Continuous local lower bound.
- No discrete local lower bound available, seems to be tough.



Adaptive Loop and Basic Assumptions

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Starting with an **initial, conforming triangulation** \mathcal{T}_0 of Ω , the standard adaptive loop SEMR

SOLVE \longrightarrow ESTIMATE \longrightarrow MARK \longrightarrow REFINES

produces a sequence

$$\{\mathcal{T}_k, \mathbb{V}_k, U_k, \{\mathcal{E}_k(T)\}_{T \in \mathcal{T}_k}, \mathcal{M}_k\}_k,$$

where

- \mathcal{T}_k is a conforming triangulation produced by refinement of $\mathcal{T}_{k-1}, \dots, \mathcal{T}_0$;
- $\mathbb{V}_k = \mathbb{V}(\mathcal{T}_k)$ is a finite element space over \mathcal{T}_k ;
- $U_k \in \mathbb{V}_k$ is the unique **Ritz-Galerkin solution**:

$$U_k \in \mathbb{V}_k : \quad \mathcal{B}[U_k, V] = \langle f, V \rangle \quad \forall V \in \mathbb{V}_k, \quad (\mathbf{P}_k)$$

which requires a **discrete inf-sup condition**;

- $\mathcal{E}_k(T)$ is an **error indicator** associated with an element $T \in \mathcal{T}_k$;
- $\mathcal{M}_k \subset \mathcal{T}_k$ is the set of **selected elements for refinement**.



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Assumptions on Refinement

Use **bisectional refinement** and denote by \mathbb{T} the set of all possible, conforming refinements of \mathcal{T}_0 .

- Refinement can be generalized to more general grids and quasi-regular element subdivisions that generate locally quasi-uniform grids.

Assumptions on Finite Element Spaces

The **finite element spaces** have the following properties:

- 1 for any $\mathcal{T} \in \mathbb{T}$, $\mathbb{V}(\mathcal{T}) \subset \mathbb{V}$ is a **conforming finite dimensional space**;
- 2 the spaces are **nested**: if \mathcal{T}' is a refinement of \mathcal{T} then $\mathbb{V}(\mathcal{T}) \subset \mathbb{V}(\mathcal{T}')$;
- 3 the spaces satisfy a **uniform discrete inf-sup condition**.

- Nesting of spaces follows from properties of refinement in combination with appropriate local function spaces.
- Coercivity of \mathcal{B} implies the uniform inf-sup condition.



Convergence of Mesh Size Functions

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Define the local mesh size function $h_k \in L^\infty(\Omega)$ by

$$h_k|_T := |T|^{1/d} \approx \text{diam}(T) \quad \forall T \in \mathcal{T}_k.$$

Lemma (Morin, S. Veeger '08)

For any realization of SEMR there exists a unique $h_\infty \in L^\infty(\Omega)$ such that

$$\lim_{k \rightarrow \infty} \|h_k - h_\infty\|_{\infty; \Omega} = 0.$$

Idea of the Proof.

For any $x \in \Omega$ the sequence $\{h_k(x)\}_k$ is monotone and bounded from below:

$$h_\infty(x) := \lim_{k \rightarrow \infty} h_k(x) \geq 0 \quad \text{exists for all } x \in \Omega.$$

Convergence in L^∞ the follows from

$$T \text{ is refined into } T_1, T_2 \implies |T_1| = |T_2| = \frac{1}{2} |T|. \quad \square$$



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In general, $h_\infty \not\equiv 0$ in Ω . If $h_\infty(x) > 0$, then there is an element $T \ni x$ and $K = K(x)$ such that

$$T \in \mathcal{T}_k \quad \forall k > K.$$

Splitting of \mathcal{T}_k

- 1 Set of elements that are **not refined anymore**

$$\mathcal{T}_k^+ := \{T \in \mathcal{T}_k \mid T \in \mathcal{T}_\ell \ \forall \ell \geq k\};$$

- 2 Set of elements that are **refined at least once**

$$\mathcal{T}_k^0 := \mathcal{T}_k \setminus \mathcal{T}_k^+.$$

Corollary (Morin, S. Veerer '08)

The mesh size functions vanish uniformly in $\Omega_k^0 = \Omega(\mathcal{T}_k^0) := \bigcup\{T : T \in \mathcal{T}_k^0\}$:

$$\lim_{k \rightarrow \infty} \|h_k\|_{\infty; \Omega_k^0} = 0.$$



Convergence of Galerkin Solutions

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Lemma (Morin, S. Veerer '08)

For any realization of SEMR there exists a unique $u_\infty \in \mathbb{V}$ such that

$$\lim_{k \rightarrow \infty} \|U_k - u_\infty\|_{\mathbb{V}} = 0.$$

Proof for coercive \mathcal{B} .

The space

$$\mathbb{V}_\infty = \overline{\bigcup_k \mathbb{V}_k}^{\|\cdot\|_{\mathbb{V}}}$$

is a **closed subspace** of \mathbb{V} . The **Lax-Milgram theorem** then implies the existence of a unique solution u_∞ to

$$u_\infty \in \mathbb{V}_\infty : \quad \mathcal{B}[u_\infty, v] = \langle f, v \rangle \quad \forall v \in \mathbb{V}_\infty.$$

Convergence follows from the **quasi-best approximation property**

$$\|U_k - u_\infty\|_{\mathbb{V}} \leq c_{\mathcal{B}}^{-1} \|\mathcal{B}\| \min_{V \in \mathbb{V}_k} \|V - u_\infty\|_{\mathbb{V}} \rightarrow 0 \quad \text{as } k \rightarrow \infty$$

by construction of \mathbb{V}_∞ . □



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Consequences for a Convergence Proof

It suffices to show $u_\infty = u$ since convergence

$$\lim_{k \rightarrow \infty} U_k \rightarrow u_\infty \quad \text{in } \mathbb{V}$$

is established for any adaptive iteration SEMR.

The residual $\mathcal{R}(u_\infty)$ and $u_\infty = u$

Using the residual $\mathcal{R}(w) \in \mathbb{V}^*$ defined by

$$\mathcal{R}(w) := \mathcal{B}[w - u, v] = \mathcal{B}[w, v] - \langle f, v \rangle \quad \forall v, w \in \mathbb{V}.$$

we reformulate

$$u_\infty = u \quad \iff \quad \mathcal{R}(u_\infty) = 0 \quad \text{in } \mathbb{V}^*$$

- 1 In case $\mathbb{V}_\infty = \mathbb{V}$ **definition of u_∞** implies $\mathcal{R}(u_\infty) = 0$.
- 2 In case $\mathbb{V}_\infty \neq \mathbb{V}$ **properties of ESTIMATE and MARK** have to yield $\mathcal{R}(u_\infty) = 0$.



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Local Approximation Property of the Finite Element Spaces

Let $\mathbb{W} \subset \mathbb{V}$ be **dense**, $q > 0$. Assume that for any $T \in \mathbb{T}$ there exists an interpolation operator $I_T : \mathbb{W} \rightarrow \mathbb{V}(T)$ such that for all $w \in \mathbb{W}$

$$\|w - I_T w\|_{\mathbb{V}(T)} \lesssim \|h_T^q\|_{\infty; T} \|w\|_{\mathbb{W}(T)} \quad \forall T \in \mathbb{T}.$$

Claim

$$\mathbb{V}_\infty = \mathbb{V} \quad \iff \quad h_\infty \equiv 0 \quad \text{in } \Omega$$

- 1 $h_\infty \neq 0$: Then $\mathcal{T}_k^+ \neq \emptyset$ for $k \geq K$ which implies $\mathbb{V} \not\subset \mathbb{V}_\infty$.
- 2 $h_\infty \equiv 0$: Use density of finite element spaces: for $v \in \mathbb{V}$ and $w \in \mathbb{W}$ estimate

$$\begin{aligned} \|v - I_k w\|_{\mathbb{V}(\Omega)} &\leq \|v - w\|_{\mathbb{V}(\Omega)} + \|w - I_k w\|_{\mathbb{V}(\Omega)} \\ &\lesssim \|v - w\|_{\mathbb{V}(\Omega)} + \|h_k\|_{\infty; \Omega} \|w\|_{\mathbb{W}(\Omega)} \stackrel{!}{\leq} \varepsilon \end{aligned}$$

by first choosing w close to v and then k large.



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For $h_\infty \neq 0$ we still obtain for any $v \in \mathbb{V}$ and $w \in \mathbb{W}$

$$\begin{aligned} \|v - I_k w\|_{\mathbb{V}(\Omega_k^0)} &\leq \|v - w\|_{\mathbb{V}(\Omega_k^0)} + \|w - I_k w\|_{\mathbb{V}(\Omega_k^0)} \\ &\lesssim \|v - w\|_{\mathbb{V}(\Omega)} + \|h_k\|_{\infty; \Omega_k^0} \|w\|_{\mathbb{W}(\Omega)} \stackrel{!}{\leq} \varepsilon \end{aligned}$$

by first choosing w close to v and then k large, thanks to

$$\|h_k\|_{\infty; \Omega_k^0} \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

Remarks

- This local density property we are going to use **explicitly** in the convergence proof. It replaces a **(discrete) local lower bound**.
 - Needs a way to build in local features via the upper bound!
- The local density property is already **implicitly** used in all other convergence proofs.



Prior Results

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Directly Related Convergence Results

- Babuška, Vogelius '86: $u'' = f$ in 1d, convergence
- Dörfler '96: Poisson problem in 2d, convergence into tolerance
- Morin, Nochetto, S. '00, '02: constant coefficient matrix, convergence
- Veeseer '02: p -Laplacian
- S. Veeseer '06: obstacle problem
- S. Veeseer '06: convergence for the equidistribution strategy
- Morin, S. Veeseer '08: general convergence with discrete lower bound

Convergence and Optimality Results

- Binev, Dahmen, DeVore '02: MNS with coarsening
- Stevenson '06: Modification of Dörfler
- Cascon, Kreuzer, Nochetto, S. '08: Plain SEMR
- Chen, Holst, Xu '08: Mixed formulation of Poisson problem



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Convergence proof without lower bound for **symmetric elliptic problems**:

$$\mathcal{B}[v, w] := \int_{\Omega} \nabla v^T \mathbf{A} \nabla w + c v w \, dx \quad v, w \in \mathbb{V} := H_0^1(\Omega)$$

with the residual estimator

$$\mathcal{E}_T^2(T) := \|h_T(-\operatorname{div}(\mathbf{A} \nabla U_T + c U_T - f))\|_{2;T}^2 + \|h_T^{1/2} [\mathbf{A} \nabla U_T]\|_{2; \partial T}^2$$

and Dörfler marking with $0 < \theta \leq 1$

$$\text{Choose } \mathcal{M} \subset \mathcal{T} : \quad \theta \mathcal{E}_T(\mathcal{T}) \leq \mathcal{E}_T(\mathcal{M}).$$

Theorem (Cascon, Kreuzer, Nochetto, S. '08)

SEMR is a **contraction**, i. e., there exists $0 < \alpha < 1$ and $\beta > 0$ such that

$$\|U_k - u\|_{\Omega}^2 + \beta \mathcal{E}_k(\mathcal{T}_k) \leq \alpha (\|U_{k-1} - u\|_{\Omega}^2 + \beta \mathcal{E}_{k-1}(\mathcal{T}_{k-1})).$$

If, in addition, θ is sufficiently small and \mathcal{M}_k minimal, then SEMR is **quasi-optimal** in terms of DOFs.

- Optimality proof utilizes the **global continuous lower bound**.



Error Estimation and Marking

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Assumptions on the Estimator

- We assume an upper bound with the following build-in localization: For any subset $\mathcal{S} \subset \mathcal{T}$ holds:

$$|\langle \mathcal{R}(U_T), v \rangle| \lesssim \mathcal{E}_T(\mathcal{S}) \|v\|_{\Omega(\mathcal{S})} + \mathcal{E}_T(\mathcal{T} \setminus \mathcal{S}) \|v\|_{\Omega(\mathcal{T} \setminus \mathcal{S})} \quad \forall v \in \mathbb{V}.$$

- We assume stability of the indicators: there exists $D \in L^2(\Omega)$ such that

$$\mathcal{E}_T(\mathcal{T}) \lesssim \|U_T\|_{\mathbb{V}(\mathcal{T})} + \|D\|_{2;T} \quad \forall \mathcal{T} \in \mathcal{T}$$

Remarks

- The continuous inf-sup condition and the upper bound for $\mathcal{S} = \mathcal{T}$ imply

$$\|U_T - u\|_{\mathbb{V}} \lesssim \|\mathcal{R}(U_T)\|_{\mathbb{V}^*} = \sup_{\|v\|_{\mathbb{V}}=1} |\langle \mathcal{R}(U_T), v \rangle| \lesssim \mathcal{E}_T(\mathcal{T}).$$

- Boundedness of $\{U_k\}_k$ and stability of the indicators yield

$$\sup_k \mathcal{E}_k(\mathcal{T}_k) \lesssim 1.$$



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Assumption on Marking

We assume the existence of $g \in C^0(\mathbb{R}_0^+; \mathbb{R}_0^+)$ with $g(0) = 0$ such that the set of marked elements \mathcal{M} satisfies

$$\mathcal{E}_{\mathcal{T}}(T) \leq g(\max\{\mathcal{E}_{\mathcal{T}}(T) \mid T \in \mathcal{M}\}) \quad \forall T \in \mathcal{T} \setminus \mathcal{M}.$$

Additional Assumption on Refinement

All marked elements are refined at least once.

Remarks

- 1 The assumption on marking includes standard marking strategies like **Maximum**, **Equidistribution** and **Minimal Dörfler marking** with $g(s) = s$.
- 2 Assumption on refinement implies $\mathcal{M}_k \subset \mathcal{T}_k^0$.
- 3 Convergence of the Galerkin Solutions, stability of the indicators, and assumption on marking and refinement yield

$$\max\{\mathcal{E}_k(T) \mid T \in \mathcal{T}_k\} \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$



Convergence of the Error

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Theorem (S. '08)

Assume that the above assumptions on refinement, finite element spaces, estimator, and marking are satisfied. Then SEMR converge, i. e.,

$$\lim_{k \rightarrow \infty} \|U_k - u\|_{\mathbb{V}} = 0.$$

Proof

Since $U_k \rightarrow u_{\infty}$ in \mathbb{V} , it remains to show

$$\langle \mathcal{R}(u_{\infty}), v \rangle = 0 \quad \forall v \in \mathbb{V} \quad \iff \quad \langle \mathcal{R}(u_{\infty}), w \rangle = 0 \quad \forall w \in \mathbb{W},$$

by density of \mathbb{W} in \mathbb{V} . Using continuity of $\mathcal{R}: \mathbb{V} \rightarrow \mathbb{V}^*$ this reduces to

$$\lim_{k \rightarrow \infty} \langle \mathcal{R}(U_k), w \rangle = 0 \quad \forall w \in \mathbb{W}, \|w\|_{\mathbb{W}} = 1.$$

The sets \mathcal{T}_k^+ are nested, which grants for $k \geq \ell$

$$\mathcal{T}_{\ell}^+ \subset \mathcal{T}_k^+ \subset \mathcal{T}_k \quad \text{and} \quad \Omega_{\ell}^0 = \Omega(\mathcal{T}_{\ell}^0) = \Omega(\mathcal{T}_k \setminus \mathcal{T}_{\ell}^+).$$



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Proof (continued)

Use the upper bound with $\mathcal{S} = \mathcal{T}_k \setminus \mathcal{T}_{\ell}^+$ for $w \in \mathbb{W}$, $\|w\|_{\mathbb{W}} = 1$

$$\begin{aligned} |\langle \mathcal{R}(U_k), w \rangle| &= |\langle \mathcal{R}(U_k), w - I_k w \rangle| \\ &\lesssim \mathcal{E}_k(\mathcal{T}_k \setminus \mathcal{T}_{\ell}^+) \|w - I_k w\|_{\mathbb{V}(\Omega_{\ell}^0)} + \mathcal{E}_k(\mathcal{T}_{\ell}^+) \|w - I_k w\|_{\mathbb{V}(\Omega_{\ell}^+)}, \\ &\lesssim \|h_k\|_{\infty; \Omega_{\ell}^0} + \mathcal{E}_k(\mathcal{T}_{\ell}^+) \stackrel{!}{\leq} \varepsilon \end{aligned}$$

- 1 Choose ℓ sufficiently large such that

$$\|h_k\|_{\infty; \Omega_{\ell}^0} \leq \|h_{\ell}\|_{\infty; \Omega_{\ell}^0} \leq \frac{\varepsilon}{2}.$$

- 2 Then choose $k \geq \ell$ such that

$$\mathcal{E}_k(T) \leq \frac{\varepsilon}{2} (\#\mathcal{T}_{\ell}^+)^{-1/2} \quad \forall T \in \mathcal{T}_{\ell}^+,$$

which implies

$$\mathcal{E}_k(\mathcal{T}_{\ell}^+) \leq \frac{\varepsilon}{2}. \quad \square$$



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Remark

The theorem does not imply convergence of the estimator, since it includes non-efficient estimators and allows for strong overestimation!

Continuous Lower Bound

Let the indicators satisfy

$$\mathcal{E}_{\mathcal{T}}(T) \lesssim \|U_{\mathcal{T}} - u\|_{\mathbb{V}(\omega(T))} + \text{osc}_{\mathcal{T}}(\omega(T)),$$

where **oscillation** can be estimated by

$$\text{osc}_{\mathcal{T}}(T) \lesssim \|h_{\mathcal{T}}^r\|_{\infty; T} (\|U_{\mathcal{T}}\|_{\mathbb{V}(\omega(T))} + \|D\|_{2; \omega(T)})$$

for some $r > 0$ and $D \in L^2(\Omega)$.

Corollary (S. '08)

If, in addition, the estimator satisfies the continuous local lower bound, then SEMR yields

$$\lim_{k \rightarrow \infty} \mathcal{E}_k(\mathcal{T}_k) = 0.$$



Proof

As in the previous proof we split for $k \geq \ell$

$$\mathcal{E}_k(\mathcal{T}_k) \lesssim \mathcal{E}_k(\mathcal{T}_k \setminus \mathcal{T}_\ell^0) + \mathcal{E}_k(\mathcal{T}_\ell^+) \lesssim \|U_k - u\|_{\mathbb{V}(\Omega_\ell^0)} + \text{osc}_k(\Omega_\ell^0) + \mathcal{E}_k(\mathcal{T}_\ell^+). \quad (*)$$

- 1 The error is controlled by the previous theorem:

$$\|U_k - u\|_{\mathbb{V}(\Omega_\ell^0)} \leq \|U_k - u\|_{\mathbb{V}} \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

- 2 Oscillation can be estimated in Ω_ℓ^0 by assumption in an a priori way:

$$\begin{aligned} \text{osc}_k(\Omega_\ell^0) &\lesssim \|h_\ell^r\|_{\infty; \Omega_\ell^0} (\|U_k\|_{\mathbb{V}} + \|D\|_{L^2(\Omega)}) \\ &\lesssim \|h_\ell^r\|_{\infty; \Omega_\ell^0} \rightarrow 0 \quad \text{as } \ell \rightarrow \infty. \end{aligned}$$

- 3 The remaining part of the estimator can be handled as before:

$$\mathcal{E}_k(\mathcal{T}_\ell^+) \rightarrow 0 \quad \text{for } \ell \text{ fixed and } k \rightarrow \infty.$$

Summarizing: The right hand side of (*) can be made arbitrarily small by first choosing ℓ large and then $k \geq \ell$ even larger. \square



- 1 General convergence proof for adaptive finite elements with mild assumptions on the ingredients, most easy to verify.
- 2 Convergence does not need the lower bound, "practical" convergence and **convergence into tolerance** need efficient estimators:

- Includes strategies, where the given **tolerance enters the selection**, like the equidistribution strategy:

$$\mathcal{M} = \{T \in \mathcal{T} \mid \mathcal{E}_T(T) \geq \theta \text{TOL} (\#T)^{-1/2}\},$$

- 3 For efficient estimators, the assumption on marking can be generalized such that it is **essentially necessary**:

$$\text{if } \lim_{k \rightarrow \infty} \max\{\mathcal{E}_k(T) \mid T \in \mathcal{M}_k\} = 0$$

$$\text{then } \forall T \in \mathcal{T}^+ : \lim_{k \rightarrow \infty} \mathcal{E}_k(T) = 0,$$

where

$$\mathcal{T}^+ = \bigcup_{k \geq 0} \bigcap_{\ell \geq k} \mathcal{T}_\ell$$

is the set of elements that are not refined.