

Convergence and Optimality of an AFEM for General Elliptic Operators

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Multiscale Problems: Modeling, Adaptive Discretization,
Stabilization, and Solvers.
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Outline

1 Main Results

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- 2 Morin-Nochetto-Siebert (MNS) and Stevenson Algorithms

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Main Results

$$\begin{aligned}\mathcal{L}u &:= -\operatorname{div} \mathbf{A} \nabla u + \mathbf{b} \cdot \nabla u + c u = f && \text{in } \Omega \\ u &= 0 && \text{on } \partial\Omega\end{aligned}$$

- Ω polyhedral bounded domain in \mathbb{R}^d ($d=2,3$).
- $\mathbf{A} : \Omega \mapsto \mathbb{R}^{d \times d}$ is piecewise Lipschitz in \mathcal{T}_0 and symmetric positive definite.
- $\mathbf{b} \in [L^\infty(\Omega)]^d$ (in general $\operatorname{div} \mathbf{b} \neq 0$ in Ω).
- $c \in L^\infty(\Omega)$ ($c \geq 0$ in Ω).
- $f \in L^2(\Omega)$.

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- **Theorem** [Convergence of AFEM] There exists $0 < \alpha < 1$ such that

$$\|u - u_{k+1}\|_{\Omega}^2 + \operatorname{osc}(u_{k+1}, \mathcal{T}_{k+1})^2 \leq \alpha^2 \left\{ \|u - u_k\|_{\Omega}^2 + \operatorname{osc}(u_k, \mathcal{T}_k)^2 \right\}.$$

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- **Theorem [Optimality of AFEM]** Let $(u, f, \mathbf{A}, \mathbf{b}, c)$ satisfy PDE and for all $\epsilon > 0$

$$\exists \mathcal{T} \supset \mathcal{T}_0 : \inf_{v_{\mathcal{T}} \in \mathbb{V}_{\mathcal{T}}} \left\{ \|u - v_{\mathcal{T}}\|_{\Omega} + \operatorname{osc}(v_{\mathcal{T}}, \mathcal{T}) \right\} \leq \epsilon \quad \& \quad \#\mathcal{T} - \#\mathcal{T}_0 \preccurlyeq \epsilon^{-1/s}.$$

Then

$$\|u - u_k\|_{\Omega} + \operatorname{osc}(u_k, \mathcal{T}_k) \preccurlyeq \{\#\mathcal{T}_k - \#\mathcal{T}_0\}^{-s}.$$

Morin-Nochetto-Siebert Algorithm

- 1 $\mathcal{T}_0, u_{-1} = 0, k = 0.$
- 2 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1}).$
- 3 $\{\eta_k, \text{osc}_k\} = \text{ESTIMATE}(\mathcal{T}_k, u_k).$
- 4 $\mathcal{M}_k = \text{MARK}(\eta_k, \text{osc}_k, \theta_{\text{est}}, \theta_{\text{osc}}, \mathcal{T}_k).$
- 5 $\mathcal{T}_{k+1} = \text{REFINE}(\mathcal{T}_k, \mathcal{M}_k).$
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- 7 go to 2.

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- **Procedure** $\text{MARK}(\eta_k, \text{osc}_k, \theta_{\text{est}}, \theta_{\text{osc}}).$

Select $\mathcal{M}_k \subset \mathcal{T}_k$ such that

$$\eta_k(\mathcal{M}_k)^2 \geq \theta_{\text{est}} \eta_k^2,$$

if necessary *enlarge* \mathcal{M}_k to satisfy

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- η_k has **large** indicators in a small area \Rightarrow Dörfler's Property holds for a **small** number of elements.
- $\text{osc}(u_k, \mathcal{T}_k) \ll \eta_k.$
- $\text{osc}(u_k, \mathcal{T}_k)$ is uniformly distributed in $\mathcal{T}_k \Rightarrow$ A **large percentage** of triangles must be marked.

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\Rightarrow **MNS is suboptimal!!**

MNS Algorithm - Counterexample

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- $\mathbf{A} = \sum_{i=1}^4 a_i \mathbf{I}_{\chi_i}$ (Checkerboard pattern).

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- $\mathbf{A} = \sum_{i=1}^4 a_i \mathbf{I} \chi_i$ (Checkerboard pattern).
- (u, f) :

$$u(\mathbf{x}) = u_K(\mathbf{x}) + u_S(\mathbf{x}) \quad f(\mathbf{x}) = f_K(\mathbf{x}) + f_S(\mathbf{x})$$

- ▶ Kellogg's example

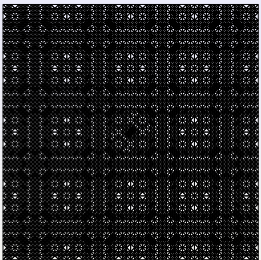
$$u_K(r, \theta) = r^{0.1} \phi(\theta) \quad \Rightarrow \quad f_K = 0$$

- ▶ Smooth function

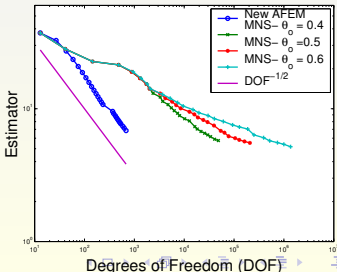
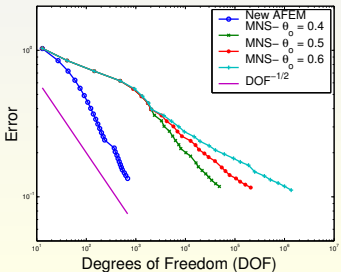
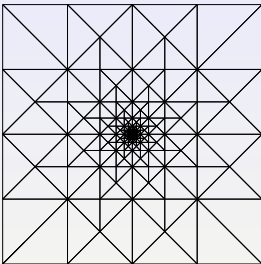
$$u_S(\mathbf{x}) = 10^{-6} \sum_{i=1}^4 a_i^{-1} \chi_i \sin^2(10\pi x) \sin^2(10\pi y) \quad \Rightarrow \quad f_S \text{ smooth function}$$

MNS Algorithm - Counterexample

MNS Algorithm (iter 15, DOFs 8077)



New Algorithm (iter 15, DOFs 368)



Stevenson Algorithm (Simplified)

- 1 $\mathcal{T}_0, u_{-1} = 0, \delta_0 = \|f\|_{H^{-1}},$
 $k = 0, \text{tol} > 0.$
- 2 do $\delta_k = \delta_k/2,$
 $[\mathcal{T}_{k,*}, f_k] = \text{RHS}(\mathcal{T}_k, f, \delta_k)$
 $\mathcal{T}_k = \text{MAKECONFORM}(\mathcal{T}_{k,*})$
 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1})$
 $\eta_k = \text{ESTIMATE}(\mathcal{T}_k, u_k)$
if $C_1\eta_k < \text{tol}$ STOP
until $\delta_k < \omega\eta_k$
- 3 $\mathcal{M}_k = \text{MARK}(\eta_k, \theta_{\text{est}}, \mathcal{T}_k)$
- 4 $\mathcal{T}_{k+1} = \text{REFINE}(\mathcal{T}_k, \mathcal{M}_k)$
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- RHS is not specified for $f \in H^{-1}.$
- The algorithm does not work with general elliptic operators $\mathcal{L}.$
- RHS reduces the oscillation of f relative to the current error indicator $\eta_k.$
- It has to solve the linear system (most expensive task) at all steps of the inner loop.

Preliminary Results

$$\mathcal{T}_0 \rightarrow \{\mathcal{T}_k\}_{k \in \mathbb{N}}.$$

$$\mathbb{V} := H_0^1(\Omega), \quad \|\cdot\|_{\mathbb{V}}.$$

\mathbb{V}_k , Continuous piecewise **polynomials of degree one** over \mathcal{T}_k .

$$\mathcal{B}(u, v) := \langle \mathbf{A} \nabla u, \nabla v \rangle + \langle \mathbf{b} \cdot \nabla u + c u, v \rangle.$$

$$\|v\|_{\Omega} := \mathcal{B}(v, v)^{1/2}.$$

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- **Continuous Problem**

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

- **Discrete Problem**

$$u_k \in \mathbb{V}_k : \quad \mathcal{B}(u_k, v) = \langle f, v \rangle \quad \forall v \in \mathbb{V}_k.$$

Preliminary Results

- **Orthogonality** (Case $\mathbf{b} = 0$).

$$\mathbb{V}_l \subset \mathbb{V}_m \subset \mathbb{V}_n, \quad \mathbb{V}_\infty = \mathbb{V}, \quad u_\infty = u,$$
$$\|u_n - u_m\|_\Omega^2 = \|u_n - u_l\|_\Omega^2 - \|u_m - u_l\|_\Omega^2.$$

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- **A-posteriori Error Estimate and Oscillation** ($v \in \mathbb{V}_k$)

$$R(v) := (f + \operatorname{div} \mathbf{A} \nabla v - \mathbf{b} \cdot \nabla v - cv)|_T,$$

$$J(v) := (\llbracket \mathbf{A} \nabla v \rrbracket \cdot \boldsymbol{\nu})|_{\partial T},$$

$$\eta_k(T)^2 := \|hR(u_k)\|_{L^2(T)}^2 + \|h^{\frac{1}{2}}J(u_k)\|_{L^2(\partial T)}^2,$$

$$\operatorname{osc}(v, T)^2 := \|h_T^2(R(v) - \bar{R}(v))\|_{L^2(T)}^2.$$

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- For any subset $\mathcal{M} \in \mathcal{T}_k$ we use the convention

$$\eta_k(\mathcal{M})^2 = \sum_{T \in \mathcal{M}} \eta_k(T)^2, \quad \operatorname{osc}(v, \mathcal{M})^2 = \sum_{T \in \mathcal{M}} \operatorname{osc}(v, T)^2.$$

- For $\mathcal{M} = \mathcal{T}_k$ we use the notation: $\eta_k^2 = \sum_{T \in \mathcal{M}} \eta_k(T)^2$.

Notation

- $\mathcal{T}_{k,*} \supset \mathcal{T}_k$ nonconforming refinement.
- Marked elements set:
 $\mathcal{M}_{k,*}$: elements of \mathcal{T}_k that give rise to $\mathcal{T}_{k,*}$ after their not conforming refinement.
 $\mathcal{M}_{k,\text{est}} := \text{MARK}(\eta_k, \theta_{\text{est}}, \mathcal{T}_k)$.
 $\mathcal{M}_{k,\text{osc}}^i := \text{MARK}(\text{osc}_k^i, \theta_{\text{osc}}, \mathcal{T}_k^i)$.
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- $\mathbf{D} := (\mathbf{A}, \mathbf{b}, c)$

$$\text{osc}(\mathbf{D}, T)^2 := h_T^2 \max_j \max_{x,y \in T} |\partial_i \{a_{ij}(x) - a_{ij}(y)\}|^2 + \dots$$

$$\text{osc}(\mathbf{D}, \mathcal{T}_k)^2 := \max_{T \in \mathcal{T}_k} \text{osc}(\mathbf{D}, T)^2.$$

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- $e_k := \|\| u - u_k \|\|_{\Omega} \quad \epsilon_k := \|\| u_{k+1} - u_k \|\|_{\Omega}.$

A Posteriori Error Estimates

- **Upper bound**

$$\begin{aligned} \text{Global :} \quad & \|u - u_k\|_{\Omega}^2 \leq C_1 \eta_k^2 \\ \text{Localized :} \quad & \|u_{k,*} - u_k\|_{\Omega}^2 \leq C_1 \eta_k (\overline{\mathcal{M}}_{k,*})^2 \end{aligned}$$

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- **Lower bound**

$$\begin{aligned} \text{Global :} \quad & C_2 \eta_k^2 \leq \|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \\ \text{Localized :} \quad & C_2 \eta_k (\mathcal{M}_{k,*})^2 \leq \|u_{k,*} - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \\ & + \text{Interior Node Property} \end{aligned}$$

A Posteriori Error Estimates

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$$\text{OPTIMALITY} \Leftarrow \text{Localized : } \|u_{k,*} - u_k\|_{\Omega}^2 \leq C_1 \eta_k (\overline{\mathcal{M}}_{k,*})^2$$

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+ Interior Node Property

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Oscillation

We consider $\mathcal{T}_k \subset \mathcal{T}_{k,*}$, $\mathbb{V}_k \subset \mathbb{V}_{k,*}$

- **Local:** $\forall T \in \mathcal{T}_{k,*}$

$$\text{osc}(u_{k,*}, T)^2 \leq 2 \text{osc}(u_k, T)^2 + 2 \text{osc}(\mathbf{D}, T)^2 \|u_{k,*} - u_k\|_{\Omega}^2.$$

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- **Global:**

$$\text{osc}(u_{k,*}, \mathcal{T}_{k,*})^2 \leq 2 \text{osc}(u_k, \mathcal{T}_{k,*})^2 + 2 \text{osc}(\mathbf{D}, \mathcal{T}_{k,*})^2 \|u_{k,*} - u_k\|_{\Omega}^2.$$

By monotonicity,

$$\text{osc}(u_{k,*}, \mathcal{T}_{k,*})^2 \leq 2 \text{osc}(u_k, \mathcal{T}_k)^2 + 2 \text{osc}(\mathbf{D}, \mathcal{T}_k)^2 \|u_{k,*} - u_k\|_{\Omega}^2.$$

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- **SOLVE**: solves the discrete linear system.

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- 3 $\{\eta_k, \text{osc}_k\} = \text{ESTIMATE}(\mathcal{T}_k, u_k).$

- **ESTIMATE**: computes local a posteriori error indicators and oscillation.

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- 2 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1}).$
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- 4 $\mathcal{M}_{k,\text{est}} = \text{MARK}(\eta_k, \theta_{\text{est}}, \mathcal{T}_k).$

- **MARK**: marks elements according to relative size of **error indicators** (Dörfler Marking).

New Algorithm

- 1 $\mathcal{T}_0, u_{-1} = 0, k = 0.$
- 2 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1}).$
- 3 $\{\eta_k, \text{osc}_k\} = \text{ESTIMATE}(\mathcal{T}_k, u_k).$
- 4 $\mathcal{M}_{k,\text{est}} = \text{MARK}(\eta_k, \theta_{\text{est}}, \mathcal{T}_k).$
- 5 if $\text{osc}(u_k, \mathcal{T}_k)^2 > \sigma_k = \delta\eta_k^2$
 $\{\mathcal{T}_k^+, \mathcal{M}_k^+\} = \text{REDUCE-OSC}(\mathcal{T}_k, \mathcal{M}_{k,\text{est}}, \text{osc}(u_k, \mathcal{T}_k), u_k, \sigma_k).$

- **REDUCE-OSC**: reduces the oscillation **only relative** to the estimator η_k . It generates a refinement \mathcal{T}_k^+ satisfying

$$\text{osc}(u_k, \mathcal{T}_k^+)^2 < \delta\eta_k^2.$$

New Algorithm

- 1 $\mathcal{T}_0, u_{-1} = 0, k = 0.$
- 2 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1}).$
- 3 $\{\eta_k, \text{osc}_k\} = \text{ESTIMATE}(\mathcal{T}_k, u_k).$
- 4 $\mathcal{M}_{k,\text{est}} = \text{MARK}(\eta_k, \theta_{\text{est}}, \mathcal{T}_k).$
- 5 if $\text{osc}(u_k, \mathcal{T}_k)^2 > \sigma_k = \delta\eta_k^2$
 $\{\mathcal{T}_k^+, \mathcal{M}_k^+\} = \text{REDUCE-OSC}(\mathcal{T}_k, \mathcal{M}_{k,\text{est}}, \text{osc}(u_k, \mathcal{T}_k), u_k, \sigma_k).$
- 6 $\mathcal{T}_{k+1} = \text{REFINE}(\mathcal{T}_k^+, \mathcal{M}_{k,\text{est}}^+).$

- **REFINE**: refines the mesh \mathcal{T}_k^+ locally according to marked elements $\mathcal{M}_{k,\text{est}}^+$.

New Algorithm

- 1 $\mathcal{T}_0, u_{-1} = 0, k = 0.$
- 2 $u_k = \text{SOLVE}(\mathcal{T}_k, u_{k-1}).$
- 3 $\{\eta_k, \text{osc}_k\} = \text{ESTIMATE}(\mathcal{T}_k, u_k).$
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 $\{\mathcal{T}_k^+, \mathcal{M}_k^+\} = \text{REDUCE-OSC}(\mathcal{T}_k, \mathcal{M}_{k,\text{est}}, \text{osc}(u_k, \mathcal{T}_k), u_k, \sigma_k).$
- 6 $\mathcal{T}_{k+1} = \text{REFINE}(\mathcal{T}_k^+, \mathcal{M}_{k,\text{est}}^+).$
- 7 $k \leftarrow k + 1.$

New Algorithm

- 1 $\mathcal{T}_0, u_{-1} = 0, k = 0.$
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 $\{\mathcal{T}_k^+, \mathcal{M}_k^+\} = \text{REDUCE-OSC}(\mathcal{T}_k, \mathcal{M}_{k,\text{est}}, \text{osc}(u_k, \mathcal{T}_k), u_k, \sigma_k).$
- 6 $\mathcal{T}_{k+1} = \text{REFINE}(\mathcal{T}_k^+, \mathcal{M}_{k,\text{est}}^+).$
- 7 $k \leftarrow k + 1.$
- 8 go to 2.

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

New Algorithm - REDUCE-OSC

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① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

② do

$\mathcal{M}_{\text{osc}}^i = \text{MARK}(\text{osc}^i, \mathcal{T}^i, \theta_{\text{osc}}).$

$\mathcal{M}^+ = \text{UPDATE}(\mathcal{M}^+, \mathcal{M}_{\text{osc}}^i).$

$\mathcal{T}^{i+1} = \text{REFINE}(\mathcal{T}^i, \mathcal{M}_{\text{osc}}^i).$

$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

② do

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$\mathcal{M}^+ = \text{UPDATE}(\mathcal{M}^+, \mathcal{M}_{\text{osc}}^i).$

$\mathcal{T}^{i+1} = \text{REFINE}(\mathcal{T}^i, \mathcal{M}_{\text{osc}}^i).$

$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

- **MARK**: standard **Dörfler marking** of oscillation,

$$\sum_{T \in \mathcal{M}_{\text{osc}}^i} \text{osc}(v, T)^2 \geq \theta_{\text{osc}} \text{osc}(v, \mathcal{T}^i)^2$$

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

② do

$\mathcal{M}_{\text{osc}}^i = \text{MARK}(\text{osc}^i, \mathcal{T}^i, \theta_{\text{osc}}).$

$\mathcal{M}^+ = \text{UPDATE}(\mathcal{M}^+, \mathcal{M}_{\text{osc}}^i).$

$\mathcal{T}^{i+1} = \text{REFINE}(\mathcal{T}^i, \mathcal{M}_{\text{osc}}^i).$

$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

- **UPDATE**: deletes the triangles in \mathcal{M}^+ that belong to $\mathcal{M}^+ \cap \mathcal{M}_{\text{osc}}^i$.

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

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② do

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$\mathcal{M}^+ = \text{UPDATE}(\mathcal{M}^+, \mathcal{M}_{\text{osc}}^i).$

$\mathcal{T}^{i+1} = \text{REFINE}(\mathcal{T}^i, \mathcal{M}_{\text{osc}}^i).$

$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

- **REFINE**: conforming local refinement of \mathcal{T}^i according to $\mathcal{M}_{\text{osc}}^i$.

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

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$\mathcal{T}^{i+1} = \text{REFINE}(\mathcal{T}^i, \mathcal{M}_{\text{osc}}^i).$

$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

- **ESTIMATE-OSC**: computes oscillation indicators
(v is fixed throughout).

New Algorithm - REDUCE-OSC

$$\{\mathcal{T}^+, \mathcal{M}^+\} = \text{REDUCE-OSC}(\mathcal{T}, \mathcal{M}, \text{osc}(v, \mathcal{T}), v, \sigma)$$

① $\text{osc}^0 = \text{osc}(v, \mathcal{T}), \mathcal{T}^0 = \mathcal{T}, \mathcal{M}^+ = \mathcal{M}, i = 0.$

② do

$\mathcal{M}_{\text{osc}}^i = \text{MARK}(\text{osc}^i, \mathcal{T}^i, \theta_{\text{osc}}).$

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$\text{osc}^{i+1} = \text{ESTIMATE-OSC}(\mathcal{T}^i, v)$

$i \leftarrow i + 1.$

while($\text{osc}^i > \sigma^{1/2}$).

③ $\mathcal{T}^+ = \mathcal{T}^i.$

Convergence (Case $\mathbf{b} = 0$)

Theorem[Convergence] Let $\{u_k\}_k$ be the sequence generated by AFEM. If $\delta = \delta(C_1, C_2, \theta_{\text{est}})$ is adequately chosen and

$$\text{osc}(\mathbf{D}, \mathcal{T}_0)^2 \leq \frac{\theta_{\text{est}} C_2}{8C_1},$$

then there exists $0 < \alpha < 1$ depending on \mathcal{T}_0 and θ_{est} such that

$$\|u - u_{k+1}\|_{\Omega}^2 + \text{osc}(u_{k+1}, \mathcal{T}_{k+1})^2 \leq \alpha^2 \left\{ \|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right\}.$$

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Proof: Outline:

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Proof: Outline:

- Estimate e_{k+1}^2 in terms of e_k^2 , $\text{osc}(u_k, \mathcal{T}_k)^2$.

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Proof: Outline:

- Estimate e_{k+1}^2 in terms of e_k^2 , $\text{osc}(u_k, \mathcal{T}_k)^2$.
- Estimate $\text{osc}(u_{k+1}, \mathcal{T}_{k+1})^2$ in terms of e_k^2 , $\text{osc}(u_k, \mathcal{T}_k)^2$.

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Proof: Outline:

- Estimate e_{k+1}^2 in terms of e_k^2 , $\text{osc}(u_k, \mathcal{T}_k)^2$.
- Estimate $\text{osc}(u_{k+1}, \mathcal{T}_{k+1})^2$ in terms of e_k^2 , $\text{osc}(u_k, \mathcal{T}_k)^2$.
- Sum and determine $0 < \alpha < 1$.

Optimality of AFEM

Theorem [Optimality of AFEM] Let $(u, \mathbf{A}, \mathbf{b}, c, f)$ satisfy PDE and for all $\epsilon > 0$

$$\exists \mathcal{T} \supset \mathcal{T}_0 : \inf_{v_{\mathcal{T}} \in \mathbb{V}_{\mathcal{T}}} \left\{ \|u - v_{\mathcal{T}}\|_{\Omega} + \text{osc}(v_{\mathcal{T}}, \mathcal{T}) \right\} \leq \epsilon \quad \& \quad \#\mathcal{T} - \#\mathcal{T}_0 \preccurlyeq \epsilon^{-1/s}.$$

Then

$$\|u - u_k\|_{\Omega} + \text{osc}(u_k, \mathcal{T}_k) \preccurlyeq \left\{ \#\mathcal{T}_k - \#\mathcal{T}_0 \right\}^{-s}.$$

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$$\Leftrightarrow \#\mathcal{T}_k - \#\mathcal{T}_0 \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_{\Omega} + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

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Definition. For $\epsilon > 0$, let \mathcal{P}_{ϵ} be the set of all nonconforming partitions \mathcal{T} of \mathcal{T}_0 such that

$$\mathcal{P}_{\epsilon} = \left\{ \mathcal{T} \supset \mathcal{T}_0 : \inf_{v \in \mathbb{V}_{\mathcal{T}}} \left\{ \|u - v_{\mathcal{T}}\|_{\Omega} + \text{osc}(v_{\mathcal{T}}, \mathcal{T}) \right\} < \epsilon \right\},$$

Optimality of AFEM

Theorem [Optimality of AFEM] Let $(u, \mathbf{A}, \mathbf{b}, c, f)$ satisfy PDE and for all $\epsilon > 0$

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and define

$$|u, f, \mathbf{D}|_s := \sup_{\epsilon > 0} \left\{ \epsilon \inf_{\mathcal{T} \in \mathcal{P}_{\epsilon}} (\#\mathcal{T} - \#\mathcal{T}_0)^s \right\}$$

Optimality of AFEM

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \asymp |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Proof: Outline

Optimality of AFEM

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Proof: Outline

- **Theorem 0.** (Binev, Dahmen, DeVore).

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \preccurlyeq \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

Optimality of AFEM

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- **Theorem 1.** (Upper bound of DOFs added by η_k)

$$\#\mathcal{M}_{k,\text{est}} \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Optimality of AFEM

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

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- **Theorem 1.** (Upper bound of DOFs added by η_k)

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- **Theorem 2.** (Upper bound of DOFs added by $\text{osc}(u_k, \mathcal{T}_k)$)

$$\#\mathcal{M}_{k,\text{osc}}^i \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \text{osc}(u_k, \mathcal{T}_k^i)^{-1/s}$$

Optimality of AFEM

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \lesssim |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Proof: Outline

- **Theorem 0.** (Binev, Dahmen, DeVore).

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \lesssim \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

- **Theorem 1.** (Upper bound of DOFs added by η_k)

$$\#\mathcal{M}_{k,\text{est}} \lesssim |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

- **Theorem 2.** (Upper bound of DOFs added by $\text{osc}(u_k, \mathcal{T}_k)$)

$$\#\mathcal{M}_{k,\text{osc}}^i \lesssim |u, f, \mathbf{D}|_s^{1/s} \text{osc}(u_k, \mathcal{T}_k^i)^{-1/s}$$

- **Theorem**[Optimality].

Thms 0,1,2 + Geometric Convergence \Rightarrow

Optimality

Theorem 1 (Upper bound for DOFs added by η_k)

Lemma (Optimal marking for η_k)

Total error reduction $\mathcal{T}_k \rightarrow \mathcal{T}_{k,*} \Rightarrow$ A Dörfler's property for $(\eta_k, \mathcal{M}_{k,*})$:

$$\eta_k(\mathcal{M}_{k,*})^2 \geq \theta \eta_k^2 \quad (0 < \theta < 1).$$

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- **Link** between nonlinear approximation theory and AFEM.

Theorem 1 (Upper bound for DOFs added by η_k)

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Total error reduction $\mathcal{T}_k \rightarrow \mathcal{T}_{k,*} \Rightarrow$ A Dörfler's property for $(\eta_k, \mathcal{M}_{k,*})$:

$$\eta_k(\mathcal{M}_{k,*})^2 \geq \theta \eta_k^2 \quad (0 < \theta < 1).$$

- **Link** between nonlinear approximation theory and AFEM.
 - ▶ If best approximation satisfies a total error reduction, then it also satisfies a Dörfler's property for $(\eta_k, \mathcal{M}_{k,*})$.
 - ▶ The solution of AFEM satisfies an **optimal** Dörfler's property for $(\eta_k, \mathcal{M}_{k,\text{est}})$.

$$\Rightarrow \quad \#\mathcal{M}_{k,\text{est}} \leq \#\mathcal{M}_{k,*}.$$

Theorem 1 - Optimal marking for η_k

Lemma (Optimal marking for η_k) Let $\mathcal{T}_{k,*} \supset \mathcal{T}_k$ be such that

$$\|u - u_{k,*}\|_{\Omega}^2 + \text{osc}(u_{k,*}, \mathcal{T}_{k,*})^2 \leq \lambda_{\text{est}} \left\{ \|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right\}$$

for $\lambda_{\text{est}} = \lambda_{\text{est}}(\theta_{\text{est}}) < \frac{1}{2}$, then

$$\eta_k(\overline{\mathcal{M}}_{k,*})^2 \geq \theta_{\text{est}} \eta_k^2.$$

Proof: Localized upper bound

Theorem 1 - Optimal marking for η_k

Lemma (Optimal marking for η_k) Let $\mathcal{T}_{k,*} \supset \mathcal{T}_k$ be such that

$$\|u - u_{k,*}\|_{\Omega}^2 + \text{osc}(u_{k,*}, \mathcal{T}_{k,*})^2 \leq \lambda_{\text{est}} \left\{ \|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right\}$$

for $\lambda_{\text{est}} = \lambda_{\text{est}}(\theta_{\text{est}}) < \frac{1}{2}$, then

$$\eta_k(\overline{\mathcal{M}}_{k,*})^2 \geq \theta_{\text{est}} \eta_k^2.$$

Proof: Localized upper bound

$$\Rightarrow \eta_k(\overline{\mathcal{M}}_{k,*})^2 \geq \frac{(1 - 2\lambda_{\text{est}})C_2}{1 + (1 + 2\text{osc}(\mathbf{D}, \mathcal{T}_0)^2)C_1} \eta_k^2 \geq \theta_{\text{est}} \eta_k^2$$

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$$\theta_{\text{est}} \in (0, \theta_{\text{est}}^*), \quad \theta_{\text{est}}^* := \frac{C_2}{1 + C_1(1 + 2\text{osc}(\mathbf{D}, \mathcal{T}_0)^2)}.$$

Theorem 1 - Proof I

$$\#\mathcal{T}_{k,\text{est}} - \#\mathcal{T}_k \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Proof:

Theorem 1 - Proof I

$$\#\mathcal{T}_{k,\text{est}} - \#\mathcal{T}_k \lesssim |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

Proof:

$$\epsilon^2 = \lambda_{\text{est}} \Lambda^{-1} \left(\|u - u_k\|_\Omega^2 + \text{osc}(u_k, \mathcal{T}_k) \right)^2$$

$$\begin{aligned} \mathcal{T}_\epsilon \supset \mathcal{T}_0 \quad : \quad & \|u - u_\epsilon\|_\Omega^2 + \text{osc}(u_\epsilon, \mathcal{T}_\epsilon)^2 < \epsilon^2 \\ & \#\mathcal{T}_\epsilon - \#\mathcal{T}_0 \lesssim \epsilon^{-1/s} |u, f, \mathbf{D}|_s^{1/s} \end{aligned}$$

$$\mathcal{T}_{k,*} = \mathcal{T}_\epsilon \cup \mathcal{T}_k$$

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$$\mathcal{T}_{k,*} = \mathcal{T}_\epsilon \cup \mathcal{T}_k \quad \Rightarrow \quad \#\mathcal{T}_{k,*} - \#\mathcal{T}_k \leq \#\mathcal{T}_\epsilon - \#\mathcal{T}_0$$

- $\mathcal{T}_k, \mathcal{T}_\epsilon$ are refinements of \mathcal{T}_0 .
- $\mathcal{T}_{k,*}$ is generated from \mathcal{T}_k adding **some elements** of \mathcal{T}_ϵ , anyway less than the number of triangles that must be added to go from \mathcal{T}_0 to \mathcal{T}_ϵ .

Theorem 1 - Proof II

- There is a **total error reduction** between $\mathcal{T}_{k,*}$ and \mathcal{T}_k .
- From **Lemma**(Optimal marking for η_k) we have

$$\#\mathcal{M}_{k,\text{est}} \leq \#\overline{\mathcal{M}}_{k,*} \leq C\mathcal{M}_{k,*}$$

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$$\begin{aligned} \Rightarrow \#\mathcal{M}_{k,\text{est}} &\preccurlyeq \#\mathcal{M}_{k,*} \\ &\preccurlyeq \#\mathcal{T}_{k,*} - \#\mathcal{T}_k \preccurlyeq \#\mathcal{T}_\epsilon - \#\mathcal{T}_0 \\ &\preccurlyeq \lambda_{\text{est}}^{-1/s} |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s} \end{aligned}$$

- λ_{est} gives the discrepancy between θ_{est} and θ_{est}^* .

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$$\Rightarrow \#\mathcal{M}_{k,\text{est}} \preccurlyeq \#\mathcal{M}_{k,*}$$

$$\preccurlyeq \#\mathcal{T}_{k,*} - \#\mathcal{T}_k \preccurlyeq \#\mathcal{T}_\epsilon - \#\mathcal{T}_0$$

$$\preccurlyeq \lambda_{\text{est}}^{-1/s} |u, f, \mathbf{D}|_s^{1/s} \left\{ \|u - u_k\|_\Omega + \text{osc}(u_k, \mathcal{T}_k) \right\}^{-1/s}$$

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Theorem 2 (Upper bound for DOFs added by osc)

Lemma (Optimal marking for $\text{osc}(u_k, \mathcal{T}_k)$)

“Oscillation reduction” \Rightarrow A Dörfler’s property for $(\text{osc}(u_k, \mathcal{T}_k), \mathcal{M}_{k,*})$

Theorem 2 (Upper bound for DOFs added by osc)

Lemma (Optimal marking for $\text{osc}(u_k, \mathcal{T}_k)$) Let \mathcal{T}_0 satisfy

$$\text{osc}(\mathbf{D}, \mathcal{T}_0) \leq \frac{\delta}{24C_1}.$$

If $v \in \mathbb{V}_{\mathcal{T}}$ and $\mathcal{T}_{k,*} \supset \mathcal{T} \supset \mathcal{T}_k$ are such that

$$\text{osc}(v, \mathcal{T}_{k,*}) \leq \lambda_{\text{osc}} \text{osc}(v, \mathcal{T}_k) \quad \|u - v\|_{\Omega} \leq C_1^{1/2} \eta_k \quad \delta^{1/2} \eta_k \leq \text{osc}(u_k, \mathcal{T}_k)$$

with $\lambda_{\text{osc}} > 0$ suitably small, then

$$\text{osc}(u_k, \mathcal{M}_{k,*})^2 \geq \theta_{\text{osc}} \text{osc}(u_k, \mathcal{T}_k)^2$$

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Proof: Use relation of oscillation between different meshes.

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Proof: Use relation of oscillation between different meshes.

$$\theta_{\text{osc}} \in (0, \theta_{\text{osc}}^*), \quad \theta_{\text{osc}}^* = \frac{1 - 24 C_1 \delta^{-1} \text{osc}(\mathbf{D}, \mathcal{T}_0)^2}{4}$$

Theorem 2 - Proof

$$\#\mathcal{T}_{k,\text{osc}}^i - \#\mathcal{T}_k^i \preccurlyeq |u, f, \mathbf{D}|_s^{1/s} \text{osc}(u_k, \mathcal{T}_k^i)^{-1/s}$$

Proof:

$$\epsilon_i = \lambda_{\text{osc}} \Lambda^{-1} \text{osc}(u_k, \mathcal{T}_k)$$

$$\begin{aligned} \mathcal{T}_{\epsilon_i} \supset \mathcal{T}_0 \quad : \quad & \|u - u_{\epsilon_i}\|_{\Omega}^2 + \text{osc}(u_{\epsilon_i}, \mathcal{T}_{\epsilon_i}) \leq \epsilon_i \\ & \#\mathcal{T}_{\epsilon_i} - \#\mathcal{T}_0 \preccurlyeq \epsilon_i^{-1/s} |u, f, \mathbf{D}|_s^{1/s} \end{aligned}$$

$$\mathcal{T}_{k,*}^i = \mathcal{T}_{\epsilon_i} \cup \mathcal{T}_k^i \quad \Rightarrow \quad \#\mathcal{T}_{k,*}^i - \#\mathcal{T}_k^i \leq \#\mathcal{T}_{\epsilon_i} - \#\mathcal{T}_0$$

- Take $u_{\epsilon_i} = v$ in **Lemma** (Optimal marking for $\text{osc}(u_k, \mathcal{T}_k)$), verify the hypothesis and conclude as in **Theorem 1**.

Optimality of AFEM - Proof I

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \preccurlyeq \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

Optimality of AFEM - Proof I

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \asymp \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

$$\sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \asymp \sum_{i=0}^{n_k-1} \text{osc}(u_k, \mathcal{T}_k^i)^{-1/s} |u, f, \mathbf{D}|_s^{1/s}$$

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$$\begin{aligned} \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i &\asymp \sum_{i=0}^{n_k-1} \text{osc}(u_k, \mathcal{T}_k^i)^{-1/s} |u, f, \mathbf{D}|_s^{1/s} \\ &\asymp |u, f, \mathbf{D}|_s^{1/s} \text{osc}(u_k, \mathcal{T}_k^{n_k-1})^{-1/s} \sum_{i=0}^{n_k-1} \rho^{i/s} \end{aligned}$$

$$\text{osc}(u_k, \mathcal{T}_k^{n_k-1}) \leq \rho^{n_k-1-i} \text{osc}(u_k, \mathcal{T}_k^i) \quad 0 \leq i \leq n_k - 1$$

Optimality of AFEM - Proof I

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \asymp \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

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$$\text{osc}(u_k, \mathcal{T}_k^{n_k-1}) \geq \delta^{1/2} \eta_k \geq \begin{cases} \frac{\delta^{1/2}}{C_1^{1/2}} \|u - u_k\|_{\Omega} \\ \delta^{1/2} \text{osc}(u_k, \mathcal{T}_k) \end{cases}$$

Optimality of AFEM - Proof I

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Optimality of AFEM - Proof II

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$$\#\mathcal{T}_n - \#\mathcal{T}_0 \asymp \sum_{k=0}^{n-1} \left(\|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right)^{-1/2s} |u, f, \mathbf{D}|_s^{1/s}$$

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$$\begin{aligned} \#\mathcal{T}_n - \#\mathcal{T}_0 &\asymp \sum_{k=0}^{n-1} \left(\|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right)^{-1/2s} |u, f, \mathbf{D}|_s^{1/s} \\ &\leq |u, f, \mathbf{D}|_s^{1/s} \left(\|u - u_n\|_{\Omega}^2 + \text{osc}(u_n, \mathcal{T}_n)^2 \right)^{-1/2s} \sum_{k=1}^n \alpha^{k/s} \end{aligned}$$

$$\|u - u_n\|_{\Omega}^2 + \text{osc}(u_n, \mathcal{T}_n)^2 \leq \alpha^{2(n-k)} \left(\|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right),$$

$$0 \leq k \leq n-1$$

Optimality of AFEM - Proof II

$$\#\mathcal{T}_n - \#\mathcal{T}_0 \asymp \sum_{k=0}^{n-1} \left(\#\mathcal{M}_{k,\text{est}} + \sum_{i=0}^{n_k-1} \#\mathcal{M}_{k,\text{osc}}^i \right)$$

$$\begin{aligned} \#\mathcal{T}_n - \#\mathcal{T}_0 &\asymp \sum_{k=0}^{n-1} \left(\|u - u_k\|_{\Omega}^2 + \text{osc}(u_k, \mathcal{T}_k)^2 \right)^{-1/2s} |u, f, \mathbf{D}|_s^{1/s} \\ &\leq |u, f, \mathbf{D}|_s^{1/s} \left(\|u - u_n\|_{\Omega}^2 + \text{osc}(u_n, \mathcal{T}_n)^2 \right)^{-1/2s} \sum_{k=1}^n \alpha^{k/s} \\ &\leq \left(\|u - u_n\|_{\Omega}^2 + \text{osc}(u_n, \mathcal{T}_n)^2 \right)^{-1/2s} |u, f, \mathbf{D}|_s^{1/s} \end{aligned}$$

Conclusions and Open problems

- We modify the MNS algorithm with a **new procedure for oscillation reduction relative to error estimator**, and prove its **convergence** and **optimality**.
- **Keys for Convergence**: orthogonality, discrete lower bound.
- **Keys for Optimality**: localized upper bound, optimal marking Lemma (error/oscillation reduction implies a Dörfler's property).
- Characterization of the **approximation class**

$$\mathcal{A}_s := \{(u, f, \mathbf{D}) : |u, f, \mathbf{D}|_s < \infty\}.$$

- Extension to the Laplace-Beltrami operator and AFEMs in $H(\text{div})$ and $H(\text{curl})$.
- Extension to a posteriori error estimators which **do not dominate** the oscillation (local problems).