

# Axioms of adaptivity

## Part I

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## Outline

- 1 Notation & Assumptions
- 2 Abstract framework
- 3 The axioms
- 4 Optimal convergence of the adaptive algorithm

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- 1 Notation & Assumptions
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- $\mathcal{X}$  is a vector space
- $\mathcal{T}_0$  is an initial shape-regular triangulation
- $\mathbb{T} := \{\mathcal{T} : \mathcal{T} \text{ is an admissible refinement of } \mathcal{T}_0\}$
- $\mathbb{T}(N) := \{\mathcal{T} \in \mathbb{T} : |\mathcal{T}| - |\mathcal{T}_0| \leq N\}$
- Each  $\mathcal{T} \in \mathbb{T}$  is split into  $C_{son} \geq 2$  elements in case of refinement
- Each  $\mathcal{T} \in \mathbb{T}$  induces a finite dimensional space  $\mathcal{X}(\mathcal{T})$
- There exists a solver

$$U(\cdot) : \mathbb{T} \rightarrow \mathcal{X}(\cdot)$$

which provides an approximation  $U(\mathcal{T}) \in \mathcal{X}(\mathcal{T})$  to a limit

$$u \in \mathcal{X}$$

- $T \in \mathcal{T}$  admits a refinement indicator

$$\eta_T(\mathcal{T}; \cdot) : \mathcal{X}(\mathcal{T}) \rightarrow [0, \infty)$$

- The summation of  $\eta_T$  over all  $T \in \mathcal{T}$  gives the global error estimator

$$\eta(\mathcal{T}; V)^2 := \sum_{T \in \mathcal{T}} \eta_T(\mathcal{T}; V)^2$$

for all  $V \in \mathcal{X}(\mathcal{T})$

- We assume the existence of an error measure  $d[\mathcal{T}; \cdot, \cdot]$  in  $\mathcal{X} \cup \mathcal{X}(\mathcal{T})$  which fulfills

- $d[\mathcal{T}; v, w] \geq 0,$
- $d[\mathcal{T}; v, w] \leq C_\Delta d[\mathcal{T}; w, v]$
- $C_\Delta^{-1} d[\mathcal{T}; v, y] \leq d[\mathcal{T}; v, w] + d[\mathcal{T}; w, y]$

for all  $v, w, y \in \mathcal{X} \cup \mathcal{X}(\mathcal{T}), C_\Delta > 0$

- For a refinement  $\hat{\mathcal{T}} \in \mathbb{T}$  of  $\mathcal{T} \in \mathbb{T}$ ,  $d[\hat{\mathcal{T}}; \cdot, \cdot]$  is well-defined on  $\mathcal{X} \cup \mathcal{X}(\mathcal{T}) \cup \mathcal{X}(\hat{\mathcal{T}})$  with

$$d[\hat{\mathcal{T}}; v, w] = d[\mathcal{T}; v, w]$$

for all  $v \in \mathcal{X}, w \in \mathcal{X}(\mathcal{T})$

- For each  $\varepsilon > 0$  there exists a refinement  $\hat{\mathcal{T}} \in \mathbb{T}$  of  $\mathcal{T} \in \mathbb{T}$  such that

$$d[\hat{\mathcal{T}}; u, U(\hat{\mathcal{T}})] \leq \varepsilon$$

- $\theta$  is a bulk parameter with

$$0 < \theta \leq 1$$

$\theta = 1$  : uniform refinement

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## Adaptive algorithm (Alg. 1)

INPUT:  $\mathcal{T}_0$  and  $\theta$

For  $\ell = 0, 1, 2, \dots$  do

- 1 Compute discrete approximation  $U(\mathcal{T}_\ell)$
- 2 Compute refinement indicators  $\eta_T(\mathcal{T}_\ell : U(\mathcal{T}_\ell))$  for all  $T \in \mathcal{T}_\ell$
- 3 Determine set  $\mathcal{M}_\ell \subseteq \mathcal{T}_\ell$  of (almost) minimal cardinality such that

$$\theta \eta(\mathcal{T}_\ell; U(\mathcal{T}_\ell))^2 \leq \sum_{T \in \mathcal{M}_\ell} \eta_T(\mathcal{T}_\ell; U(\mathcal{T}_\ell))^2$$

- 4 Refine (at least) the marked elements  $T \in \mathcal{M}_\ell$  to generate the triangulation  $\mathcal{T}_{\ell+1}$

OUTPUT: Discrete approximations  $U(\mathcal{T}_\ell)$  and error estimators  $\eta(\mathcal{T}_\ell : U(\mathcal{T}_\ell))$  for all  $\ell \in \mathbb{N}_0 := 0, 1, 2, 3, \dots$



## Conditions on the mesh

- The used mesh-refinement strategy has to satisfy

$$|\mathcal{T} \setminus \widehat{\mathcal{T}}| \leq |\widehat{\mathcal{T}}| - |\mathcal{T}|, \quad (1)$$

$$|\mathcal{T}_{\ell+1}| - |\mathcal{T}_\ell| \leq (C_{son} - 1) |\mathcal{T}_\ell| \text{ for all } \ell \in \mathbb{N}, \quad (2)$$

$$|\mathcal{T}_\ell| - |\mathcal{T}_0| \leq C_{mesh} \sum_{k=0}^{\ell-1} |\mathcal{M}_k| \text{ for all } \ell \in \mathbb{N} \quad (3)$$

for all refinements  $\widehat{\mathcal{T}} \in \mathbb{T}$  of  $\mathcal{T} \in \mathbb{T}$ , some  $\mathbb{T}$ -dependent constant  $C_{mesh} > 0$  and for any two meshes  $\mathcal{T}, \mathcal{T}' \in \mathbb{T}$  there is a coarsest common refinement  $\mathcal{T} \oplus \mathcal{T}' \in \mathbb{T}$  such that

$$|\mathcal{T} \oplus \mathcal{T}'| \leq |\mathcal{T}| + |\mathcal{T}'| - |\mathcal{T}_0| \quad (4)$$

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## Set of axioms

- (A1) Stability
- (A2) Reduction property
- (A3) General quasi-orthogonality
- (A4) Discrete reliability

## Stability (A1)

**Stability on non-refined element domains:** For all refinements  $\hat{\mathcal{T}} \in \mathbb{T}$  of a triangulation  $\mathcal{T} \in \mathbb{T}$ , for all subsets  $\mathcal{S} \subseteq \mathcal{T} \cap \hat{\mathcal{T}}$  of non-refined element domains and for all  $V \in \mathcal{X}(\mathcal{T})$ ,  $\hat{V} \in \mathcal{X}(\hat{\mathcal{T}})$ , it holds that

$$\left| \left( \sum_{T \in \mathcal{S}} \eta_T(\hat{\mathcal{T}}; \hat{V})^2 \right)^{1/2} - \left( \sum_{T \in \mathcal{S}} \eta_T(\mathcal{T}; V)^2 \right)^{1/2} \right| \leq C_{stab} d[\hat{\mathcal{T}}; \hat{V}, V] \quad (\text{A1})$$

with a constant  $C_{stab} > 0$ .

## Reduction property (A2)

**Reduction property on refined element domains:** Any refinement  $\hat{\mathcal{T}} \in \mathbb{T}$  of a triangulation  $\mathcal{T} \in \mathbb{T}$  satisfies

$$\sum_{T \in \hat{\mathcal{T}} \setminus \mathcal{T}} \eta_T(\hat{\mathcal{T}}; U(\hat{\mathcal{T}}))^2 \leq \rho_{red} \sum_{T \in \mathcal{T} \setminus \hat{\mathcal{T}}} \eta_T(\mathcal{T}; U(\mathcal{T}))^2 + C_{red} d[\hat{\mathcal{T}}; U(\hat{\mathcal{T}}), U(\mathcal{T})] \quad (\text{A2})$$

with  $C_{red} > 0$  and  $0 < \rho_{red} < 1$ .

## General quasi-orthogonality (A3)

**General quasi-orthogonality:** There exist constants

$$0 \leq \varepsilon_{qo} < \varepsilon_{qo}^*(\theta) := \sup_{\delta > 0} \frac{1 - (1 + \delta)(1 - (1 - \rho_{red})\theta)}{C_{rel}^2(C_{red} + (1 + \delta^{-1})C_{stab}^2)}$$

and  $C_{qo}(\varepsilon_{qo}) \geq 1$  such that the output of Alg. 1 satisfies, for all  $\ell, N \in \mathbb{N}_0$  with  $N \geq \ell$ , that

$$\begin{aligned} \sum_{k=\ell}^N (\mathfrak{d}[\mathcal{T}_{k+1}; U(\mathcal{T}_{k+1}), U(\mathcal{T}_k)]^2 - \varepsilon_{qo} \mathfrak{d}[\mathcal{T}_k; u, U(\mathcal{T}_k)]^2) \\ \leq C_{qo}(\varepsilon_{qo}) \eta(\mathcal{T}_\ell; U(\mathcal{T}_\ell))^2. \quad (\text{A3}) \end{aligned}$$

## Discrete reliability (A4)

**Discrete reliability:** For all refinements  $\hat{\mathcal{T}} \in \mathbb{T}$  of a triangulation  $\mathcal{T} \in \mathbb{T}$ , there exists a subset  $\mathcal{R}(\mathcal{T}, \hat{\mathcal{T}}) \subseteq \mathcal{T}$  and constants  $C_{ref}, C_{drel} > 0$  with  $\mathcal{T} \setminus \hat{\mathcal{T}} \subseteq \mathcal{R}(\mathcal{T}, \hat{\mathcal{T}})$  and  $|\mathcal{R}(\mathcal{T}, \hat{\mathcal{T}})| \leq C_{ref} |\mathcal{T} \setminus \hat{\mathcal{T}}|$  such that

$$d[\hat{\mathcal{T}}; U(\hat{\mathcal{T}}), U(\mathcal{T})]^2 \leq C_{drel}^2 \sum_{T \in \mathcal{R}(\mathcal{T}, \hat{\mathcal{T}})} \eta_T(\mathcal{T}; U(\mathcal{T}))^2. \quad (\text{A4})$$

## Discrete reliability implies reliability

### Lemma

*Discrete reliability (A4) implies reliability in the sense that any triangulation  $\mathcal{T} \in \mathbb{T}$  satisfies with  $C_{rel} = C_{\Delta} C_{drel}$*

$$d[\mathcal{T}; u, U(\mathcal{T})] \leq C_{rel} \eta(\mathcal{T}; U(\mathcal{T})). \quad (5)$$



## Quasi-monotonicity of the error estimator

### Lemma

*Stability (A1), reduction (A2), and discrete reliability (A4) imply quasi-monotonicity, i.e., there exists a constant  $C_{mon} > 0$  such that all refinements  $\hat{\mathcal{T}} \in \mathbb{T}$  of  $\mathcal{T} \in \mathbb{T}$  satisfy*

$$\eta(\hat{\mathcal{T}}; U(\hat{\mathcal{T}})) \leq C_{mon} \eta(\mathcal{T}; U(\mathcal{T})). \quad (6)$$

## Quasi-monotonicity of the error estimator

### Lemma

*Stability (A1), reduction (A2), reliability (5) and a Céa type best approximation, i.e., there is a constant  $C_{Céa} > 0$  such that*

$$d[\widehat{\mathcal{T}}; u, U(\widehat{\mathcal{T}})] \leq C_{Céa} \min_{V \in \mathcal{X}(\widehat{\mathcal{T}})} d[\widehat{\mathcal{T}}; u, V]. \quad (7)$$

*holds for any refinement  $\widehat{\mathcal{T}}$  of  $\mathcal{T} \in \mathbb{T}$ . Suppose that the ansatz spaces  $\mathcal{X}(\mathcal{T}) \subseteq \mathcal{X}(\widehat{\mathcal{T}})$  are nested. Then, the error estimator is quasi-monotone (6).*

## A concrete example

Consider the homogeneous Poisson problem in a bounded Lipschitz domain  $\Omega \subset \mathbb{R}^d$ ,  $d \geq 2$ , find  $u \in \mathcal{X} := H_0^1(\Omega)$  such that

$$a(u, v) := \int_{\Omega} \nabla u \cdot \nabla v \, dx = \int_{\Omega} f v \, dx =: \langle F, v \rangle \quad \forall v \in \mathcal{X}.$$

Suppose  $\mathcal{X}(\mathcal{T}) \subseteq \mathcal{X}(\widehat{\mathcal{T}}) \subseteq \mathcal{X}$  for all triangulations  $\mathcal{T} \in \mathbb{T}$  and all refinements  $\widehat{\mathcal{T}} \in \mathbb{T}$  of  $\mathcal{T}$  and suppose

$d[\mathcal{T}; v, w] = d[v, w] = \|v - w\|_{\mathcal{X}}$  for all  $\mathcal{T} \in \mathbb{T}$  and  $v, w \in \mathcal{X}$ .

The Lax-Milgram lemma yields solutions in the continuous as well as in the discrete cases. Furthermore, there holds a priori convergence

$$\lim_{\ell \rightarrow \infty} \|u - U(\mathcal{T}_{\ell})\|_{\mathcal{X}} = 0.$$

## A concrete example

Hence, stability (A1), reduction (A2) and reliability (5) already imply convergence.

Galerkin orthogonality yields

$$a(u - U(\mathcal{T}_{\ell+1}), V) = 0 \quad \forall V \in \mathcal{X}(\mathcal{T}_{\ell+1})$$

and implies the Pythagoras theorem

$$\|u - U(\mathcal{T}_{\ell+1})\|_{\mathcal{X}}^2 = \|u - U(\mathcal{T}_{\ell})\|_{\mathcal{X}}^2 - \|U(\mathcal{T}_{\ell+1}) - U(\mathcal{T}_{\ell})\|_{\mathcal{X}}^2.$$

General quasi-orthogonality (A3) is already satisfied.

It remains to show stability (A1), reduction (A2) and discrete reliability (A4).

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## References



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