

# Adaptive Virtual Element Method

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Joint work with:

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G. Vacca (Università di Bari)

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# Outline

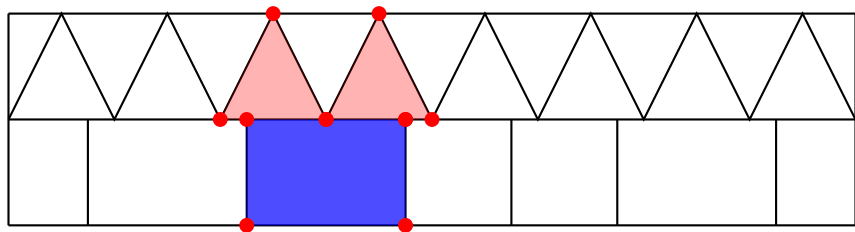
- 1 Introduction
- 2 Continuous problem and virtual discretization
- 3 Adaptive Virtual Element Method (AVEM)
- 4 Conclusions and perspectives

# Introduction

- Many different methods to solve PDEs on polytopal (i.e. polygonal/polyedral) meshes: Polytopal Finite Elements, Mixed/Hybrid Finite Volumes, Mimetic Finite Differences, Virtual Elements, Hybrid High-Order, Hybrid Discontinuous Galerkin, Polytopal Discontinuous Galerkin, Weak Galerkin, BEM-based polytopal FEM ...

# Why polygons/polyhedra?

## Gluing meshes



Conforming mesh: no hanging nodes

# Complex Geometries

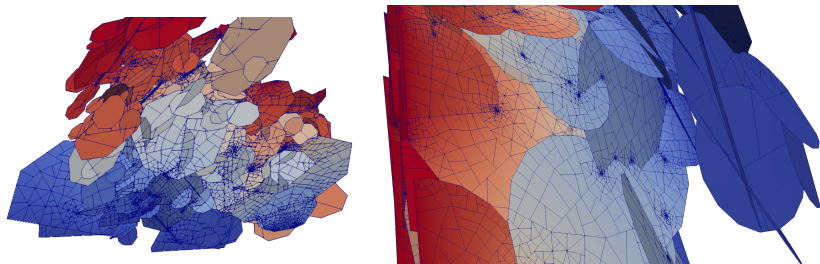


Figure: Polygonal mesh on a system of fractures (Courtesy of S. Berrone and A. D'Auria (Politecnico di Torino))

# Moving Geometries

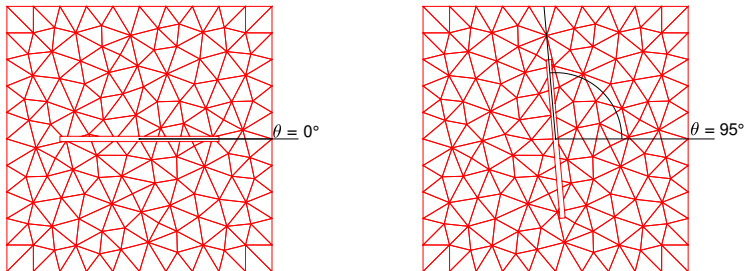
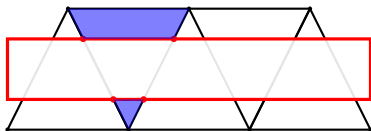


Figure: A rotating a bar on a **triangular background mesh** induces a **polygonal mesh** (from [Antonietti, Mascotto, V., Zonca, 2021])



# Adaptive Mesh refinement

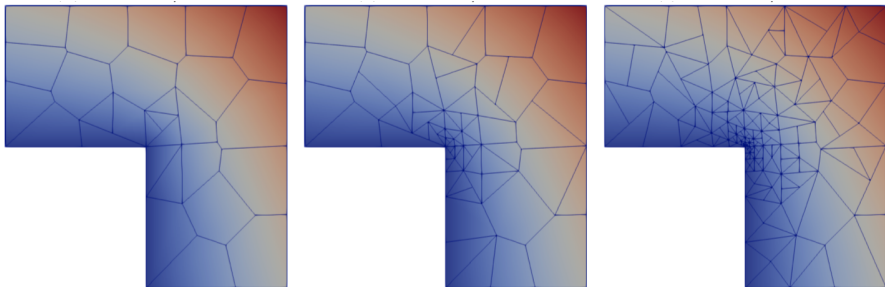
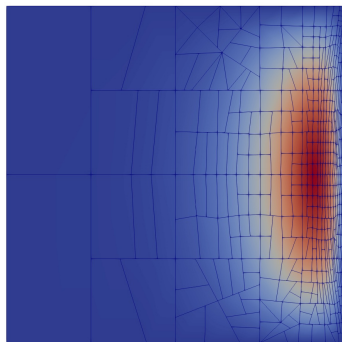
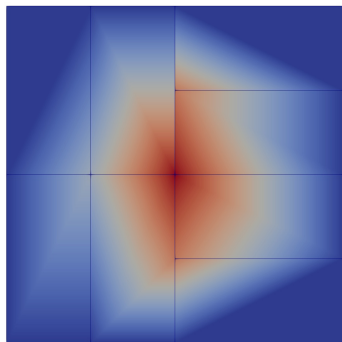


Figure: Polygonal refinement strategy based on preferential cutting direction  
(Courtesy of S. Berrone and A. D'Auria (Politecnico di Torino))



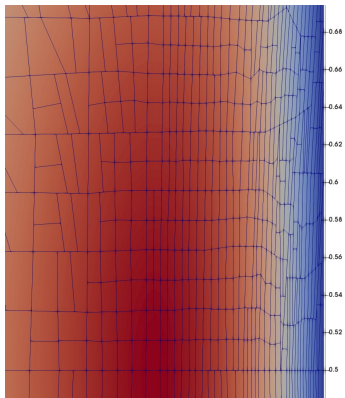
# Adaptive Mesh refinement



Anisotropic polygonal refinement

[Antonietti, Berrone, Borio, D'Auria, V., Weisser 2021]

# Adaptive Mesh refinement



Zoom of the refined mesh

# Agglomeration

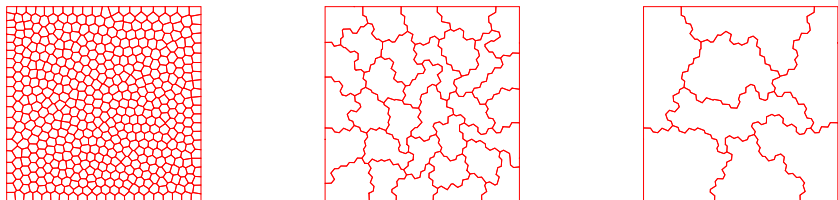


Figure: Agglomerated polygonal meshes with  $N_{el} = 512, 32, 8$

Agglomeration useful, e.g., in (adaptive) de-refinement mesh strategies

# Introduction

In this Talk we focus on the Virtual Element Method (VEM)

[Beirão da Veiga, Brezzi, Cangiani, Manzini, Marini, Russo, '13]

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[Beirão da Veiga, Brezzi, Cangiani, Manzini, Marini, Russo, '13]

**Idea of VEM:** Galerkin method where the **explicit** knowledge of the basis functions on polygons is not needed to assemble the algebraic problem (**only DOFS are needed**).

# Introduction

In this Talk we focus on the Virtual Element Method (VEM)

[Beirão da Veiga, Brezzi, Cangiani, Manzini, Marini, Russo, '13]

Intense research activity on VEM (very incomplete list ...):

- Methods:

Conforming and nonconforming approximation; mixed formulation; serendipity spaces; divergence-free elements; Trefftz methods; *hp*-approximation; a posteriori error estimates and adaptivity; curved faced/edges, divergence-free elements; preconditioners; ...

- Applications:

fluidynamic problems; structural mechanics problems; contact mechanics and elasto-plastic deformation problems; phase-field models of isotropic brittle fractures; cracks in materials; elastic wave propagation phenomena; underground flows and discrete fracture networks; propagation and scattering of time-harmonic waves; eigenvalue problems; Maxwell equation; Schrodinger equation; Laplace-Beltrami equation; Cahn-Hilliard equation; obstacle and minimal surface problems; topology optimization problems; nonlocal reaction-diffusion systems describing the cardiac electric field; ...

# Introduction

Flexibility of VEM allows:

- to deal with general polygonal/polyedral meshes;
- to easily incorporate additional features and regularity properties into the discrete space (divergence free,  $C^k$ , ...).

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In this Talk:

- Beirao da Veiga, Canuto, Nochetto, Vacca, V., Adaptive VEM: Stabilization-Free A Posteriori Error Analysis, arXiv:2111.07656, 2021
- Beirao da Veiga, Canuto, Nochetto, Vacca, V., Adaptive VEM: convergence analysis, in preparation.
- Beirao da Veiga, Canuto, Nochetto, Vacca, V., Adaptive VEM: optimality analysis, in preparation.

# Literature on VEM and adaptivity

## A posteriori error estimates and numerical tests of AVEM

Residual based  $h$ -estimator: [Berrone,Borio, 2017], [Cangiani, Georgoulis,Pryer,Sutton, 2017]

Residual based  $hp$ -estimator: [Beirao,Manzini,Mascotto, 2019]

Residual based anisotropic estimator: [Antonietti,Berrone,Borio,D'Auria,V., 2021]

Mixed-VEM: [Cangiani,Munar, 2019], [Munar, Sequeira, 2020]

Gradient recovery: [Chi,Beirao,Paulino, 2019]

Equilibrated flux: [Dassi,Gedicke,Mascotto, 2020], [Dassi,Gedicke,Mascotto, 2021]

## Polytopal meshes: quality and refinement

2d: [Beirao,Manzini, 2015], [Hoshina,Menezes,Pereira, 2018], [Berrone,Borio,D'Auria, 2021], [Berrone,D'Auria, 2021], [Attene et al., 2021],[Antonietti,Manuzzi, 2022], [Sorgente,Biasotti,Manzini, Spagnuolo, 2022], ...

3d: [D'Auria, PhD thesis, 2020], [Antonietti,Dassi,Manuzzi, 2022], ...

# Continuous problem and virtual discretization

## Continuous problem

$$-\nabla \cdot (A \nabla u) + cu = f \quad \text{in } \Omega \subset \mathbb{R}^2, \quad u = 0 \quad \text{on } \partial\Omega$$

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where

$A \in (L^\infty(\Omega))^{2 \times 2}$  is symmetric and uniformly positive-definite in  $\Omega$ ,  
 $c \in L^\infty(\Omega)$  is non-negative in  $\Omega$ ,  
 $f \in L^2(\Omega)$ .

## Continuous problem

$$-\nabla \cdot (A \nabla u) + cu = f \quad \text{in } \Omega \subset \mathbb{R}^2, \quad u = 0 \quad \text{on } \partial\Omega$$

The variational formulation is

$$u \in \mathbb{V} : \mathcal{B}(u, v) = (f, v)_\Omega \quad \forall v \in \mathbb{V} = H_0^1(\Omega)$$

with  $\mathcal{B}(u, v) := a(u, v) + m(u, v)$  where

$$a(u, v) := \int_{\Omega} (A \nabla u) \cdot \nabla v, \quad m(u, v) := \int_{\Omega} c u v$$

# Virtual Discretization

## Local Virtual Space (on polygon $E \in \mathcal{T}$ )

$$V_{\mathcal{T}}(E) = \{v_{\mathcal{T}} \in H^1(E) : \Delta v_{\mathcal{T}} = 0 \text{ in } E, v_{\mathcal{T}} \in \mathbb{P}^1(e) \forall e \in \partial E\}$$

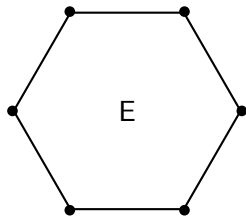
$v_{\mathcal{T}}$  **virtual** solution of Laplace problem with **prescribed** boundary datum



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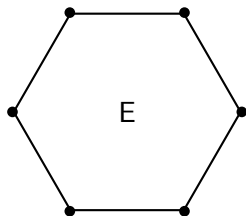
- $v_{\mathcal{T}} \in C^0(\partial E)$

**LOCAL DOFS:**  $v_{\mathcal{T}}$  at vertices

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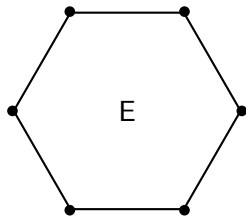
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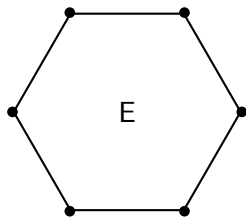
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- $\mathbb{P}^1(E) \subset V_{\mathcal{T}}(E)$

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LOCAL DOFS:  $v_{\mathcal{T}}$  at vertices

$v_{\mathcal{T}}$  **virtual** solution of Laplace problem with **prescribed** boundary datum

- $v_{\mathcal{T}} \in C^0(\partial E)$
- DOFS are unisolvent
- $\mathbb{P}^1(E) \subset V_{\mathcal{T}}(E)$
- On triangles:  $V_{\mathcal{T}}(E) = \mathbb{P}^1(E)$

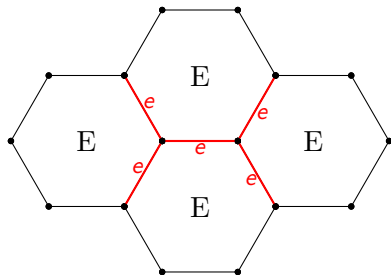
## Global Virtual Space

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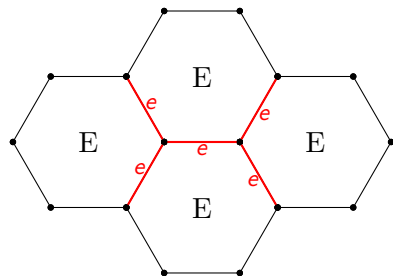


GLOBAL DOFS:  $v_T$  at vertices

# Virtual Discretization

## Global Virtual Space

$$V_{\mathcal{T}} = \{v_{\mathcal{T}} \in H_0^1(\Omega) : v_{\mathcal{T}}|_E \in V_{\mathcal{T}}(E) \forall E \in \mathcal{T}\}$$



GLOBAL DOFS:  $v_{\mathcal{T}}$  at vertices

- Local spaces  $V_{\mathcal{T}}(E)$  are  $C^0$ -glued:
  - $C^0$ -continuity at vertices (same point values of  $v_{\mathcal{T}}$ );
  - $C^0$ -continuity across edges  $e$  (same polynomial functions  $v_{\mathcal{T}}$ ).

# Virtual Discretization

## Weak formulation

$$u \in \mathbb{V} : \mathcal{B}(u, v) = (f, v)_\Omega \quad \forall v \in \mathbb{V} = H_0^1(\Omega)$$

with  $\mathcal{B}(u, v) := a(u, v) + m(u, v)$  where

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One would be tempted to write the virtual discrete problem as:

$$u_T \in V_T : \mathcal{B}(u_T, v_T) = (f, v_T)_\Omega \quad \forall v_T \in V_T$$

## BUT

this would require the explicit expression of the virtual functions in each polygon  $E$  that we do not want to employ. To set up the linear system we want to use only the DOFS.

## Virtual discretization

**Step 0:**  $\Pi_E^\nabla : V_T(E) \rightarrow \mathbb{P}_1(E)$  is the energy projector:

$$(\nabla(v - \Pi_E^\nabla v), \nabla w)_E = 0 \quad \forall w \in \mathbb{P}_1(E), \quad \int_{\partial E} (v - \Pi_E^\nabla v) = 0.$$

computable using DOFS only.

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**Step 1:** Define the local discrete bilinear form as

$$\mathcal{B}_T^E(u_T, v_T) := a^E(\Pi_E^\nabla u_T, \Pi_E^\nabla v_T) + m^E(\Pi_E^\nabla u_T, \Pi_E^\nabla v_T) + \gamma S^E(u_T, v_T)$$

where

- $\gamma > 0$  stabilization parameter
- $S^E(v_T, v_T) \simeq |v_T - \Pi_E^\nabla v_T|_{1,E}$  stabilizing form.

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Properties:

- **Consistency:**  $\mathcal{B}_T^E(q, v_T) = \mathcal{B}^E(q, v_T) \quad \forall q \in \mathbb{P}^1(E), \quad \forall v_T \in V_T(E)$
- **Stability:**  $\mathcal{B}_T^E(v_T, v_T) \simeq \mathcal{B}^E(v_T, v_T) \quad \forall v_T \in V_T(E)$

# Virtual Element discretization

**Step 2: Discrete problem.** Find  $u_T \in V_T$  such that

$$\mathcal{B}_T(u_T, v_T) = (f, v_T)_T \quad \forall v_T \in V_T$$

where

- $\mathcal{B}_T(u_T, v_T) = \sum_{E \in \mathcal{T}} \mathcal{B}_T^E(u_T, v_T)$
- $(f, v_T)_T = \sum_{E \in \mathcal{T}} \int_E f \Pi_E^\nabla v_T$

→ Optimal a priori error estimates in energy norm under suitable mesh assumptions

# Modified Local Space

## Local **Enhanced** Virtual Space (on polygon $E$ )

$$V_T(E) = \{v_T \in H^1(E) : \overbrace{\Delta v_T \in \mathbb{P}^1(E)}^{\text{ADD DOFS}}, v_T \in \mathbb{P}^1(e)\}$$

$$\underbrace{(\Pi_E^\nabla v_T, q)_{L^2(E)} = (v_T, q)_{L^2(E)} \quad \forall q \in \mathbb{P}^1(E)}_{\text{ADD CONSTRAINTS}}$$

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**DOFS** = vertex values

$L^2$ -projection on  $\mathbb{P}^1$  is computable using DOFS only .

## Typical mesh assumptions (for theory)

$E$  polygonal element of a partition  $\mathcal{T}$

- (a)  $E$  is a star-shaped polygon with respect to a circle of radius  $\rho$  and center  $z \in E$ .
- (b) The aspect ratio is uniformly bounded from above by  $\sigma$ , i.e.  $h_E/\rho < \sigma$ , being  $h_E$  the diameter of  $E$ .
- (c) For every edge  $e \subset \partial E$  it holds  $h_E \leq ch_e$ , being  $h_e$  the length of  $e$ .

Assumptions can be weakened (small edges):

[Beirão da Veiga, Lovadina, Russo, 2017], [Brenner, Sung, 2018]



# Adaptive Virtual Element Method (AVEM)

## Assumption (Coefficients and right-hand side of the equation)

*The coefficients  $A$  and  $c$  and the right-hand side  $f$  are constant in each element of the polygonal mesh  $\mathcal{T}$ .*

Study **convergence** and **optimality** properties of AVEM:

SOLVE → ESTIMATE → MARK → REFINE

Study **convergence** and **optimality** properties of AVEM:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

We follow the framework developed for AFEM (in particular, adaptive DGFEM).

Study **convergence** and **optimality** properties of AVEM:

SOLVE → ESTIMATE → MARK → REFINE

**Crucial Questions:**

Study **convergence** and **optimality** properties of AVEM:

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### Crucial Questions:

Q1: Is it possible to systematically refine general polytopes and preserve shape regularity?

Study **convergence** and **optimality** properties of AVEM:

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### Crucial Questions:

Q1: Is it possible to systematically refine general polytopes and preserve shape regularity?

*Shape regularity is critical to have robust interpolation estimates regardless of the resolution level.*

Study **convergence** and **optimality** properties of AVEM:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

### Crucial Questions:

Q1: Is it possible to systematically refine general polytopes and preserve shape regularity?

At the moment, there is no general positive answer.



Study **convergence** and **optimality** properties of AVEM:

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### Crucial Questions:

Q2: Is it possible to prove that **Error (+ Estimator)** reduces between consecutive adaptive iterations?

Study **convergence** and **optimality** properties of AVEM:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

### Crucial Questions:

Q2: Is it possible to prove that **Error (+ Estimator)** reduces between consecutive adaptive iterations?

*This is crucial to show that AVEM converges.*

Study **convergence** and **optimality** properties of AVEM:

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### Crucial Questions:

Q2: Is it possible to prove that **Error (+ Estimator)** reduces between consecutive adaptive iterations?

Comparing the stabilization terms under refinement is crucial (and problematic).

Study **convergence** and **optimality** properties of AVEM:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

### Crucial Questions:

Q3: Can we prove that the number of elements generated by REFINE is proportional to the number of elements selected by MARK?

Study **convergence** and **optimality** properties of AVEM:

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### Crucial Questions:

Q3: Can we prove that the number of elements generated by REFINE is proportional to the number of elements selected by MARK?

*This is crucial to show that AVEM leads to an error decay comparable with the best approximation in terms of degrees of freedom.*

Study **convergence** and **optimality** properties of AVEM:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

### Crucial Questions:

Q3: Can we prove that the number of elements generated by REFINE is proportional to the number of elements selected by MARK?

YES, if the refinement is **local**

**Drawback:** unlimited growth of nodes per element

Study **convergence** and **optimality** properties of AVEM:

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YES, if the refinement is **local**

**Drawback:** unlimited growth of nodes per element

$\rightsquigarrow$  **Restrict the number of hanging nodes per edge**

# Assumptions

In view of Q1, Q2 and Q3 we adopt the following framework:

- Polygonal mesh made of triangles with hanging nodes;
- Refinement based on "newest-vertex element bisection";
- Condition preventing unbounded number of hanging nodes per edge.

▶ bisection



# Assumptions

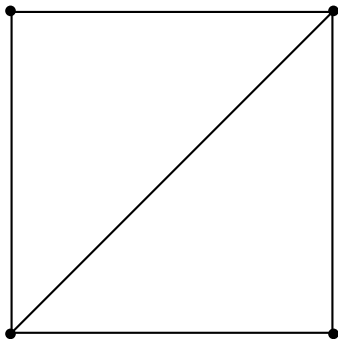
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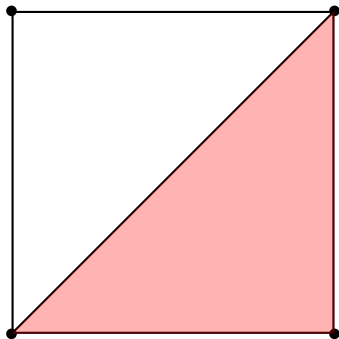
**Alternatively:** Polygonal mesh made of squares with hanging nodes (standard quad-tree refinement), or heterogeneous mesh made of triangles and squares with (bounded number of) hanging nodes .

▶ quad-tree

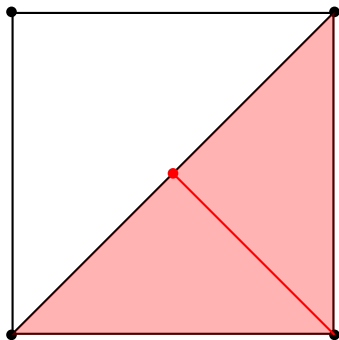
## Newest vertex bisection



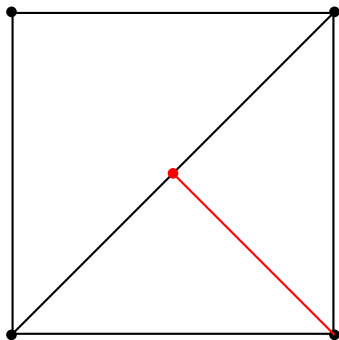
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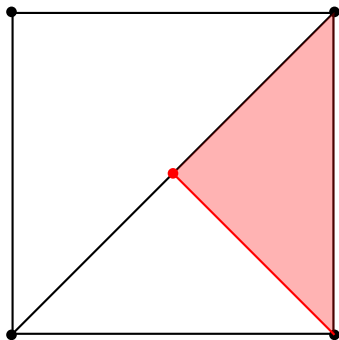
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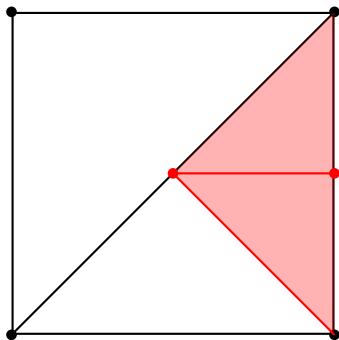
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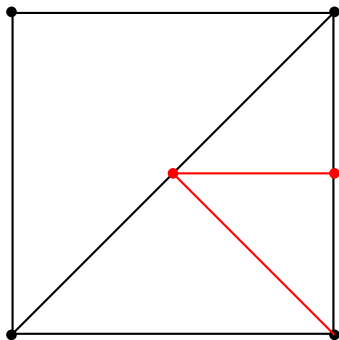
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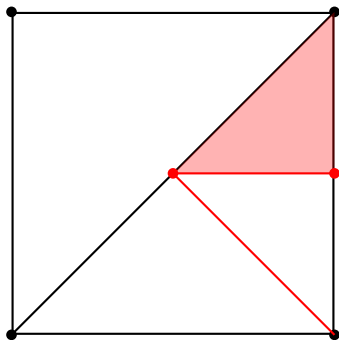


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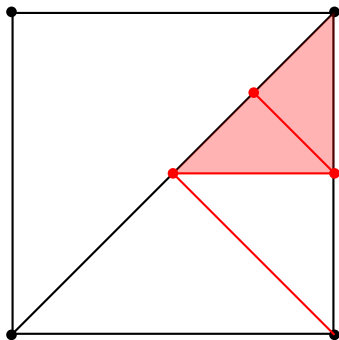




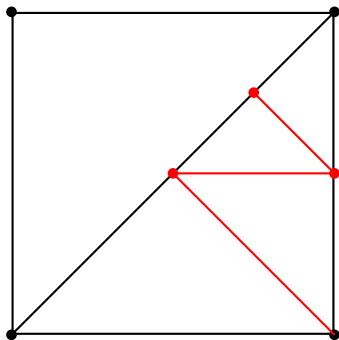
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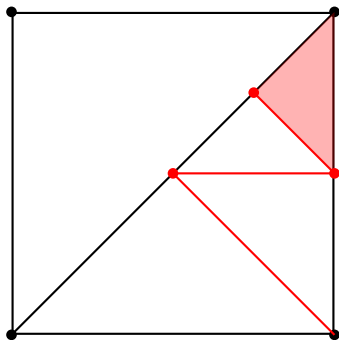
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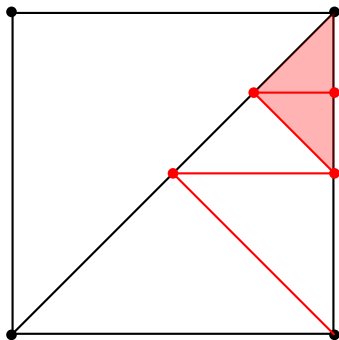
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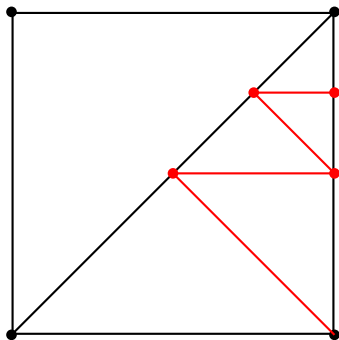
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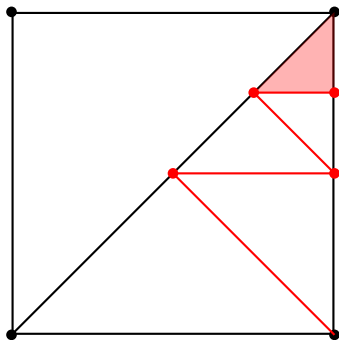
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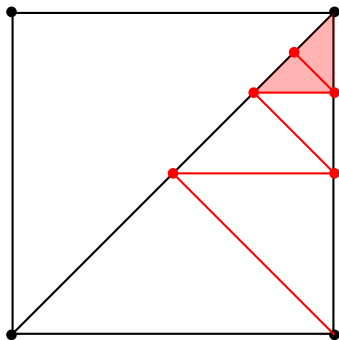
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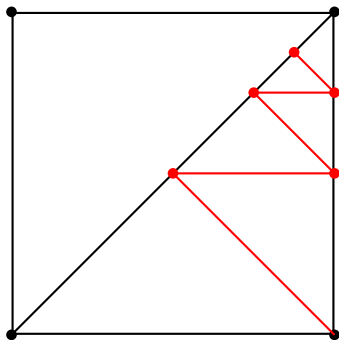


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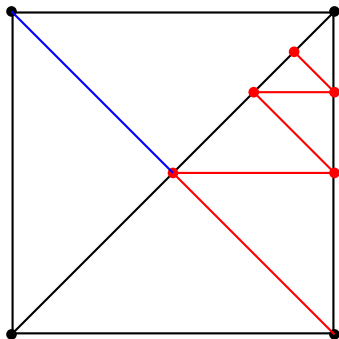




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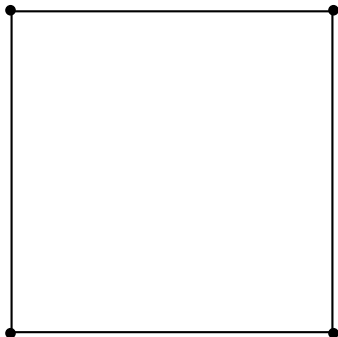
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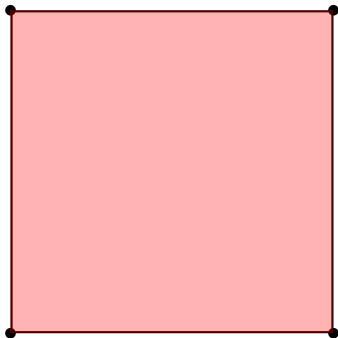
To restore the validity of the condition on the hanging nodes

▸ quad

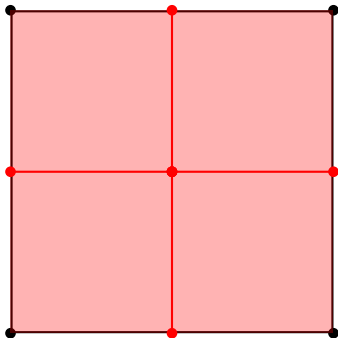
# Quad-tree refinement



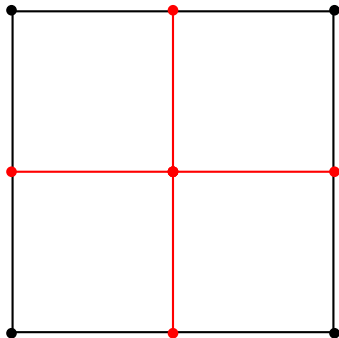
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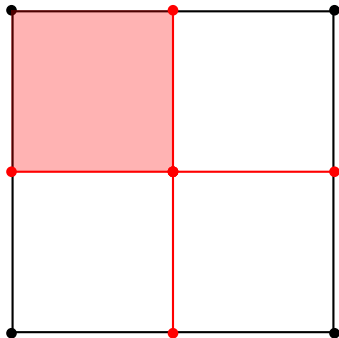
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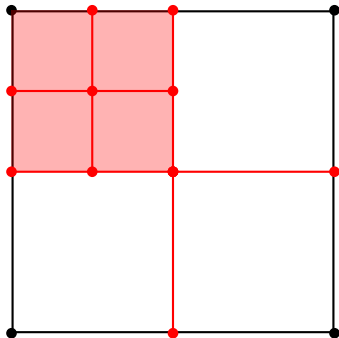
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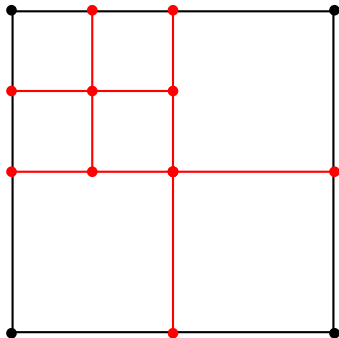


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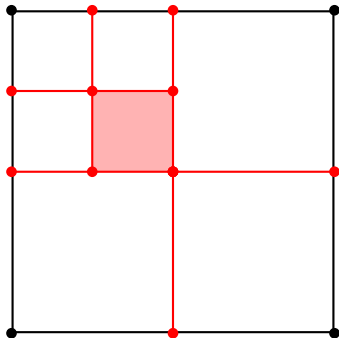




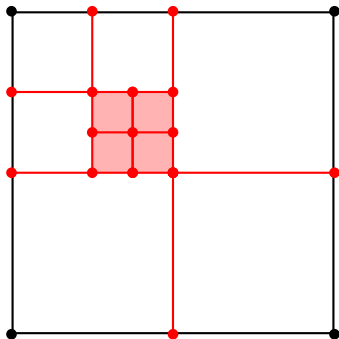
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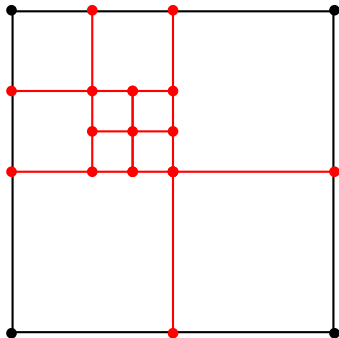
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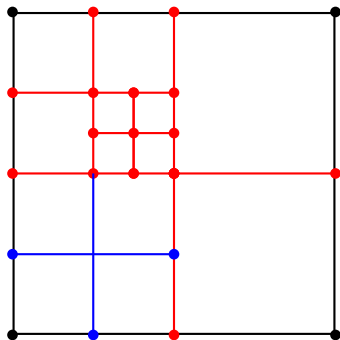
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To restore the validity of the condition on the hanging nodes

Recall the structure of **AVEM**:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

Recall the structure of **AVEM**:

SOLVE → **ESTIMATE** → MARK → REFINE

## a posteriori error estimate

Proposition ([Beirao, Canuto, Nochetto, Vacca, V. 2021])

$$|u - u_T|_{1,\Omega}^2 \leq C_{\text{apost}} (\eta_T^2(u_T, \mathcal{D}) + S_T(u_T, u_T))$$

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where

$$\begin{aligned} r_T(E; v, \mathcal{D}) &:= f_E - c_E \Pi_E^\nabla v, & j_T(e; v, \mathcal{D}) &:= \left[ [A_E \nabla \Pi_T^\nabla v] \right]_e \\ \eta_T^2(E; v, \mathcal{D}) &:= h_E^2 \|r_T(E; v, \mathcal{D})\|_{0,E}^2 + \frac{1}{2} \sum_{e \in \mathcal{E}_E} h_e \|j_T(e; v, \mathcal{D})\|_{0,e}^2, \\ \eta_T^2(v, \mathcal{D}) &:= \sum_{E \in \mathcal{T}} \eta_T^2(E; v, \mathcal{D}). \end{aligned}$$

See also [Cangiani, Georgoulis, Pryer, Sutton, 2017]

## Stabilization free a posteriori error estimate

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**Stabilization-free a posteriori error estimate:**

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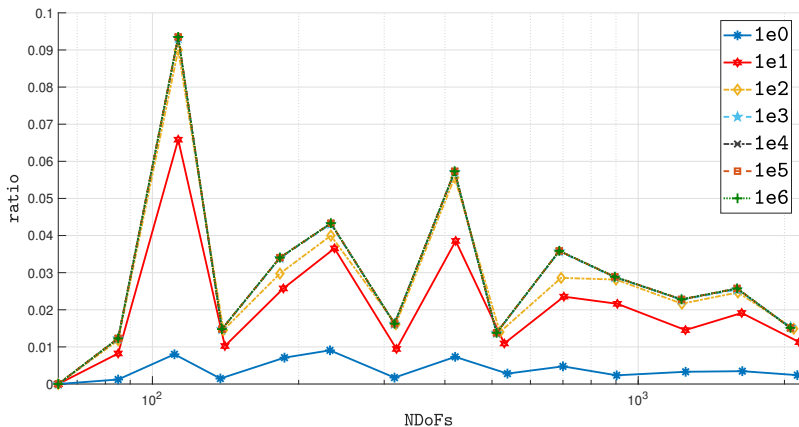
Stabilization-free a posteriori estimates opens the door to prove **convergence** of AVEM.

**Stabilization-free a posteriori error estimate:**

$$\left( C_{\text{apost}} - \frac{C_B}{\gamma^2} \right) \eta_{\mathcal{T}}^2(u_{\mathcal{T}}, \mathcal{D}) \leq |u - u_{\mathcal{T}}|_{1, \Omega}^2 \leq C_{\text{apost}} \left( 1 + \frac{C_B}{\gamma^2} \right) \eta_{\mathcal{T}}^2(u_{\mathcal{T}}, \mathcal{D})$$

Technically speaking: to obtain the result is essential to have access to a **subspace of  $V_{\mathcal{T}}$  made of continuous piecewise affine function on  $\mathcal{T}$** . This dictates our mesh assumptions!

- Recall the bound:  $\gamma^2 S_{\mathcal{T}}(u_{\mathcal{T}}, u_{\mathcal{T}}) \leq C_B \eta_{\mathcal{T}}^2(u_{\mathcal{T}}, \mathcal{D})$
- Employ AVEM with Dörfler parameter  $\theta = 0.5$  for L-shaped domain problem with  $A = I$ ,  $c = 0$ ,  $f = 1$  and vanishing boundary conditions.



$$\text{ratio} := \frac{\gamma^2 S_{\mathcal{T}}(u_{\mathcal{T}}, u_{\mathcal{T}})}{\eta_{\mathcal{T}}^2(u_{\mathcal{T}}, \mathcal{D})} \text{ for different values of } \gamma.$$



# AVEM in action

Poisson problem with piecewise constant  $a$ ,  $A = aI$ ,  $c = 0$  and  $f = 0$  ( $\rightarrow$  Kellogg's exact solution  $u \in H^{1+\epsilon}$ ,  $\epsilon < 0.1$ ).

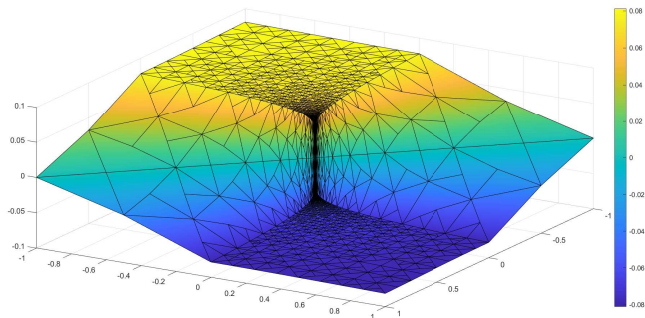
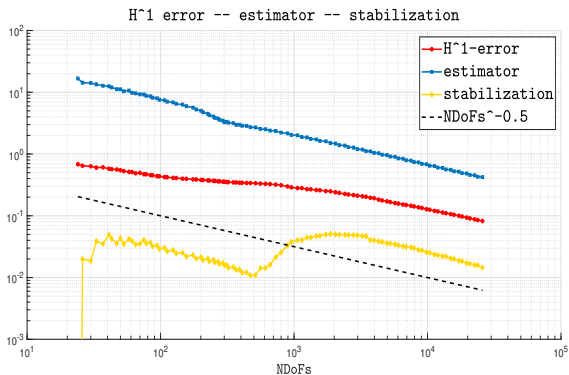
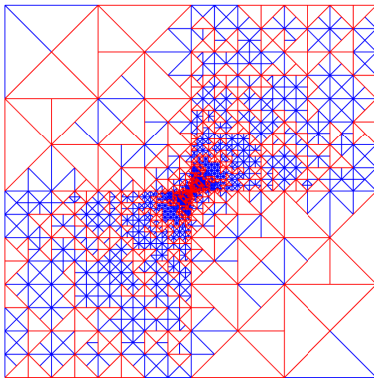


Figure: discrete solution

# AVEM in action

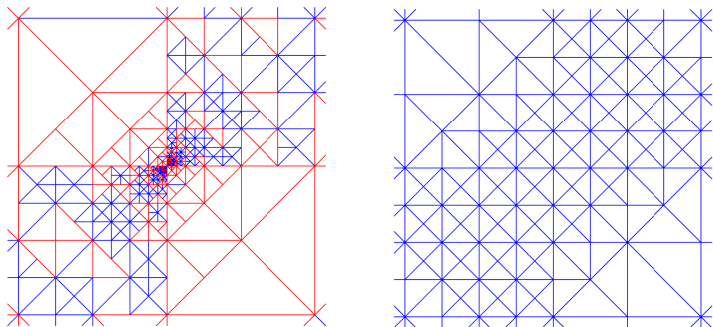


# AVEM in action



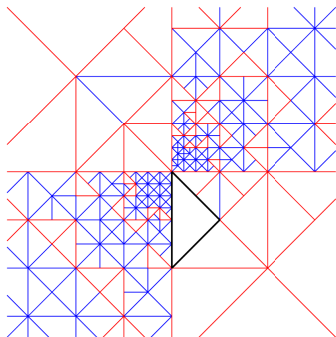
Final grid. Mesh elements having more than three vertices are drawn in red

# AVEM in action



Left: final grid  $\mathcal{T}_{VEM}$ . Right: final grid  $\mathcal{T}_{FEM}$ . Zoom to  $(-10^{-9}, 10^{-9})^2$ . VEM exhibits stronger grading at the singularity.

# AVEM in action



Final grid  $\mathcal{T}_{\text{VEM}}$ , zoom to  $(-10^{-10}, 10^{-10})^2$ . The black element is a **nonagon**

# AVEM with general data: the idea

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## Outer Loop

**Approximate data**  $(A, c, f)$  with piecewise constants up to tolerance  $\epsilon_k$

Update tolerance:  $\epsilon_k \rightarrow \epsilon_{k+1} < \epsilon_k$

Update Outer Loop counter :  $k \rightarrow k + 1$

# AVEM with general data: the idea

## Outer Loop

**Approximate data**  $(A, c, f)$  with piecewise constants up to tolerance  $\epsilon_k$

## Inner Loop

**Approximate the problem** with piecewise constant data

by iterating:

SOLVE  $\longrightarrow$  ESTIMATE  $\longrightarrow$  MARK  $\longrightarrow$  REFINE

up to tolerance  $\epsilon_k$

Update tolerance:  $\epsilon_k \rightarrow \epsilon_{k+1} < \epsilon_k$

Update Outer Loop counter :  $k \rightarrow k + 1$



## Convergence of Inner Loop

At each subiteration  $i$  we have :

$$\text{InnerError}(i)^2 + \beta \text{InnerEstimator}(i)^2 \lesssim \xi^i, \quad \xi < 1$$

$\text{InnerError}(i)$  = difference between **solution of the perturbed problem** and its **VEM approximation**

**Inner Loop:**

$\epsilon$ -approximation  
to  $\epsilon$ -perturbed  
problem.

+

**Outer Loop:**

reduces  $\epsilon$

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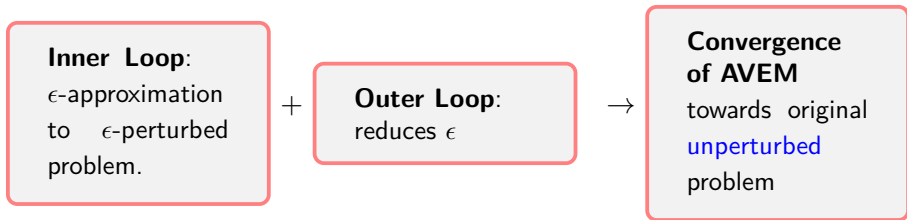
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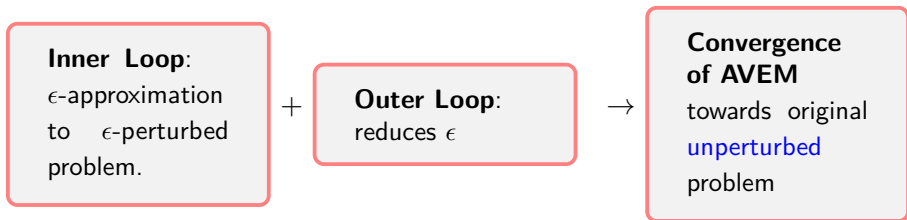
→

**Convergence  
of AVEM**

towards original  
**unperturbed**  
problem



Moreover, AVEM is quasi-optimal:



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AVEM produces an **approximation** of  $u$  with  $N$  dofs that is **comparable** with the **best  $N$ -term** VEM approximation of  $u$ .

# AVEM in action

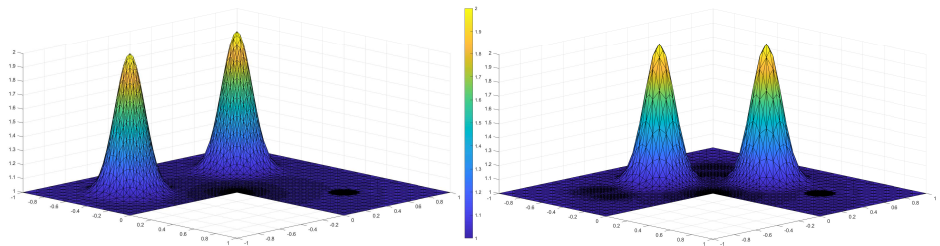
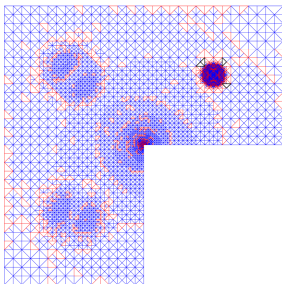
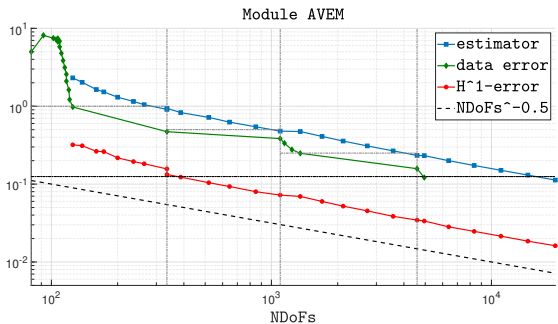


Figure: Left: graph of  $a$  ( $A = aI$ ). Right: graph of  $c$

Choose  $f$  so that

$$u_{\text{ex}}(x, y) = r^{\frac{2}{3}} \sin(2\alpha/3) + \exp(-1000((x - 0.5)^2 + (y - 0.5)^2))$$

# AVEM in action



## Conclusions:

- We discussed **convergence** and **optimality** properties of AVEM;
- We obtained theoretical results under **quite restrictive assumptions on the polygonal mesh (triangles/squares with hanging nodes)**;
- The analysis sheds light on **obstructions** to considering more general polygonal meshes.

## Perspectives:

- **Extending** convergence and optimality analysis of AVEM to **more general polygonal meshes**. This seems to require (at least):
  - 1 Refinement strategy preserving shape regularity;
  - 2 Stabilization-free a posteriori error estimates.
- **Extending** convergence and optimality analysis of AVEM to **higher order** virtual elements.



Thanks for your attention!