

**MATH 676**

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**Finite element methods in  
scientific computing**

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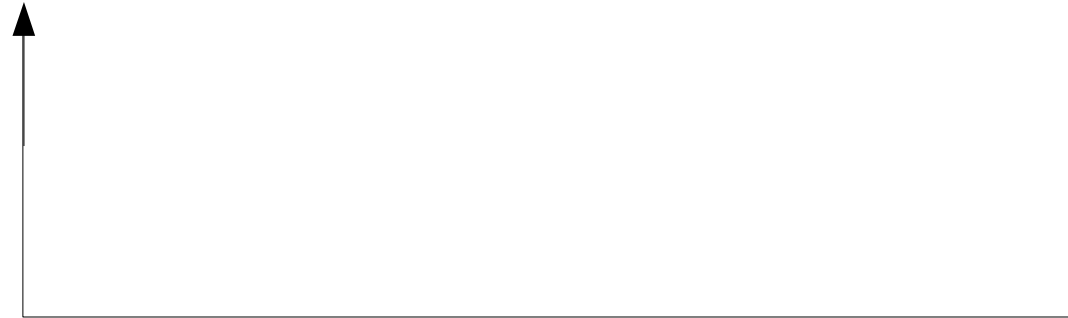
## **Lecture 17.75:**

# **Generating adaptively refined meshes: “A posteriori” error estimators**

# Adaptive mesh refinement (AMR)

## Adaptive mesh refinement happens in a loop:

SOLVE → ESTIMATE → MARK → REFINE



- SOLVE: Assemble linear system, solve it
- ESTIMATE: Compute a refinement indicator for each cell
- MARK: Determine which cells should be refined
- REFINE: Refine these cells (and possibly others)

# The ESTIMATE phase

**ESTIMATE phase:** Compute a “refinement indicator” for each cell  $K$ :

$$\eta = \{\eta_{K_1}, \eta_{K_2}, \eta_{K_3}, \dots, \eta_{K_N}\}$$

**Recall from lecture 17.25:**

A “reasonable” approach is to estimate the *interpolation error* of the solution, which leads to the Kelly indicator:

$$\eta_K = h_K^{1/2} \left( \int_{\partial K} |[\nabla u_h]|^2 \right)^{1/2}$$

# Ideas

**The “ideal” refinement indicator would be the actual error  $e_K$ :**

$$\eta_K := e_K = \|u - u_h\|_K = \left( \int_K |u - u_h|^2 dx \right)^{1/2}$$

**Questions/problems:**

1. Would a different norm be more appropriate?
2. But we don't have the exact solution  $u(x)$ !
3. Does refining a cell decrease the overall error?

# Ideas

**Question 3:** Does refining a cell decrease the overall error?

**Answers:**

- It depends on the norm in which we measure the error
- For *interpolation*, refining one cell only affects the error on this cell
- For the *Galerkin projection*, refining one cell affects the error *everywhere*

This is called “error pollution”.

# Ideas

**Question 3:** Does refining a cell decrease the overall error?

**Consider the Laplace equation:**

- Let  $T$  be a mesh,  $T^+$  be any refinement
- Let  $V_h$  be the FE space on  $T$ ,  $V_h^+$  be the FE space on  $T^+$

- Then  $V_h \subset V_h^+$

- Recall

$$E := \|\nabla(u - u_h)\| = \min_{v_h \in V_h} \|\nabla(u - v_h)\|$$

$$E^+ := \|\nabla(u - u_h^+)\| = \min_{v_h \in V_h^+} \|\nabla(u - v_h)\|$$

- Consequently:  $E^+ \leq E$

# Ideas

**Question 3:** Does refining a cell decrease the overall error?

**But:**

- It is not necessarily true that  $E^+ \leq E$  if  $E := \|u - u_h\|$
- The improvement may be marginal

To show that  $E^+ \leq \theta E$ ,  $\theta < 1$  one needs to refine *enough* cells.

**Strategy:** Dörfler marking / bulk refinement marks the cells with the largest errors so that

$$\sum_{K \in \text{marked cells}} \eta_K \geq \alpha \sum_{K \in T} \eta_K, \quad 0 < \alpha \leq 1$$



# A posteriori error estimation

**Question 2:** For evaluating the true error

$$\eta_K := e_K = \|u - u_h\|_K = \left( \int_K |u - u_h|^2 dx \right)^{1/2}$$

we would need to know the *exact* solution  $u(x)$ .

**Answer:** Yes. This is the central problem.

We need to find formulas that help us *estimate* the error only in term of  $u_h(x)$ .

This is called “a posteriori” error estimation.

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Consider

$$-\Delta u = f, \quad u|_{\partial\Omega} = 0.$$

Start with Galerkin orthogonality:

$$(\nabla u, \nabla v) = (f, v) \quad \forall v \in V$$

$$(\nabla u_h, \nabla v_h) = (f, v_h) \quad \forall v_h \in V_h \subset V$$

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$$\underbrace{(\nabla(u - u_h), \nabla v_h)}_{=: e} = 0 \quad \forall v_h \in V_h \subset V$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Then consider the energy norm of the error:

$$\begin{aligned}\|\nabla e\|_{\Omega}^2 &= (\nabla e, \nabla e)_{\Omega} = (\nabla e, \nabla(e - I_h e))_{\Omega} \\ &= \sum_K (\nabla e, \nabla(e - I_h e))_K \\ &= \sum_K (-\Delta e, e - I_h e)_K + (n \cdot \nabla e, e - I_h e)_{\partial K}\end{aligned}$$

Now use the following facts:

$$-\Delta e = -\Delta(u - u_h) = -\Delta u + \Delta u_h = \underbrace{f + \Delta u_h}_{\text{cell residual}}$$

$$(e - I_h e)|_{\partial\Omega} = ((u - u_h) - \underbrace{(I_h u - I_h u_h)}_{u_h})|_{\partial\Omega} = \underbrace{u|_{\partial\Omega}}_{=0} - \underbrace{I_h u|_{\partial\Omega}}_{=0}$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Thus:

$$\begin{aligned}\|\nabla e\|_{\Omega}^2 &= \sum_K (-\Delta e, e - I_h e)_K + (n \cdot \nabla e, e - I_h e)_{\partial K} \\ &= \sum_K (f + \Delta u_h, e - I_h e)_K + (n \cdot \nabla e, e - I_h e)_{\partial K \setminus \partial \Omega}\end{aligned}$$

The second term only integrates over interior faces. We visit each face twice (from  $K$  and from its neighbor  $K'$ ):

$$\begin{aligned}\sum_K (n_K \cdot \nabla e|_K, e - I_h e)_{\partial K \setminus \partial \Omega} &= \sum_K \frac{1}{2} (n_K \cdot \nabla e|_K, e - I_h e)_{\partial K \setminus \partial \Omega} \\ &\quad + \frac{1}{2} (\underbrace{n_{K'} \cdot \nabla e|_{K'}}_{=-n_K}, e - I_h e)_{\partial K \setminus \partial \Omega} \\ &= \sum_K \frac{1}{2} (n_K \cdot (\nabla e|_K - \nabla e|_{K'}), e - I_h e)_{\partial K \setminus \partial \Omega}\end{aligned}$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Consider the face term:

$$\sum_K (n_K \cdot \nabla e|_K, e - I_h e)_{\partial K \setminus \partial \Omega} = \sum_K \frac{1}{2} (n_K \cdot (\nabla e|_K - \nabla e|_{K'}), e - I_h e)_{\partial K \setminus \partial \Omega}$$

Observe:

$$\nabla e|_K - \nabla e|_{K'} = \underbrace{(\nabla u|_K - \nabla u|_{K'})}_{=0 \text{ because } u \text{ is smooth}} - \underbrace{(\nabla u_h|_K - \nabla u_h|_{K'})}_{=: [\nabla u_h]}$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Thus:

$$\|\nabla e\|_{\Omega}^2 = \sum_K \underbrace{(f + \Delta u_h, e - I_h e)_K}_{\text{cell residual}} - \frac{1}{2} \underbrace{(n \cdot [\nabla u_h], e - I_h e)_{\partial K \setminus \partial \Omega}}_{\text{"jump" residual}}$$

Using the Cauchy-Schwarz inequality on cell/face integrals:

$$\|\nabla e\|_{\Omega}^2 \leq \sum_K \left( \|f + \Delta u_h\|_K \|e - I_h e\|_K + \frac{1}{2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \|e - I_h e\|_{\partial K \setminus \partial \Omega} \right)$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Thus:

$$\begin{aligned}\|\nabla e\|_{\Omega}^2 &= \sum_K \underbrace{(f + \Delta u_h, e - I_h e)_K}_{\text{cell residual}} - \frac{1}{2} \underbrace{(n \cdot [\nabla u_h], e - I_h e)_{\partial K \setminus \partial \Omega}}_{\text{"jump" residual}} \\ &\leq \sum_K \left( \|f + \Delta u_h\|_K \|e - I_h e\|_K + \frac{1}{2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \|e - I_h e\|_{\partial K \setminus \partial \Omega} \right)\end{aligned}$$

Because  $e \in H^1$  we have the following basic interpolation estimates:

$$\|e - I_h e\|_K \leq Ch_K \|\nabla e\|_K$$

$$\|e - I_h e\|_{\partial K \setminus \partial \Omega} \leq Ch_K^{1/2} \|\nabla e\|_K$$

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Thus:

$$\begin{aligned} \|\nabla e\|_{\Omega}^2 &\leq \sum_K \left( \|f + \Delta u_h\|_K \underbrace{\|e - I_h e\|_K}_{\leq Ch_K \|\nabla e\|_K} + \frac{1}{2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \underbrace{\|e - I_h e\|_{\partial K \setminus \partial \Omega}}_{\leq Ch_K^{1/2} \|\nabla e\|_K} \right) \\ &\leq C \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right) \|\nabla e\|_K \end{aligned}$$

Recall Cauchy-Schwarz inequality:  $\sum_i x_i y_i \leq \left(\sum x_i^2\right)^{1/2} \left(\sum y_i^2\right)^{1/2}$

Apply this to the sum over cells:

$$\|\nabla e\|_{\Omega}^2 \leq C \left[ \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2 \right]^{1/2} \underbrace{\left[ \sum_K \|\nabla e\|_K^2 \right]^{1/2}}_{\|\nabla e\|_{\Omega}}$$



# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Thus:

$$\|\nabla e\|_{\Omega} \leq C \left[ \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2 \right]^{1/2}$$

**Interpretation:** We have bounded the error in terms of only the (computable!) discrete solution!

# A posteriori error estimation

## A posteriori error estimation for Laplace's equation:

Last reformulation, squaring both sides:

$$\|\nabla e\|_{\Omega}^2 \leq C \sum_K \underbrace{\left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)}_{=: \eta_K^2}$$

## Nomenclature:

- We call  $\eta_K$  the “residual-based” error estimator for cell  $K$ .
- It consists of the norm of the “cell residual” and the norm of the “jump residual”/“face residual”.

# A posteriori error estimation

**Note:** If we approximate

$$\begin{aligned} \|\nabla e\|_{\Omega}^2 &\leq C \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2 \\ &\approx C' \sum_K h_K \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega}^2 \end{aligned}$$

then we get exactly the “Kelly” estimator!

**Reason:** For odd polynomial degrees the jump residual dominates the cell residual.

**But:** For even polynomial degrees, it is the other way around!

# Some conclusions

**Conclusion 1:** It is possible to bound the (unknown) error exclusively in terms of the known, computed solution.

For example, for the Laplace equation we get:

$$\|\nabla e\|_{\Omega}^2 \leq C \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2$$

The estimate involves “cell” and “jump” residuals.

**Note:** Similar estimates can often be derived for other equation as well!

# Some conclusions

**Conclusion 2:** Estimates typically involve unknown constants.

For example, for the Laplace equation we get:

$$\|\nabla e\|_{\Omega}^2 \leq C \sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2$$

Here, the constant comes from the interpolation estimates. It is usually poorly known.

**Note:** For other equations, there are also stability constants, and constants from yet other estimates.

# Some conclusions

**Conclusion 3:** If we forget about the constant, we still get good refinement criteria!

For example, for the Laplace equation we get:

$$\|\nabla e\|_{\Omega}^2 \leq C \underbrace{\sum_K \left( h_K \|f + \Delta u_h\|_K + \frac{1}{2} h_K^{1/2} \|n \cdot [\nabla u_h]\|_{\partial K \setminus \partial \Omega} \right)^2}_{=: \eta_K^2}$$

Here, the  $\eta_K$  are entirely computable and yield good meshes, even though

$$\|\nabla e\|_{\Omega}^2 \not\propto \sum_K \eta_K^2$$

**Note:** The same is true for other equations.

# Some conclusions

**Conclusion 4:** That leaves the question whether we care about the  $H^1$  norm of the error,  $\|\nabla e\|_{\Omega}$ .

## Some answers:

- Sometimes we may. For the Laplace equation it has a meaning in terms of the “energy” of a physical system
- Oftentimes we don't.
- We may be much more interested in other, physical measures.
- This leads to “goal oriented” error estimates.

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