

# **Finite element methods in scientific computing**

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# **Lecture 3.95:**

## **The ideas behind the finite element method**

### **Part 6: Error estimates for the Laplace equation**

# Solutions

## Recall (ignoring boundary values):

The “weak solution”  $u \in H^1(\Omega)$  of the Laplace equation satisfies the “weak formulation”:

$$(\nabla \varphi, \nabla u) = (\varphi, f) \quad \forall \varphi \in H^1(\Omega)$$

The finite element solution  $u_h \in V_h \subset H^1(\Omega)$  satisfies the “discrete weak formulation”:

$$(\nabla \varphi_h, \nabla u_h) = (\varphi_h, f) \quad \forall \varphi_h \in V_h \subset H^1(\Omega)$$

**Here:**  $V_h$  is the space of piecewise polynomial functions of degree  $p$ , defined on the mesh.

# Previous questions

**At the end of Lecture 3.92, we had 3 questions:**

**Question 3:** Is the approximation  $u_h$  so defined “close” to the exact solution  $u$ ?

**Question 4:** Does  $u_h$  “converge” towards  $u$  in some useful sense?

**Question 5:** What is the computational effort to reach a certain accuracy? Optimality?

# Measuring errors

## How do we *measure* whether $u$ and $u_h$ are close?

- We use a “norm” of  $u-u_h$
- There are many possible norms:
  - Choice depends on the application
  - For some, estimating the error is easy, for others not.

## Examples:

- Maximal ( $L^\infty$ ) error:  $\|u-u_h\|_{L^\infty} = \max_{x \in \Omega} |u(x)-u_h(x)|$
- Mean square ( $L^2$ ) error:  $\|u-u_h\|_{L^2} = \left( \int_{\Omega} |u(x)-u_h(x)|^2 dx \right)^{1/2}$
- Gradient ( $H^1$ , energy) error:  $\|\nabla(u-u_h)\|_{L^2} = \left( \int_{\Omega} |\nabla(u(x)-u_h(x))|^2 dx \right)^{1/2}$

# Form of error estimates

## ***A priori* error estimates (available *before* computing):**

Representations of the (relative) error of the form

$$\frac{\|u - u_h\|_?}{\|u\|_?} \leq C(p) h^{\alpha(p)}$$

- Typically written as

$$\|u - u_h\|_? \leq C(p) h^{\alpha(p)} \|u\|_?,$$

- Guarantees convergence if  $\alpha(p) > 0$
- But absolute level of error unknown because  $u$  on the right hand side is unknown.

# Form of error estimates

## ***A posteriori* estimates (available *after* computing):**

Representations of the (relative) error of the form

$$\|u - u_h\|_? \leq C(p) h^{\beta(p)} Q(u_h)$$

- Does not guarantee convergence even if  $\beta(p) > 0$
- Can only be evaluated after computing  $u_h$
- Allows estimating actual size of the error
- Substantially more complicated to derive!  
(See Lecture 17.75)

# Ingredient #1: Galerkin orthogonality

## Starting point for most error estimation approaches:

- Discrete solution satisfies

$$(\nabla \varphi_h, \nabla u_h) = (\varphi_h, f) \quad \forall \varphi_h \in V_h \subset H^1(\Omega)$$

- Exact solution satisfies:

$$(\nabla \varphi, \nabla u) = (\varphi, f) \quad \forall \varphi \in H^1(\Omega)$$

- But because  $V_h \subset H^1(\Omega)$ , the exact solution also satisfies

$$(\nabla \varphi_h, \nabla u) = (\varphi_h, f) \quad \forall \varphi_h \in V_h \subset H^1(\Omega)$$

## Subtract the first from the third equation:

$$(\nabla \varphi_h, \nabla (u - u_h)) = 0 \quad \forall \varphi_h \in V_h \subset H^1(\Omega)$$



# Ingredient #1: Galerkin orthogonality

For finite element discretizations, we have:

$$(\nabla \varphi_h, \nabla (u - u_h)) = 0 \quad \forall \varphi_h \in V_h \subset H^1(\Omega)$$

This is called **Galerkin orthogonality**:

The error  $e = u - u_h$  is *perpendicular* to all elements  $\varphi_h \in V_h$  with regard to the scalar product  $(\nabla \circ, \nabla \circ)$  !

# Ingredient #2: Interpolation estimates

Recall (a variation of the theorem in) lecture 3.91:

$$\|\nabla (f - I_{h,p} f)\|_{L^2} \leq \frac{C(p, \Omega)}{p!} h^p \|\nabla^{p+1} f\|_{L^2}$$

**Here:**

- $f$  can be any function that has sufficiently many derivatives
- $I_{h,p} f$  is the function that *interpolates*  $f$ :
  - on a mesh with maximal cell diameter  $h$
  - has polynomial degree  $p$  on each cell

# Estimating the gradient error

**Start as follows:**

$$\begin{aligned}\|\nabla(u - u_h)\|_{L^2}^2 &= (\nabla(u - u_h), \nabla(u - u_h)) \\ &= (\nabla(u - u_h), \nabla(u - u_h)) + \underbrace{(\nabla\varphi_h, \nabla(u - u_h))}_{=0 \text{ (Galerkin orthogonality)}} \quad \forall \varphi_h \in V_h\end{aligned}$$

We can pick a “convenient” test function:  $\varphi_h = u_h - I_{h,p}u \in V_h$

$$= (\nabla(u - I_{h,p}u), \nabla(u - u_h))$$

Then apply the Cauchy-Schwarz inequality:

$$\leq \|\nabla(u - I_{h,p}u)\|_{L^2} \|\nabla(u - u_h)\|_{L^2}$$

Finally divide by the gradient norm of the error.

# Estimating the gradient error

**Situation now: We are comparing the error in the FE solution with the interpolation error:**

$$\underbrace{\|\nabla(u - u_h)\|_{L^2}}_{\text{Finite element error}} \leq \underbrace{\|\nabla(u - I_{h,p}u)\|_{L^2}}_{\text{Piecewise polynomial approximation error}}$$

We call this a “best approximation estimate”: the finite element error is at least as good as the interpolation error.

For other equations, error estimation is more difficult. “Best approximation” would then look like this:

$$\underbrace{\|\nabla(u - u_h)\|_{L^2}}_{\text{Finite element error}} \leq \underbrace{C_{\text{equation}}}_{\text{Specifics of the equation}} \underbrace{\|\nabla(u - I_{h,p}u)\|_{L^2}}_{\text{Piecewise polynomial approximation error}}$$

# Estimating the gradient error

**Situation now: We are comparing the error in the FE solution with the interpolation error:**

$$\underbrace{\|\nabla(u - u_h)\|_{L^2}}_{\text{Finite element error}} \leq \underbrace{\|\nabla(u - I_{h,p}u)\|_{L^2}}_{\text{Piecewise polynomial approximation error}}$$

Recall the interpolation error estimate ("Ingredient #2") to obtain the final result:

$$\|\nabla(u - u_h)\|_{L^2} \leq \underbrace{\frac{C(p, \Omega)}{p!}}_{\text{constant}} \underbrace{h^p}_{\text{convergence!}} \underbrace{\|\nabla^{p+1}u\|_{L^2}}_{\text{property of the solution}}$$

# Estimating the gradient error

For the Laplace equation, we have:

$$\|\nabla(u - u_h)\|_{L^2} \leq \underbrace{\frac{C(p, \Omega)}{p!}}_{\text{constant}} \underbrace{h^p}_{\text{convergence!}} \underbrace{\|\nabla^{p+1} u\|_{L^2}}_{\text{property of the solution}}$$

For many (but not all!) other equations, we get something like this:

$$\|\nabla(u - u_h)\|_{L^2} \leq \underbrace{C_{\text{equation}}}_{\text{specifics of the equations}} \underbrace{\frac{C(p, \Omega)}{p!}}_{\text{constant}} \underbrace{h^p}_{\text{optimal order convergence!}} \underbrace{\|\nabla^{p+1} u\|_{L^2}}_{\text{property of the solution}}$$

# Estimating the $L^2$ error

For the Laplace equation, we have:

$$\|\nabla(u - u_h)\|_{L^2} \leq \underbrace{\frac{C(p, \Omega)}{p!}}_{\text{constant}} \underbrace{h^p}_{\text{convergence!}} \underbrace{\|\nabla^{p+1} u\|_{L^2}}_{\text{property of the solution}}$$

With work ( $\rightarrow$  "Nitsche trick"), we can also estimate the  $L^2$  error:

$$\|u - u_h\|_{L^2} \leq \frac{C'(p, \Omega)}{p!} h^{p+1} \|\nabla^{p+1} u\|_{L^2}$$

# Estimating the $L^\infty$ error

For the Laplace equation, we have:

$$\|\nabla(u - u_h)\|_{L^2} \leq \underbrace{\frac{C(p, \Omega)}{p!}}_{\text{constant}} \underbrace{h^p}_{\text{convergence!}} \underbrace{\|\nabla^{p+1} u\|_{L^2}}_{\text{property of the solution}}$$

With substantially more work, we also get:

$$\|u - u_h\|_{L^\infty} \leq \frac{C''(p, \Omega)}{p!} h^{p+1} \|\nabla^{p+1} u\|_{L^2} \quad \text{if } p > 1$$

$$\|u - u_h\|_{L^\infty} \leq \frac{C''(p, \Omega)}{p!} h^{p+1} \left( \log \frac{1}{h} \right) \|\nabla^{p+1} u\|_{L^2} \quad \text{if } p = 1$$



# Summary

**For the Laplace equation, estimating errors is based on two fundamental properties:**

- Interpolation estimates
- Galerkin orthogonality

With these we get the following estimates that guarantee convergence and tell us how fast the error decreases with mesh refinement:

$$\|\nabla(u - u_h)\|_{L^2} \leq \frac{C(p, \Omega)}{p!} h^p \|\nabla^{p+1} u\|_{L^2}$$

$$\|u - u_h\|_{L^2} \leq \frac{C'(p, \Omega)}{p!} h^{p+1} \|\nabla^{p+1} u\|_{L^2}$$

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