

Adaptivity

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1. Here we examine strategies by which the grid / basis functions can be generated or modified to achieve a desired accuracy with as little computational time possible.

Two strategies:

→ **A priori** adaptivity: here the estimates employed do not require us to solve the problem (or at least we do so with a very coarse approximation); the grid is constructed before solving any system

→ **A posteriori** adaptivity: our estimates make use of our solution, so we must re-mesh and solve again (possibly multiple times).

An important (but difficult to achieve!) principle: **equidistribution of errors**.

→ in an ideal world, the contribution to the global error from each element should be roughly the same

⇒ typically $h_K^r |u|_{H^{r+1}(K)} \approx \text{constant}$ for all K .

(recall

$$\|u - u_h\|_{H^1(\Omega)} \leq \frac{M}{\alpha} C h^r |u|_{H^{r+1}(\Omega)}.)$$

→ to increase accuracy, we can refine the grid or increase the polynomial degree p (or both!)

→ here we consider grid refinement only, but note that doing both can give very effective results ('hp-refinement').

Both a priori and a posteriori refinement rely on an estimation of $|u|_{H^{r+1}(K)}$.

In fact, our estimates use

$$\|u - u_h\|_{H^1(\Omega)}^2 \leq \frac{M}{\alpha} C \sum_K h_K^{2r} |u|_{H^{r+1}(K)}^2$$

2. A priori adaptivity:

a) Here we will use a coarse approximation to u to estimate $|u|_{H^2(K)}$.



assume we compute u_h on some fairly coarse grid T_h .

Assume we are working with linear finite elements. Then we have a problem, because the approximation $u_h \notin H^2(K)$.

⇒ use reconstruction.

The idea is to find a continuous, piecewise linear approximation of ∇u_h

(at the moment, ∇u_h is discontinuous, but constant on elements).

i.e. take $(g_h)_j(x) = \sum_i g_{ij} \varphi_i(x)$

a vector, $g_h \approx \nabla u_h$. → sum over nodes.

Algorithm: we write $g_{ij} = \Pi_{ij}(x_i)$

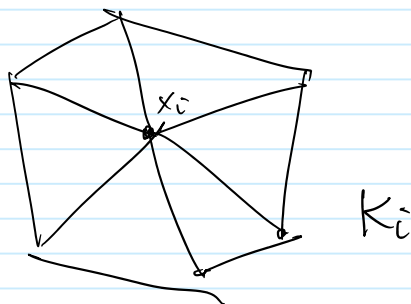
where $\Pi_{ij}(x) = \underline{A}x + \underline{b}$ the node x_i

(an affine map)

is defined by minimizing

$$\| \Pi_{ij} - \frac{\partial u_h}{\partial x_j} \|_{L^2(K_i)} \quad j=1, \dots, d.$$

where K_i is the set of all elements containing the node x_i , eg.



NB: the integrals $\int_{K_i} (A_{ij}x + b_{ij} - \frac{\partial u_h}{\partial x_j})^2$

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 can be computed analytically, giving a
 system to be solved; or quadrature
 results in an affine least-squares problem.

We then construct

$$H(u) \approx \frac{1}{2} (\nabla g_u + \nabla g_u^T)$$

↓ (to enforce symmetry of Hessian)

$$\text{Hessian } H(u)_{ij} = \frac{\partial^2 u}{\partial x_i \partial x_j}$$

b) refinement: we would like to have

$$\sum_K h_K^2 |u|_{H^2(K)}^2 \approx \sum_K h_K^2 \sum_{i,j} \|H_{ij}\|_{L^2}^2 \leq \frac{\varepsilon^2}{C^2}$$

to obtain error ε .

we can refine by inserting new vertices
 inside elements for which

$$\sum_K h_K^2 |u|_{H^2(K)}^2 > \frac{1}{N} \left(\frac{\varepsilon}{C}\right)^2$$

no. of elements.

c) Remeshing: here construct a spacing
 function

$$h \approx \frac{\varepsilon}{C \sqrt{\sum \|H_{ij}\|^2}}$$

(constant
 or each
 element)

$$|H|_K^2 = \frac{1}{C\sqrt{N}} \sqrt{\sum_{i,j} \|H_{ij}\|_{L^2(K)}^2} \quad \text{(or each element)}$$

then can apply, e.g. advancing front technique with this spacing factors etc.

3. A posteriori adaptivity: here the strategy considered is based on estimation of the residual operator

$$\begin{aligned} R : \quad \langle R, v \rangle &= F(v) - a(u_h, v) \\ &= a(u - u_h, v) \end{aligned}$$

Now, continuity/coercivity

$$\Rightarrow \alpha \|u - u_h\|_V \leq \|R\|_V \leq M \|u - u_h\|_V$$

(i.e. operator norm
 $\sup_{v \neq 0} \frac{\langle R, v \rangle}{\|v\|}$),

$$\text{Now take } \langle R, v \rangle = F(v - v_h) - a(u_h, v - v_h)$$

(as $a(u - u_h, v) = a(u - u_h, v - v_h)$ by Galilean invariance)

$$= a(u, v - v_h) - a(u_h, v - v_h)$$

$F(v-v_h)$ as u is solution.)

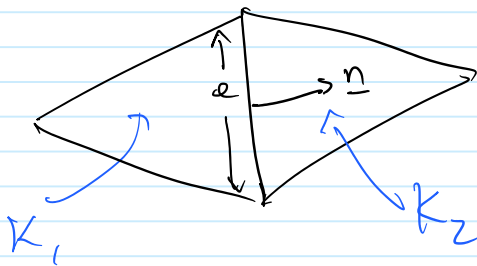
\Rightarrow (consider case of Poisson problem)

$$\langle R, v \rangle = \int_{\Omega} f(v-v_h) + \sum_K \int_K \nabla^2 u_h (v-v_h) - \sum_K \int_{\partial K} \frac{\partial u_h}{\partial n} (v-v_h)$$

$$= \sum_K \int_{\Omega} (f + \nabla^2 u_h)(v-v_h) - \sum_K \int_{\partial K} \frac{\partial u_h}{\partial n} (v-v_h)$$

To estimate $\int_{\partial K} \frac{\partial u_h}{\partial n} (v-v_h)$ define the jump of $\frac{\partial u_h}{\partial n}$ across an edge e

$$\left[\frac{\partial u_h}{\partial n} \right]_e = \left(\nabla u_h|_{K_1} - \nabla u_h|_{K_2} \right) \cdot \underline{n}$$



we can then define generalized jump by taking $\left[\frac{\partial u_h}{\partial n} \right]_e = 0$ if e is on boundary:
then

$$\sum_K \int_K \frac{\partial u_h}{\partial n} (v-v_h) = \sum_K \int_{\partial K} \frac{1}{2} \left[\frac{\partial u_h}{\partial n} \right]_e (v-v_h)$$

$$\sum_K \int_{\partial K} \frac{1}{\partial n} (v - v_h) = \sum_{\substack{K, \\ e \in K}} \int_e \left[\frac{1}{\partial n} \right]_e (v - v_h)$$

$\frac{1}{2}$ as each internal edge appears twice.

$$\Rightarrow \langle R_h v \rangle \leq \sum_K \left(\|f + \nabla^2 u_h\|_{L^2(K)} \|v - v_h\|_{L^2(K)} + \frac{1}{2} \left\| \left[\frac{\partial u_h}{\partial n} \right] \right\|_{L^2(\partial K)} \|v - v_h\|_{L^2(\partial K)} \right)$$

(Cauchy-Schwarz)

Interpolation: find a linear interpolant of v for estimation

→ Lagrange estimate has problems if v not continuous

→ can use, e.g. C1 element interpolation, obeying

$$\|v - R_h v\|_{L^2(K)} \leq C h_K |v|_{H^1(\bar{K})}$$

$$\|v - R_h v\|_{L^2(\partial K)} \leq C h_K^{1/2} \|v\|_{H^1(\bar{K})}$$

(here \bar{K} is union of all elements sharing an edge or vertex with K)

applying with Cauchy-Schwarz:

$$\langle R_h v \rangle^2 \leq C \left(\sum_K \rho_K (u_h)^2 \right) \left(\sum_K \|v\|_{H^1(\bar{K})}^2 \right)$$

$$\langle R, v \rangle \leq C \left(\sum_K \rho_K(u_h)^2 \right) \left(\sum_K \|v\|_{H^1(K)}^2 \right).$$

$$\rho_K(u_h) = h_K \|f + \nabla^2 u_h\|_{L^2(K)} + \frac{1}{2} h_K^{1/2} \left\| \left[\frac{\partial u_h}{\partial n} \right] \right\|_{L^2(K)}$$

$$\leq n \|v\|_{H^1(\Omega)}$$

max no. of
elements in \bar{K}

(bounded by regularity
assumption independent of h)

$$\text{hence } \|R\|_{V'}^2 \leq C_n \sum_K \rho_K(u_h)^2$$

and as $\alpha \|u - u_h\|_H \leq \|R\|_{V'}$, can use
as estimate.