

# Recovery-Based A Posteriori Error Estimators for Elliptic Equations

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# Scalar Elliptic Equations

- Scalar Elliptic Equations

$$-\nabla \cdot (A(x)\nabla u) + \mathbf{b} \cdot \nabla u + cu = f \quad \text{in } \Omega \subset \mathcal{R}^d$$

with boundary conditions

$$u = 0 \quad \text{on } \Gamma_D \quad \text{and} \quad \mathbf{n} \cdot A\nabla u = 0 \quad \text{on } \Gamma_N$$

- Let  $Xv = \mathbf{b} \cdot \nabla v + cv$ , rewrite the equation as

$$-\nabla \cdot (A(x)\nabla u) + Xu = f \quad \text{in } \Omega \subset \mathcal{R}^d$$

- Diffusion Dominant Case only

## $L^2$ Projection Recovery

- The finite element solution  $u_h \in \mathcal{U}_k$ ,  $\mathcal{U}_k$  ( $k \geq 1$ ) is the piecewise continuous  $k$ -th degree polynomial finite element space.

$$(A\nabla u_h, \nabla v) + (Xu_h, v) = (f, v) \quad \forall v \in \mathcal{U}_k$$

- Quantity to recover: the flux  $\sigma = -A\nabla u \in H(\text{div})$
- Recovered flux  $\sigma_h$  lies in  $H(\text{div})$  conforming finite element space  $RT_{k-1}$  or  $BDM_k$ .
- Find  $\sigma_h \in RT_{k-1}/BDM_k$ , s.t.

$$(A^{-1}\sigma_h, \tau) = -(\nabla u_h, \tau) \quad \forall \tau \in RT_{k-1}/BDM_k$$

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# $L^2$ Projection Error Estimators

- $L^2$  projection error estimator

$$\xi_{L^2} = \|A^{1/2} \nabla u_h + A^{-1/2} \sigma_h\|_{0,\Omega}$$

- When Linear Elements and Diffusion Dominated, even for discontinuous  $A$

$$\xi_{L^2} \sim \|A^{1/2} \nabla (u - u_h)\|_{0,\Omega}$$



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# Existing recovery based error estimators for higher order finite elements

- (Bank, Xu, and B. Zhang 2007)  
Superconvergence, gradient recovery
- (Naga and Z. Zhang 2005)  
Polynomial Preserving Recovery of the gradient on mildly structured mesh
- (Bartels and Carstensen 2002)  
Averaging Scheme for the gradient, Poisson equations

# Why $L^2$ recovery works for linear elements?

- Residual based error estimator:

$$\eta_{Res}^2 := \sum_{e \in \mathcal{E}} h_e \| [A \nabla u_h \cdot \mathbf{n}] \|_{0,e}^2 + \sum_{K \in \mathcal{T}} h_K^2 \| f + \nabla \cdot (A \nabla u_h) - \chi u_h \|_{0,K}^2$$

- (Carstensen and Verfürth 1999), for the linear element case, edge jump terms are dominant, and element residual terms are higher order terms.
- $\eta_{edge}^2 = \sum_{e \in \mathcal{E}} h_e \| [A \nabla u_h \cdot \mathbf{n}] \|_{0,e}^2$
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# Why $L^2$ recovery may fail for higher order elements?

- (D.Yu 91), for rectangular grids, edge jump terms are dominant for the odd-order element case, while element residual terms are dominant for the even-order element case.
- Simple  $L^2$  projection recovery of the flux may fail for higher order finite elements.

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# How to fix it?

Recover a  $\sigma$ , such that element residual terms are **higher order terms**.

# An H(div) Problem Recovery

- Find  $\sigma_h \in RT_{k-1}/BDM_k$ , s.t.,

$$\begin{aligned} (A^{-1}\sigma_h, \tau) + (\nabla \cdot \sigma_h, \nabla \cdot \tau) = \\ (-\nabla u_h, \tau) + (f - Xu_h, \nabla \cdot \tau) \quad \forall \tau \in RT_{k-1}/BDM_k \end{aligned}$$

- Too costly to solve?  
Fast Full-Multigrid H(div) Solvers or Direct Solvers

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# New Error Estimator

$$\xi_{hdiv,K} = \|A^{-1/2}\sigma_h + A^{1/2}\nabla u_h\|_{0,K}.$$

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# Analysis

- $e = u - u_h$ .
- Notation:  $\|h g(x)\|_0 = (\sum_{K \in \mathcal{T}} h_K^2 \|g\|_{0,K}^2)^{1/2}$
- Reliability bound.

$$\|A^{1/2} \nabla e\|_0 \leq C(\xi_{hdiv} + \|h(f - Xu_h - \nabla \cdot \sigma_h)\|_0)$$

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- Since  $(f - Xu_h - \nabla \cdot \sigma_h, \nabla \cdot \tau) = (\nabla u_h + A^{-1} \sigma_h, \tau)$ , we can prove

$$\|f - Xu_h - \nabla \cdot \sigma_h\|_0 \leq C \xi_{hdiv} + \|R - \mathcal{P}_{k-1} R\|_0$$

Where  $R = f - Xu_h - \nabla \cdot \sigma_h$ , and  $\mathcal{P}_{k-1}$  is the  $L^2$  projection operator onto the discontinuous piecewise polynomial space of degree  $k - 1$  with respect to the triangulation  $\mathcal{T}$ .

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# Two Criteria for Recovery Based Error Estimators

- Optimality of the mesh
  - Smooth solution and uniform mesh
    - Poisson equation  $-\Delta u = f$  and  $u \in H^{1+k}$ ,
    - $u_h \in \mathcal{U}_k$  is the finite element solution in the piecewise continuous  $k$ -th degree finite element space.
    - $\mathcal{T}_h$  is the mesh with uniform mesh size  $h$
    - $N$ : Number of the unknowns  $\approx h^{-d}$ ,  $d = 1, 2$ , or  $3$ .
    - $\|\nabla e\|_0 \leq Ch^k \|D^{1+k} u\|_0 = CN^{-k/d}$ .
    - The slope of  $\log(N)$ - $\log(\|\nabla e\|_0)$  line is  $-k/d$ .
  - Adaptive mesh generated by error indicators: should have similar error decay.
- Effectivity index  $\frac{\text{error estimator}}{\text{error}}$  is close to 1

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## A 1-D $P_2$ element example

$-u'' = f$  on  $(0, 1)$ ,  $u(0) = u(1) = 0$ , with the right-hand side function  $f = 30x^4 - 20x^3$  and the exact solution  $u = x^5(1 - x)$ .

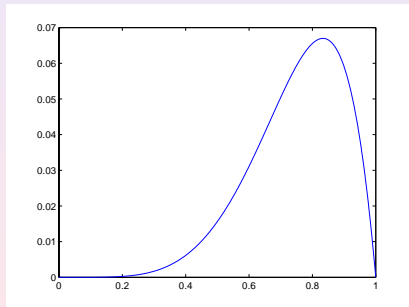


Figure: True solution  $u$

# $L^2$ projection error estimator fails

$u_h \in \mathcal{U}_2$  is the quadratic finite element solution.

$L^2$  projection recovery:

Find  $\sigma_h \in \mathcal{U}_2$ , s.t.,  $(\sigma_h, \tau) = -(u'_h, \tau) \forall \tau \in \mathcal{U}_2$

$\xi_{L2} = \|\sigma + u'_h\|_0$

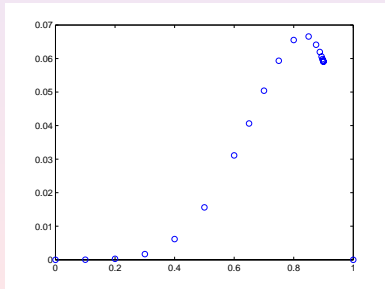


Figure: solution

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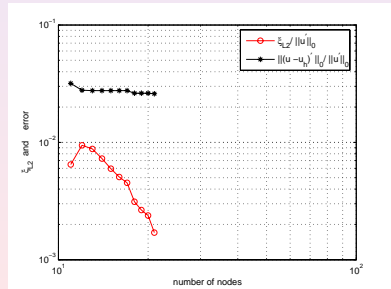


Figure: error and  $\xi_{L2}$

Recovery-Based A Posteriori Error Estimators

## Error estimator $\xi_{hdiv}$

$u_h \in \mathcal{U}_2$  is the quadratic finite element solution.

Recover  $\sigma_h \in \mathcal{U}_2$

$$(\sigma_h, \tau) + (\sigma'_h, \tau') = -(u'_h, \tau) + (f, \tau') \quad \forall \tau \in \mathcal{U}_2$$

Error Estimator

$$\xi_{hdiv} = \|\sigma_h + u'_h\|_0$$



For  $P_2$  element, error estimator  $\xi_{hdiv}$  works

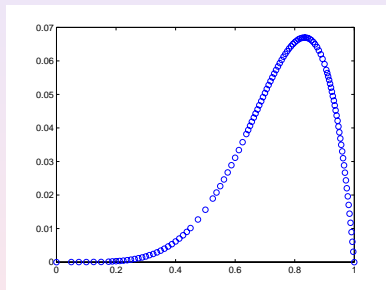


Figure: solution

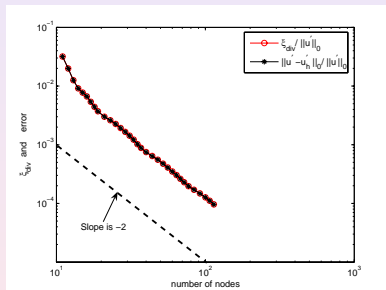


Figure: error and  $\xi_{hdiv}$

## 2-D $P_2$ Element Example

- Interface problem

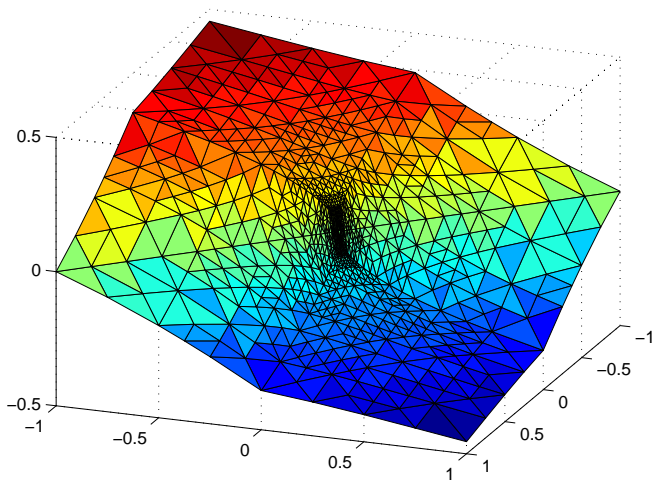
$$\begin{cases} -\nabla \cdot (a \nabla u) = f & \text{in } \Omega = (-1, 1)^2 \\ u = g & \text{on } \partial\Omega \end{cases}$$

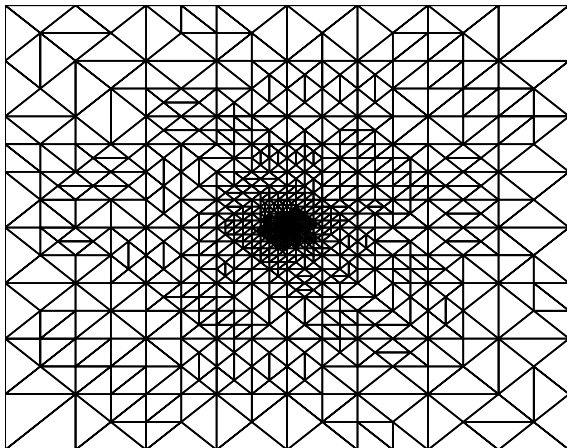
with  $a = R$  in  $(0, 1)^2 \cup (-1, 0)^2$  and 1 in  $(-1, 0) \times (0, 1) \cup (0, 1) \times (-1, 0)$

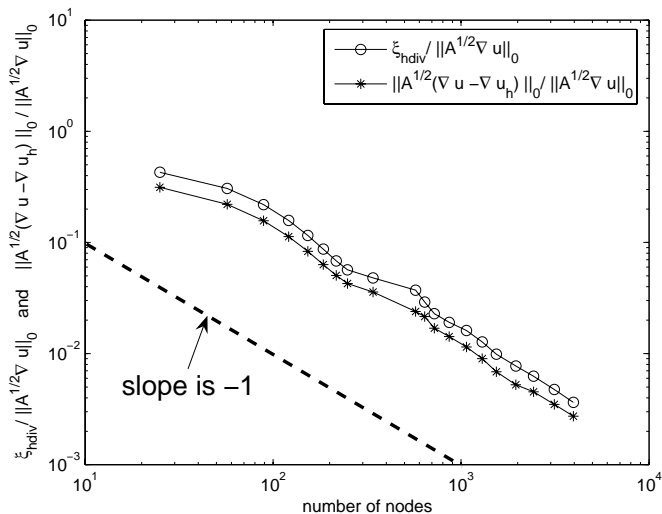
- Exact solution  $u(r, \theta) = r^\alpha \mu(\theta) \in H^{1+\alpha-\epsilon}(\Omega)$  with

$$\mu(\theta) = \begin{cases} \cos\left(\left(\frac{\pi}{2} - \sigma\right)\alpha\right) \cdot \cos\left(\left(\theta - \frac{\pi}{2} + \rho\right)\alpha\right) & \text{if } 0 \leq \theta \leq \frac{\pi}{2}, \\ \cos(\rho\alpha) \cdot \cos\left(\left(\theta - \pi + \sigma\right)\alpha\right) & \text{if } \frac{\pi}{2} \leq \theta \leq \pi, \\ \cos(\sigma\alpha) \cdot \cos\left(\left(\theta - \pi - \rho\right)\alpha\right) & \text{if } \pi \leq \theta \leq \frac{3\pi}{2}, \\ \cos\left(\left(\frac{\pi}{2} - \rho\right)\alpha\right) \cdot \cos\left(\left(\theta - \frac{3\pi}{2} - \sigma\right)\alpha\right) & \text{if } \frac{3\pi}{2} \leq \theta \leq 2\pi. \end{cases}$$

- Example when  $\alpha = 0.5$ , then  $R \approx 5.8284271247461907$ ,  $\rho = \pi/4$ , and  $\sigma \approx -2.3561944901923448$ .







## Concluding Remarks

- An Extension of  $L^2$  Recovery
- Flux Recovery for Higher Order Finite Elements
- No Regularity Assumptions are Required