

Error estimates for an average interpolation on anisotropic Q_1 elements

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Joint work with Ariel Lombardi

- The classic Hardy inequality and its dual inequality.
- Their application in error estimates for functions in weighted Sobolev spaces.
- Advantages over compactness arguments.
- Example of application in 1D. Graded meshes for singularly perturbed problems.
- Error estimates for narrow elements in 2D and 3D. Necessity of Average Interpolants.
- The generalized Hardy inequality in 2D and 3D.
- Estimates for average interpolants in anisotropic rectangular elements (in 2D and 3D).
- Applications.
- Numerical examples.

We are interested in estimates with weighted norms on the right hand side.

REASONS: to approximate singular functions or functions with large derivatives.

THE USE OF HARDY INEQUALITY

THE 1D CASE:

CLASSIC HARDY INEQUALITY:

$$\left\| \frac{v}{d} \right\|_{L^2(a,b)} \leq 2 \|v'\|_{L^2(a,b)}$$

$v \in H_0^1(a, b)$ $d(x)$ distance to the boundary of (a, b)

DUAL INEQUALITY:

$$u \in H^1(a, b) \quad , \quad \int_a^b u = 0$$

then,

$$\|u\|_{L^2(a,b)} \leq 2 \|du'\|_{L^2(a,b)}$$

Proof : Define $v \in H_0^1(a, b)$

$$v(x) = - \int_a^x u(y) dy$$

Using the Hardy inequality for v , we have

$$\begin{aligned} \|u\|_{L^2}^2 &= \int_a^b u'(x)v(x) dx \\ &\leq \left\| \frac{v}{d} \right\|_{L^2} \|du'\|_{L^2} \leq 2 \|u\|_{L^2} \|du'\|_{L^2} \end{aligned}$$

ERROR ESTIMATE FOR THE DERIVATIVE:

$I = (a, b)$ $u_I \in \mathcal{P}_1$ Lagrange interpolation

$u \in H^2(I)$ $0 \leq \alpha \leq 1$

$$\|(u - u_I)'\|_{L^2(I)} \leq 2|I|^{1-\alpha} \|d^\alpha u''\|_{L^2(I)}$$

Proof : Use that

$$\int_I (u - u_I)' = 0$$

and the DUAL HARDY INEQUALITY.

REMARKS: Estimates of this kind can be proved by compactness arguments. However, our method has the following advantages:

1- In the n-dimensional case we obtain explicit information on the dependence of the constants on the geometry of the elements. This is important in our analysis for anisotropic elements.

2- Our argument gives better results: One can not obtain the case $\alpha = 1$ by compactness. This case is of interest in some applications.

$$H^{1,d} = \{v \in L^2(I) : dv' \in L^2(I)\}$$

with norm

$$\|u\|_{H^{1,d}} = \|u\|_{L^2}^2 + \|du'\|_{L^2}^2$$

The inclusion $H^{1,d} \subset L^2$ IS NOT COMPACT!

EXAMPLE (Ariel Lombardi): $I = (0, 1)$

Consider the sequence

$$u_n(x) = \begin{cases} nx & 0 < x < \frac{1}{n} \\ 2 - nx & \frac{1}{n} \leq x < \frac{2}{n} \\ 0 & \frac{2}{n} \leq x < 1 \end{cases}$$

and $w_n = \sqrt{n}u_n$. Then, $\|w_n\|_{H^{1,d}}^2 = 10/3$. If $H^{1,d} \subset L^2$ is compact, there exists a subsequence w_n such that

$$w_n \rightarrow w \quad \text{in } L^2$$

but $w_n(x) \rightarrow 0 \quad \forall x \in I$ and so $w = 0$. But,

$$\|w_n\|_{L^2}^2 = 2/3$$

CONTRADICTION!

APPLICATIONS

GRADED MESHES: AN EXAMPLE IN 1D

CONVECTION-DIFFUSION EQUATION:

$$\begin{aligned} -\varepsilon u'' - b(x)u' + c(x)u &= f \quad \text{in } (0, 1) \\ u(0) = u(1) &= 0 \end{aligned}$$

$$b(x) \geq b_0 > 0 \quad \forall x \in (0, 1)$$

There is a boundary layer at $x = 0$.

GRADED MESH:

$$x_0 = 0 < x_1 < \cdots < x_N$$

u_I piecewise \mathcal{P}_1 Lagrange interpolation

Error estimate for the first interval $(0, x_1)$:

$$\begin{aligned}\varepsilon \|(u - u_I)'\|_{L^2(0, x_1)}^2 &\leq 4\varepsilon \|xu''\|_{L^2(0, x_1)}^2 \\ &\leq 4\varepsilon^{-2\beta} x_1^{2(1-\alpha)} \varepsilon^{1+2\beta} \|x^\alpha u''\|_{L^2(0, x_1)}^2\end{aligned}$$

REMARK: We will use this estimate for $\alpha < 1$, but it is important to have a constant independent of α .

Choose:

$$\beta = 1 - \alpha = \frac{1}{\log(\frac{1}{\varepsilon})}$$

So, $\varepsilon^{-\beta} = e$ and then,

$$\varepsilon \|(u - u_I)'\|_{L^2(0, x_1)}^2 \leq C x_1^{2(1-\alpha)} \varepsilon^{1+2\beta} \|x^\alpha u''\|_{L^2(0, x_1)}^2$$

Take $h > 0$ and $x_1 \leq h^{\frac{1}{1-\alpha}}$. Then,

$$\varepsilon \|(u - u_I)'\|_{L^2(0, x_1)}^2 \leq Ch^2 \varepsilon^{1+2\beta} \|x^\alpha u''\|_{L^2(0, x_1)}^2$$

Error estimate for the other intervals (x_j, x_{j+1}) :

$$\begin{aligned} \varepsilon \|(u - u_I)'\|_{L^2(x_j, x_{j+1})}^2 \\ \leq 4\varepsilon^{-2\beta} (x_{j+1} - x_j)^2 \varepsilon^{1+2\beta} \|u''\|_{L^2(x_j, x_{j+1})}^2 \end{aligned}$$

Now choose x_j such that:

$$x_{j+1} \leq x_j + hx_j^\alpha$$

Then,

$$\begin{aligned}\varepsilon \|(u - u_I)'\|_{L^2(x_j, x_{j+1})}^2 &\leq Ch^2 x_j^{2\alpha} \varepsilon^{1+2\beta} \|u''\|_{L^2(x_j, x_{j+1})}^2 \\ &\leq Ch^2 \varepsilon^{1+2\beta} \|x^\alpha u''\|_{L^2(x_j, x_{j+1})}^2\end{aligned}$$

WEIGHTED A PRIORI ESTIMATE:

$$\varepsilon^{1+2\beta} \|x^\alpha u''\|_{L^2}^2 \leq C$$

$$\text{if } \alpha \geq 0, \beta \geq 0, \alpha + \beta = 1$$

Consequently,

$$\varepsilon \|(u - u_I)'\|_{L^2(0,1)}^2 \leq Ch^2$$

with C independent of ε .

N : Number of nodes in graded mesh $\implies h \sim \frac{\log N}{N}$

Therefore,

$$\varepsilon \|(u - u_I)'\|_{L^2(0,1)}^2 \leq C \frac{\log N}{N}$$

Similar weighted estimates, but with different powers of $d(x)$, can be proved for the L^2 interpolation error.

L^2 - ERROR ESTIMATE:

$$\|u - u_I\|_{L^2(I)} \leq \frac{C}{1 - 2\alpha} |I|^{1-\alpha} \|d^\alpha u'\|_{L^2(I)}$$

for $0 \leq \alpha < \frac{1}{2}$.

The following example shows that the estimate is not true for $\alpha > \frac{1}{2}$:

$$u_n(x) = \begin{cases} nx & \text{if } 0 \leq x \leq \frac{1}{n} \\ 1 & \text{if } \frac{1}{n} < x \leq 1 \end{cases}$$

Then,

$$\|u_n - u_{n,I}\|_{L^2(0,1)} \rightarrow \left(\int_0^1 (1-x)^2 dx \right)^{\frac{1}{2}} = \frac{1}{\sqrt{3}}$$

while

$$\|x^\alpha u_n'\|_{L^2(0,1)}^2 = \int_0^{\frac{1}{n}} n^2 x^{2\alpha} dx = \frac{1}{2\alpha + 1} n^{1-2\alpha} \rightarrow 0$$

for $\alpha > \frac{1}{2}$

Using these estimates and the weighted a priori estimate

$$\varepsilon^{2\beta} \|x^\alpha u'\|_{L^2}^2 \leq C$$

$$\text{if } \alpha \geq 0, \beta \geq 0, \alpha + \beta = \frac{1}{2}$$

Choosing,

$$\beta = \frac{1}{2} - \alpha = \frac{1}{\log \frac{1}{\varepsilon}}$$

ERROR ESTIMATE IN ENERGY NORM

$$\|v\|_\varepsilon^2 = \|v\|_{L^2}^2 + \varepsilon^2 \|v'\|_{L^2}^2$$

$$\|u - u_I\|_\varepsilon \leq C \log \frac{1}{\varepsilon} \frac{\log N}{N}$$

THE 2D AND 3D CASES

Classic theory uses “regularity assumption”:

$$\frac{h_T}{\rho_T} \leq C$$

h_T exterior diameter, ρ_T interior diameter. For both Lagrange and Average Interpolants.

BUT: IT'S KNOWN THAT IT IS NOT NEEDED! First works: Babuska-Aziz, Jamet (1976).

Other references: Krizek, Al Shenk, Dobrowolski, Apel, Nicaise, Formaggia, Perotto, Acosta, D., etc..

FOR EXAMPLE: RECTANGULAR ELEMENTS

K reference element

Given $u \in H^2(K)$, let $p \in \mathcal{P}_1$ be such that

$$\left\| \frac{\partial}{\partial x}(u - p) \right\|_{L^2(K)} \leq C \left\| \nabla \frac{\partial u}{\partial x} \right\|_{L^2(K)}$$

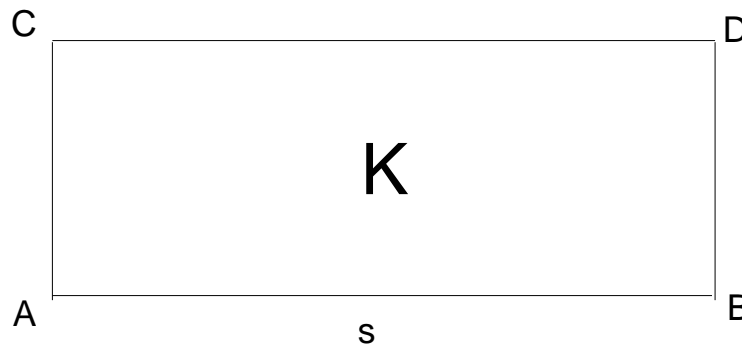
For example: p_1 the averaged Taylor polynomial of degree 1.

Let $u_I \in \mathcal{Q}_1$ be the Lagrange interpolation.

$$\left\| \frac{\partial}{\partial x}(u - u_I) \right\|_{L^2(K)} \leq \left\| \frac{\partial}{\partial x}(u - p) \right\|_{L^2(K)} + \left\| \frac{\partial}{\partial x}(p - u_I) \right\|_{L^2(K)}$$

So, it is enough to estimate $\left\| \frac{\partial}{\partial x}(p - u_I) \right\|_{L^2(K)}$

We use: for $v = p - u_I \in \mathcal{Q}_1(K)$



$$\left\| \frac{\partial v}{\partial x} \right\|_{L^2(K)}^2 \sim |v(B) - v(A)|^2 + |v(D) - v(C)|^2$$

$$\begin{aligned}
|v(B) - v(A)| &= |(p(B) - u(B)) - (p(A) - u(A))| \\
&= \left| \int_s \frac{\partial}{\partial x} (p - u) \right| \leq C \left\{ \left\| \frac{\partial}{\partial x} (p - u) \right\|_{L^2(K)} + \left\| \nabla \frac{\partial u}{\partial x} \right\|_{L^2(K)} \right\}
\end{aligned}$$

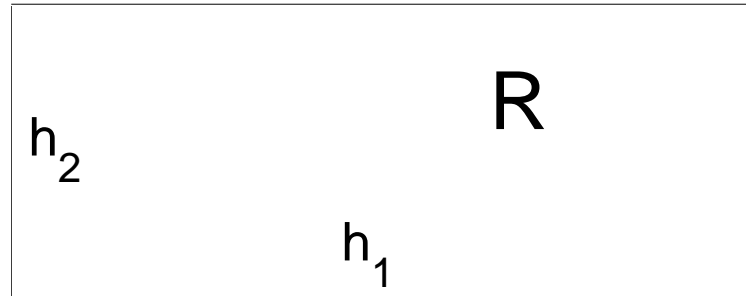
where we have used a trace theorem.

Analogously we bound $|v(D) - v(C)|$ and so we obtain:

$$\left\| \frac{\partial}{\partial x} (u - u_I) \right\|_{L^2(K)} \leq C \left\| \nabla \frac{\partial u}{\partial x} \right\|_{L^2(K)}$$

$$\frac{\partial^2 u}{\partial y^2} \quad \text{DOES NOT APPEAR!}$$

Therefore, changing variables we obtain for a rectangle R :



$$\left\| \frac{\partial}{\partial x}(u - u_I) \right\|_{L^2(R)} \leq C \left\{ h_1 \left\| \frac{\partial^2 u}{\partial x^2} \right\|_{L^2(R)} + h_2 \left\| \frac{\partial^2 u}{\partial x \partial y} \right\|_{L^2(R)} \right\}$$

THE CONSTANT C IS INDEPENDENT OF THE RELATION BETWEEN h_1 and h_2 !

A SIMILAR ESTIMATE IN 3D IS NOT TRUE !!

WHAT FAILS IN 3D IN THE ARGUMENT GIVEN ABOVE?

THE TRACE THEOREM:

$$\|u\|_{L^2(s)} \leq C \|u\|_{H^1(R)},$$

WHERE s IS AN EDGE OF R , IS NOT TRUE!

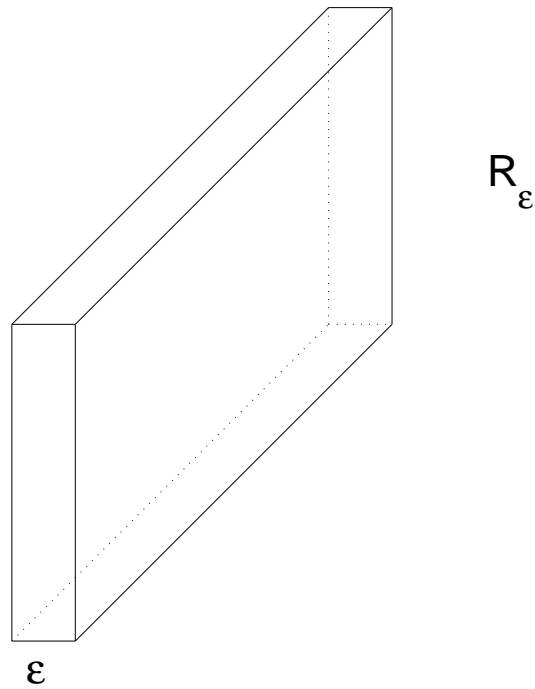
Counterexamples for the interpolation error estimate were given by:

Apel-Dobrowolski (Computing 1992), Al Shenk (Math. Comp. 1994).

They showed that the constant in the estimate

$$\|u - u_I\|_{H^1(R_\varepsilon)} \leq C_\varepsilon h \|u\|_{H^2(R_\varepsilon)}$$

goes to ∞ when $\varepsilon \rightarrow 0$



THIS IS ONE REASON TO WORK WITH AVERAGE INTERPOLANTS.

The other reason is the classic one: to approximate non-smooth functions.

GENERALIZED HARDY INEQUALITY:

$D \subset \mathbb{R}^n$ convex domain, $u \in H_0^1(D)$

$d(x)$ distance to the boundary

$$\left\| \frac{u}{d} \right\|_{L^2(D)} \leq 2 \|\nabla u\|_{L^2(D)}$$

ANISOTROPIC VERSION

$R = \prod_{i=1}^n (a_i, b_i)$ $h_i = b_i - a_i$

$u \in H_0^1(R)$, δ_R is a “normalized distance”:

$$\delta_R(x) = \min \left\{ \frac{x_i - a_i}{h_{R,i}}, \frac{b_i - x_i}{h_{R,i}} : 1 \leq i \leq n \right\}$$

$$\left\| \frac{u}{\delta} \right\|_{L^2(R)} \leq 2 \sum_{i=1}^n h_i \left\| \frac{\partial u}{\partial x_i} \right\|_{L^2(R)} .$$

DUAL INEQUALITY

$$\frac{1}{\delta} \leq h_i \leq \delta, \quad \psi \in C_0(R), \quad \int_R \psi = 1.$$

$$u \in H^1(R) \quad \text{such that} \quad \int_R u \psi = 0$$

$$\|u\|_{L^2(R)} \leq C \|d\nabla u\|_{L^2(R)}$$

with C depending only on δ and ψ .

Proof : REPEAT THE ARGUMENT GIVEN IN 1D:

$v := u - (\int_R u) \psi$ has vanishing mean value.

So, there exists $F \in H_0^1(R)^2$ such that

$$-\operatorname{div} F = v$$

and

$$\|F\|_{H_0^1(R)^2} \leq C \|v\|_{L^2(R)}$$

C DEPENDS ONLY ON δ : It follows from the explicit bound given in DM.

Since $\int_R u \psi = 0$, then

$$\|u\|_{L^2(R)}^2 = \int_R uv = - \int_R u \operatorname{div} F$$

and the proof finish as in the 1D case.

AN AVERAGE INTERPOLANT

Our construction is a modification of that in D. (Math. Comp. 1999).

DIFFERENCE: We do not use reference elements for the definition!

In this way we can relax the regularity assumptions on the mesh.

ASSUMPTION: local regularity in each direction

R, S neighboring elements

$$\frac{h_{R,i}}{h_{S,i}} \leq \sigma \quad 1 \leq i \leq n.$$

OUR ERROR ESTIMATES DEPEND ONLY ON σ .

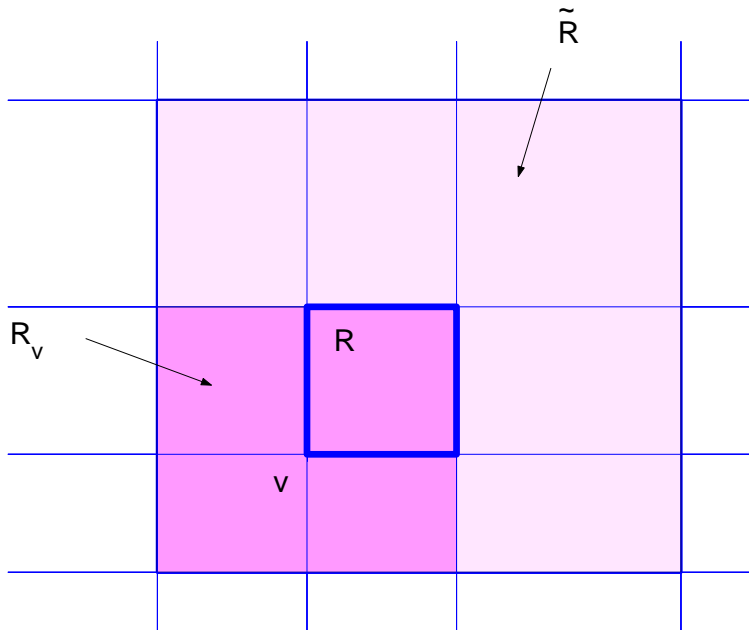
\mathcal{N}_{in} set of interior nodes. For $z \in \mathcal{N}_{in}$:

$$h_{z,i} = \min\{h_{R,i} : z \text{ is a vertex of } R\}, \quad 1 \leq i \leq n$$

$$\tilde{R} = \bigcup\{S \in \mathcal{T} : S \text{ is a neighboring element of } R\}.$$

and

$$R_v = \bigcup\{S \in \mathcal{T} : v \text{ is a vertex of } S\}.$$



$Tu(x, y)$ Taylor polynomial of u of degree 1 at the point x :

$$Tu(x, y) = u(x) + \nabla u(x) \cdot (y - x)$$

AVERAGED TAYLOR POLYNOMIAL

$$\psi \in C^\infty(\mathbb{R}^n) \quad \int \psi = 1$$

$$\text{supp } \psi \subset B(0, r) \quad r \leq 1/\sigma$$

For $z \in \mathcal{N}_{in}$ define:

$$\psi_z(x) = \frac{1}{h_{z,1}h_{z,2}h_{z,3}} \psi \left(\frac{z_1 - x_1}{h_{z,1}}, \frac{z_2 - x_2}{h_{z,2}}, \frac{z_3 - x_3}{h_{z,3}} \right)$$

We introduce the averaged Taylor polynomial of order 1 of u at $z \in \mathcal{N}_{in}$:

$$T_{1,z}(u)(y) = \int Tu(x, y) \psi_z(x) dx$$

Analogously, we introduce the average of u at $z \in \mathcal{N}_{in}$:

$$T_{0,z}(u) = \int u(x)\psi_z(x)dx$$

INTERPOLANT:

For $u \in H_0^1(\Omega)$ define Πu as the unique piecewise \mathcal{Q}_1 function such that, for $z \in \mathcal{N}_{in}$,

$$\Pi u(z) = T_{1,z}(u)(z)$$

$\Pi u(z) = 0$ for boundary nodes z .

$$\Pi u(x) = \sum_{z \in \mathcal{N}_{in}} T_{1,z}(u)(z) \lambda_z(x)$$

λ_z : standard basis functions

ERROR ESTIMATES IN WEIGHTED NORMS

L^2 - ERROR ESTIMATES:

THEOREM: If R is an interior element,

$$\|u - \Pi u\|_{L^2(R)} \leq C \sum_{i=1}^n h_{R,i} \left\| \delta_{\tilde{R}} \frac{\partial u}{\partial x_i} \right\|_{L^2(\tilde{R})}$$

$$C = C(\sigma, \psi)$$

Proof : First we prove stability:

$$\|\Pi u\|_{L^2(R)} \leq C \|u\|_{L^2(\tilde{R})}$$

then, for z_1 a vertex of R ,

$$\begin{aligned} \|u - \Pi u\|_{L^2(R)} &\leq \|u - T_{0,z_1}(u)\|_{L^2(R)} \\ &\quad + \|\Pi(T_{0,z_1}(u) - u)\|_{L^2(R)} \\ &\leq C \|u - T_{0,z_1}(u)\|_{L^2(R)} \end{aligned}$$

and use the dual Hardy inequality. \square

H^1 - ERROR ESTIMATES:

THEOREM: If R is an interior element,

$$\left\| \frac{\partial}{\partial x_j} (u - \Pi u) \right\|_{L^2(R)} \leq C \sum_{i=1}^n h_{R,i} \left\| \delta_{\tilde{R}} \frac{\partial^2 u}{\partial x_i \partial x_j} \right\|_{L^2(\tilde{R})}$$

Proof : VERY TECHNICAL!

IDEA: Decompose the error in two parts (as in the proof of the estimate for Lagrange interpolation)

$$u - \Pi u = \left(u - T_{1,z_1}(u) \right) + \left(T_{1,z_1}(u) - \Pi u \right)$$

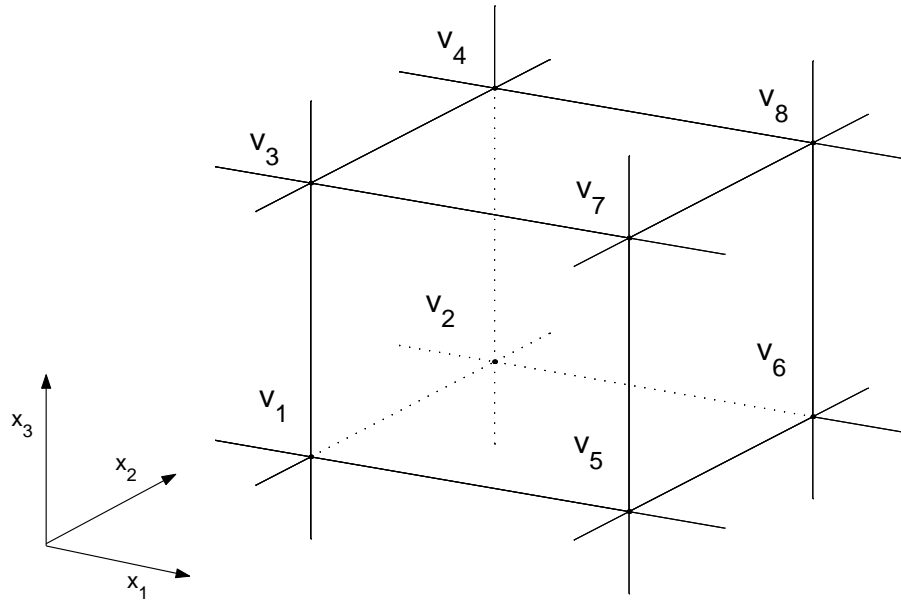
First term:

$$\left\| \frac{\partial(u - T_{1,z_1}(u))}{\partial x_1} \right\|_{L^2(R)}$$

and use the dual Hardy inequality.

Second term: $w := T_{1,z_1}(u) - \Pi u \in \mathcal{Q}_1$ then

$$\frac{\partial w}{\partial x_1} = \sum_{i=1}^4 \left(w(z_i) - w(z_{i+4}) \right) \frac{\partial \lambda_{z_i}}{\partial x_1}$$



So,

$$\left\| \frac{\partial w}{\partial x_1} \right\|_{L^2(R)} \leq \sum_{i=1}^4 |w(z_i) - w(z_{i+4})| \left\| \frac{\partial \lambda_{z_i}}{\partial x_1} \right\|_{L^2(R)}$$

But,

$$\left\| \frac{\partial \lambda_{z_i}}{\partial x_1} \right\|_{L^2(R)} \leq C \left(\frac{h_{z_i,2} h_{z_i,3}}{h_{z_i,1}} \right)^{\frac{1}{2}}$$

So, we have to estimate $|w(z_i) - w(z_{i+4})|$

For example,

$$\begin{aligned} w(z_1) - w(z_5) &= T_{1,z_5}(u)(z_5) - T_{1,z_1}(u)(z_5) \\ &= \int Tu(x, z_5)\psi_{z_5}(x)dx - \int Tu(x, z_1)\psi_{z_1}(x)dx \end{aligned}$$

Which, after long technical details! can be bounded by

$$C \frac{1}{h_{z_1,2}h_{z_1,3}} \sum_{i=1}^3 h_{z_1,i} \int \left| \frac{\partial^2 u}{\partial x_1 \partial x_i}(\bar{x}) \right| \psi(\bar{x}) d\bar{x}$$

where the function $\psi(\bar{x})$ has support in \tilde{R} . Then, use the Cauchy-Schwarz inequality and the Hardy inequality for $\psi(\bar{x})$. □

APPLICATIONS

REACTION-DIFFUSION EQUATION:

$$\begin{aligned} -\varepsilon^2 \Delta u + u &= f & \text{in } \Omega &= (0, 1)^2 \\ u &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\|v\|_\varepsilon^2 = \|v\|_{L^2}^2 + \varepsilon^2 \|\nabla v\|_{L^2}^2$$

$$\|u - u_h\|_\varepsilon \leq \|u - \Pi u\|_\varepsilon$$

THEOREM: With graded mesh:

$$\|u - u_h\|_\varepsilon \leq C \frac{\log N}{\sqrt{N}}$$

CONVECTION-DIFFUSION EQUATION:

$$\begin{aligned} -\varepsilon \Delta u - u_x - u_y &= 1 & \text{in } \Omega = (0, 1)^2 \\ u &= 0 & \text{on } \partial\Omega \end{aligned}$$

For graded meshes it follows from our weighted error estimates:

$$\|u - \Pi u\|_\varepsilon \leq C \log \frac{1}{\varepsilon} \frac{\log N}{\sqrt{N}}$$

However, in this case we don't have:

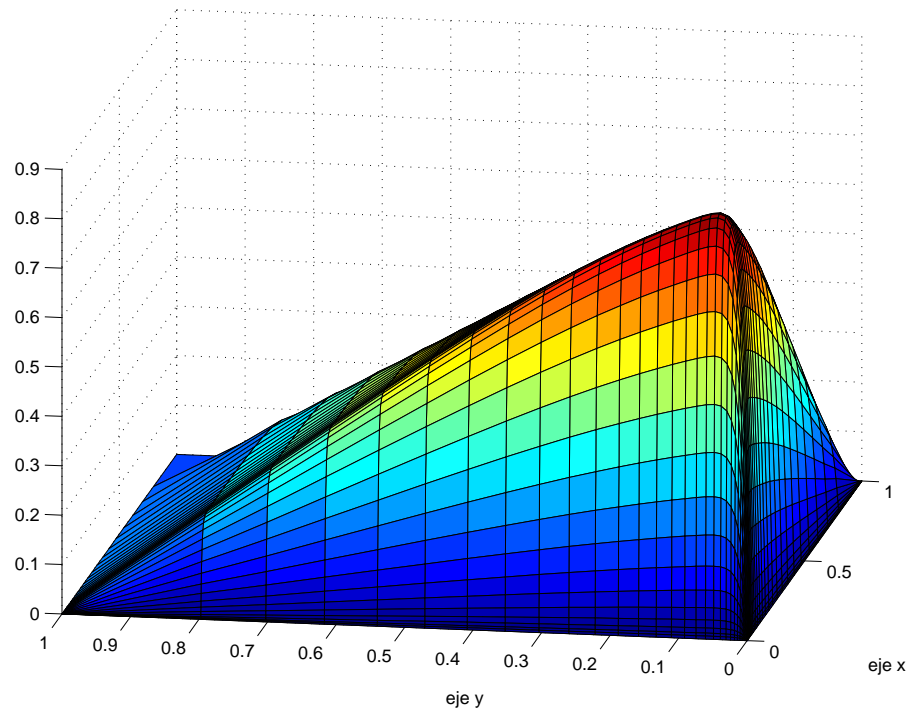
$$\|u - u_h\|_\varepsilon \leq C \|u - \Pi u\|_\varepsilon$$

So

$$\|u - u_h\|_\varepsilon \quad ? \quad \text{WE DON'T KNOW!}$$

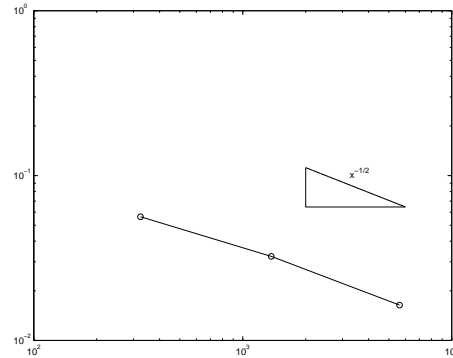
PRELIMINARY NUMERICAL EXPERIMENTS SHOW GOOD RESULTS!

Numerical Solution with graded mesh



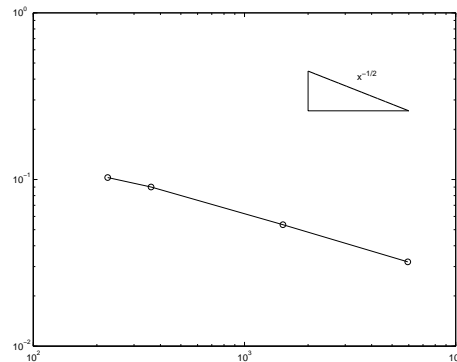
$$\varepsilon = 0.01$$

Order of convergence with graded mesh and Q_1 elements:



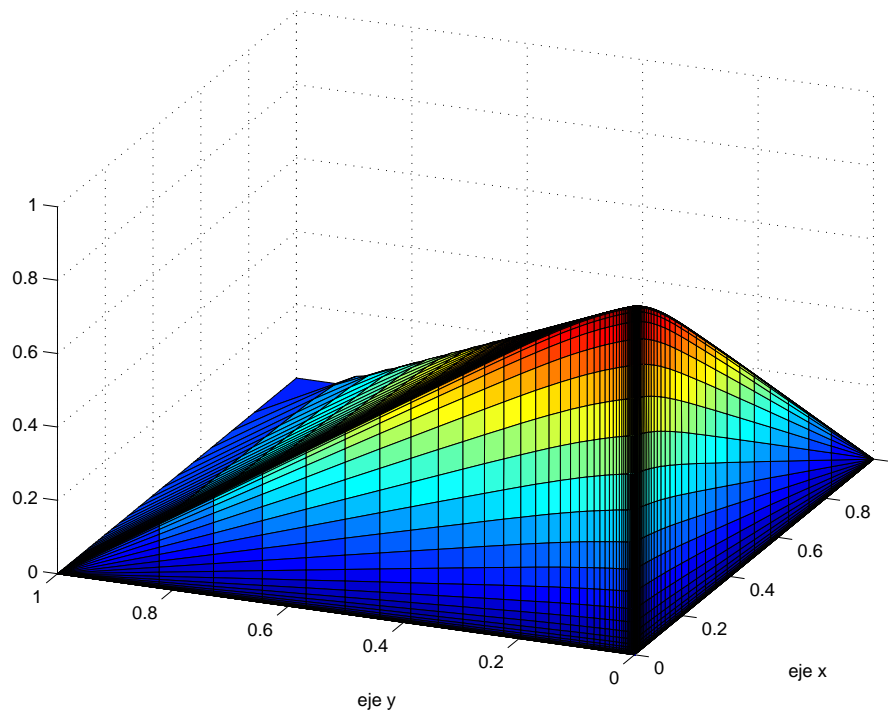
$$\varepsilon = 0.01$$

Order of convergence with Shishkin mesh and Q_1 elements:



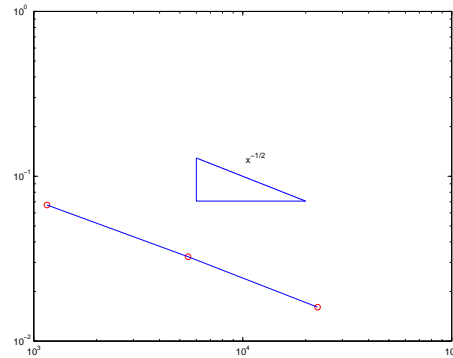
$$\varepsilon = 0.01$$

Numerical Solution with graded mesh



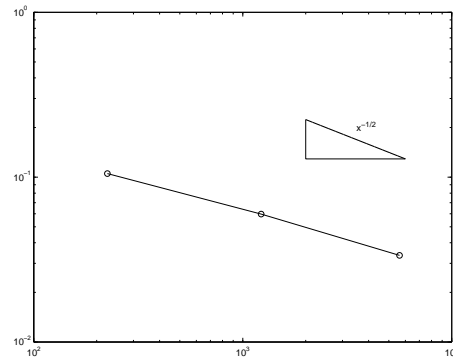
$$\varepsilon = 0.0001$$

Order of convergence with graded mesh and Q_1 elements:



$$\varepsilon = 0.0001$$

Order of convergence with Shishkin mesh and Q_1 elements:



$$\varepsilon = 0.0001$$

ORDER OF CONVERGENCE

Graded Meshes, $\varepsilon = 0.01$

# Nodes	Est. Error	Est. Order
324	0.056315	
1369	0.032338	0.38
5625	0.016383	0.48

Shishkin Meshes, $\varepsilon = 0.01$

# Nodes	Est. Error	Est. Order
225	0.102680	
361	0.090082	0.27
1521	0.053507	0.36
5929	0.032026	0.37

ORDER OF CONVERGENCE

Graded Meshes, $\varepsilon = 0.0001$

# Nodes	Est. Error	Est. Order
1156	0.066923	
5476	0.032484	0.46
22801	0.016070	0.49

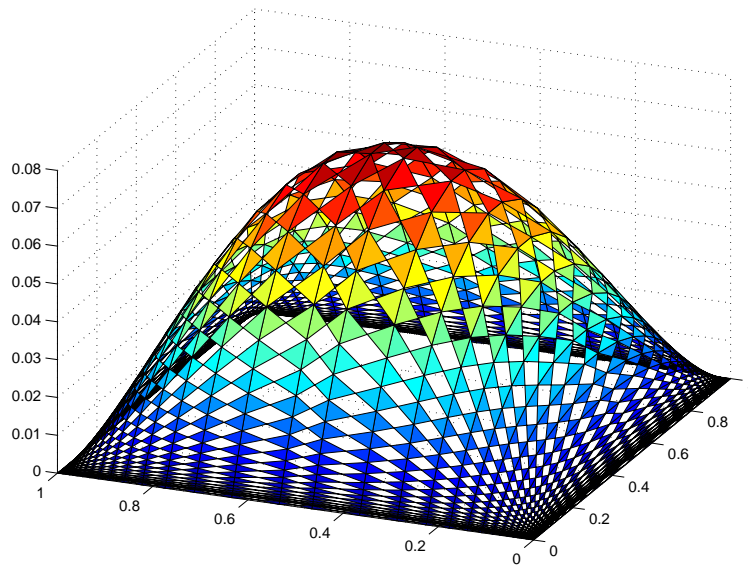
Shishkin Meshes, $\varepsilon = 0.0001$

# Nodes	Est. Error	Est. Order
225	0.105182	
1225	0.059729	0.33
5625	0.033547	0.37

FOURTH ORDER MODEL EQUATION

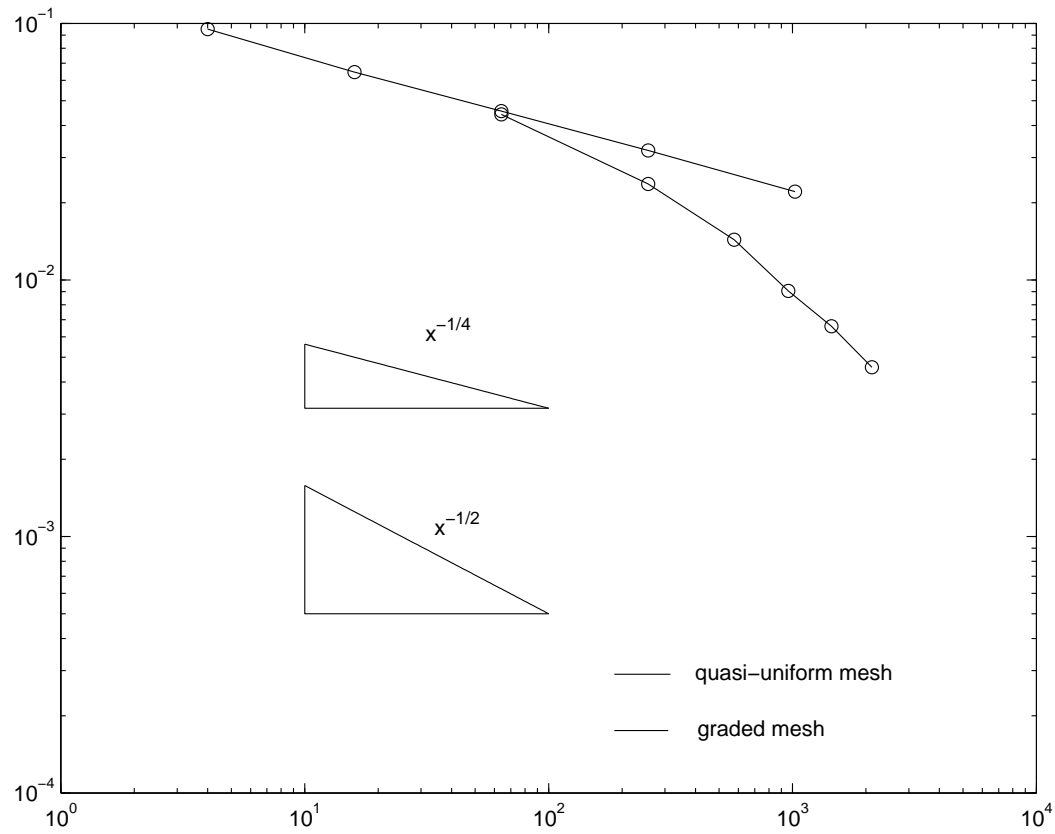
$$\begin{aligned} -\varepsilon^2 \Delta^2 u + \Delta u &= 1 && \text{in } \Omega \\ u &= 0 && \text{on } \partial\Omega \\ \frac{\partial u}{\partial n} &= 0 && \text{on } \partial\Omega \end{aligned}$$

Numerical solution with Adini's element:



$$\varepsilon = 0.001$$

Order of convergence with uniform and graded meshes:



$$\varepsilon = 0.001$$

For uniform meshes:

$$\|u - u_h\|_\varepsilon \leq \frac{C}{\sqrt[4]{N}}$$

this can be proved (but not yet written!).

Expected order for graded meshes:

$$\|u - u_h\|_\varepsilon \leq \frac{C}{\sqrt{N}}$$

BUT: NO THEORY FOR ANISOTROPIC ELEMENTS!

WE ARE WORKING ON THAT!