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Adaptive discontinuous Galerkin multiscale methods for elliptic problems

Energy norm a posteriori error estimate

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Papers

- ▶ D. Elfverson, E. Georgoulis and A. Målqvist, Adaptive discontinuous Galerkin multiscale method (submitted).

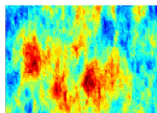
Model problem

Poisson's equation

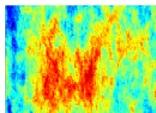
Given a polygonal domain $\Omega \subset \mathbb{R}^d$. We want to find u such that

$$\begin{aligned} -\nabla \cdot \alpha \nabla u &= f \text{ i } \Omega, \\ n \cdot \nabla u &= 0 \text{ on } \partial\Omega, \end{aligned}$$

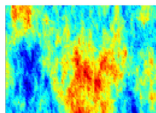
where α is bounded $0 < \beta \leq \alpha(x) \in L^\infty(\Omega)$, $f \in L^2(\Omega)$ and $\int_\Omega f \, dx = 0$.



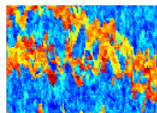
(a) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(b) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(c) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(d) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^6$

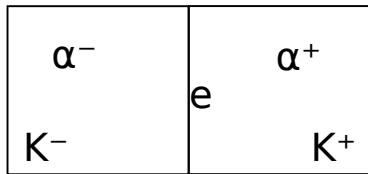
Figure: Permeabilities α projected in log scale and taken from the Society of Petroleum Engineer <http://www.spe.org/>

Discontinuous Galerkin discretization

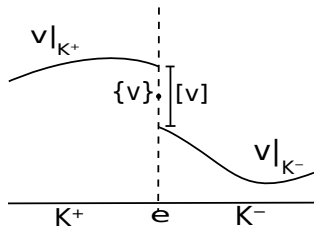
Discretization

- ▶ Let Ω be subdivided into the partition $\mathcal{K} = \{K\}$ and Γ^I be the union of all interior edges.
- ▶ Let also \mathcal{V}_h be the space of all discontinuous piecewise (bi)linear polynomials.
- ▶ Define the weighted average and jump on face e as:

$$\{v\}_w = \frac{\alpha^+ v^-}{\alpha^+ + \alpha^-} + \frac{\alpha^- v^+}{\alpha^+ + \alpha^-} \quad \text{and} \quad [v] = v^+ - v^-.$$



(a) Here $\mathcal{K} = \{K^+, K^-\}$ and $\Gamma^I = \{e\}$



(b) Example of $\{v\}$ and $[v]$

Consider a symmetric inconsistent interior penalty discontinuous Galerkin method

- ▶ Expanded DG space: $\mathcal{V} = \mathcal{V}_h + H^{1+\epsilon}$ with $\epsilon > 0$.
- ▶ Denote $\Pi : (L^2(\Omega))^d \rightarrow (\mathcal{V}_h)^d$ the L^2 -projection onto $(\mathcal{V}_h)^d$

The bilinear form $a(\cdot, \cdot) : \mathcal{V} \times \mathcal{V} \rightarrow \mathbb{R}$ and right hand side $l(\cdot) : \mathcal{V} \rightarrow \mathbb{R}$ are defined as:

$$\begin{aligned} a(v, z) &= \sum_{K \in \mathcal{K}} (\alpha \nabla v, \nabla z)_{L^2(K)} - \sum_{e \in \Gamma'} \left((\mathbf{n} \cdot \{\alpha \Pi \nabla v\}_w, [z])_{L^2(e)} \right. \\ &\quad \left. + (\mathbf{n} \cdot \{\alpha \Pi \nabla z\}_w, [v])_{L^2(K)} - \frac{\sigma_e \gamma_e}{h_e} ([v], [z])_{L^2(e)} \right), \\ l(v) &= (f, v)_{L^2(\Omega)}. \end{aligned}$$

Comments

- ▶ Why use weighted averages?
- ▶ Using Π , since we want to assume as little regularity as possible in u for the a posteriori error analysis.
- ▶ For $v \in (\mathcal{V}_h)^d$ then $\Pi v = v$ and $a(\cdot, \cdot)$ is reduced to a more familiar fashion.

Multiscale method

Motivation

In many applications, solution exist on several different scales e.g. flow in porous media and in composite materials.

- ▶ Secondary oil recovery.
- ▶ Sequestration of Carbon Dioxide.

Why do we need to resolve the coefficients?

Example with periodic coefficient

Consider Poisson's equation with periodic coefficient $\alpha = \alpha(x/\epsilon)$. For the finite element method, we have

$$\|\sqrt{\alpha}\nabla(u - u_h)\|_{L^2(\Omega)} \leq C \frac{H}{\epsilon} \|f\|_{L^2(\Omega)}$$

- ▶ Need $H \ll \epsilon$ for reliable results.
- ▶ Too computationally expensive to solve on a single mesh for many applications e.g. flow in porous media and in composite materials.
- ▶ Want to eliminate the ϵ dependence by using a multiscale method (Målqvist-Peterseim).

Framework for Multiscale methods

The problem is split into one coarse and fine scale contribution

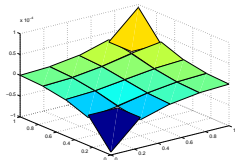
$$\mathcal{V}_h = \mathcal{V}_c \oplus \mathcal{V}_f.$$

- ▶ Let subdivide Ω into a coarse mesh $\mathcal{K}_c = \{K_c\}$.
- ▶ $\mathcal{V}_c = \text{span}\{\phi_i\} = \Pi_c \mathcal{V}_h$ and $\mathcal{V}_f = \{v \in \mathcal{V}_h : \Pi_c v = 0\}$, where $\Pi_c : \mathcal{V}_h \rightarrow \mathcal{V}_c$ is the L^2 projection onto the coarse mesh.
- ▶ Define the map $\mathcal{T} : \mathcal{V}_c \rightarrow \mathcal{V}_f$ as

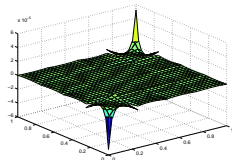
$$a(\mathcal{T}v_c, v_f) = -a(v_c, v_f), \quad \forall v_c \in \mathcal{V}_c, v_f \in \mathcal{V}_f$$

Split $u_h = u_c + \mathcal{T}u_c + u_f$ and $v = v_c + v_f$ where $u_c \in \mathcal{V}_c$, $v_f \in \mathcal{V}_f$.

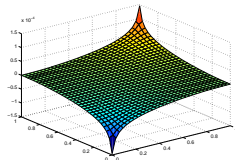
$$a(u_c + \mathcal{T}u_c + u_f, v_c + v_f) = l(v_c + v_f), \quad \forall v_c \in \mathcal{V}_c, v_f \in \mathcal{V}_f$$



(c) u_c



(d) $\mathcal{T}u_c + u_f$



(e) $u_h = u_c + \mathcal{T}u_c + u_f$

Fine scale

Let $v_c = 0$ to get the fine scale equations

$$a(\mathcal{T}u_c + u_f, v_f) = l(v_f) - a(u_c, v_f),$$

split into two equations

$$\begin{aligned} a(u_f, v_f) &= l(v_f) \quad \forall v_f \in \mathcal{V}_f, \\ a(\mathcal{T}u_c, v_f) &= -a(u_c, v_f) \quad \forall v_f \in \mathcal{V}_f. \end{aligned}$$

Coarse scale

Let $v_f = 0$ on the coarse scale

$$a(u_c + \mathcal{T}u_c, v_c) = l(v_c) - a(u_f, v_c) \quad \forall v_c \in \mathcal{V}_c$$

Comments

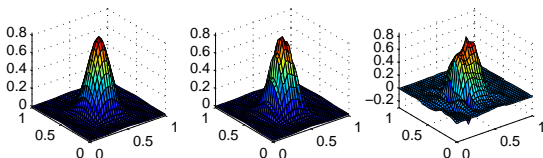
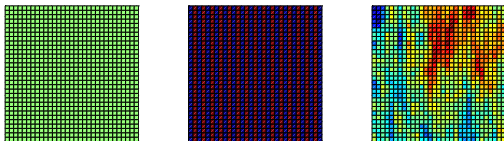
- ▶ Equally hard to solve as the original problem.
- ▶ Other choices than Π_c can be coincided.
- ▶ A symmetric split can also be considered for the coarse scale

View solution as span of modified basis functions

- ▶ Let $\mathcal{V}_c = \text{span}\{\phi_i\}$ and $\mathcal{V}^{ms} = \text{span}\{\phi_i + \mathcal{T}\phi_i\}$.
- ▶ View $\phi_i + \mathcal{T}\phi_i$ as a modified basis function.

From the multiscale map we have, $\mathcal{V}_h = \mathcal{V}_{ms} \perp_a \mathcal{V}_f$, for all i

$$a(\phi_i + \mathcal{T}\phi_i, v) = 0, \quad v \in \mathcal{V}_f$$



Approximation of $\mathcal{T}\phi_i$

- ▶ The fast decay of $\mathcal{T}\phi_i$ motivates approximations of $\mathcal{T}\phi_i$ to patches $\omega_i^l \subset \Omega$.

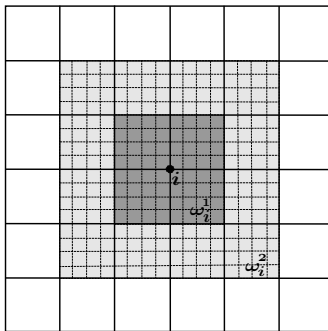
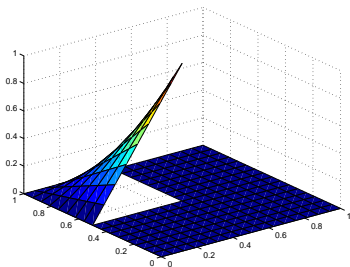


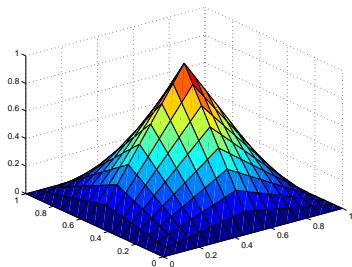
Figure: Example of a one layer patch ω_i^1 and a two layer patch ω_i^2

Multiscale method discretization

- ▶ $\tilde{\mathcal{T}}$ is the restriction of \mathcal{T} to a patch $\omega \subset \Omega$
- ▶ $\tilde{U}_f = \sum_{i \in \mathcal{N}} \tilde{U}_{f,i}$ where \mathcal{N} is the number of nodes, be the approximation of u_f .
- ▶ Let \mathcal{M}_i be all j s.t $\phi_j = 1$ in node i .
- ▶ Let also $\Phi_i = \sum_{j \in \mathcal{M}_i} \phi_j$



(a) ϕ_j



(b) $\Phi_i = \sum_{j \in \mathcal{M}_i} \phi_j$

Bilinear form for the fine scale problem

- ▶ Let $\mathcal{K}(\omega_i^L) = \{K : K \cap \omega_i^L \neq \emptyset\}$.
- ▶ Let also $\Gamma^I(\omega_i^L)$ be all interior edges on $\mathcal{K}(\omega_i^L)$.

Define $a_i : \mathcal{V}_f(\omega_i) \times \mathcal{V}_f(\omega_i) \rightarrow \mathbb{R}$, as

$$\begin{aligned} a_i(v, z) &= \sum_{K \in \mathcal{K}(\omega_i^L)} (\alpha \nabla v, \nabla z)_{L^2(K)} - \sum_{e \in \Gamma^I(\omega_i^L)} \left((\mathbf{n} \cdot \{\alpha \Pi \nabla v\}_w, [z])_{L^2(e)} \right. \\ &\quad \left. + (\mathbf{n} \cdot \{\alpha \Pi \nabla z\}_w, [v])_{L^2(K)} - \frac{\sigma_e \gamma_e}{h_e} ([v], [z])_{L^2(e)} \right), \\ l_i(v) &= (\Phi_i f, v)_{L^2(\Omega)}. \end{aligned}$$

Fine scale equations

For all $i \in \mathcal{N}$: find $\tilde{T}\phi_j \in \mathcal{V}_f(\omega_i^L)$ and $U_{f,i} \in \mathcal{V}_f(\omega_i^L)$ for $j \in \mathcal{M}_i$ s.t

$$a_i(\tilde{T}\phi_j, v_f) = -a_i(\phi_j, v_f), \quad \forall v_f \in \mathcal{V}_f(\omega_i^L),$$

$$a_i(\tilde{U}_{f,i}, v_f) = l_i(\Phi_i v_f), \quad \forall v_f \in \mathcal{V}_f(\omega_i^L).$$

Coarse scale equation

Find $U_c \in \mathcal{V}_c$ s.t

$$a(U_c + \tilde{T}U_c, v_c) = l(v_c) - (\tilde{U}_f, v_c), \quad \forall v_c \in \mathcal{V}_c.$$

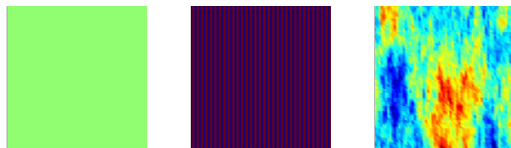
Decay in V_f

Problem setting

- ▶ Let the computational domain be ω_i^L for $L = 1, 2, \dots, N$ where $\omega_i^L \subseteq \Omega$.
- ▶ Let also $\Phi_i = \sum_{j \in \mathcal{M}_i} \phi_j$
- ▶ The problem reads: find $\tilde{\mathcal{T}}\Phi_i \in \mathcal{V}_h(\omega_i^L)$

$$a(\tilde{\mathcal{T}}\Phi_i, v) = -a(\Phi_i, v), \quad \forall v \in \mathcal{V}_h(\omega_i^L).$$

- ▶ The reference solution $\mathcal{T}\Phi_i$ is the solution computed on $\omega_i^N = \Omega$.
- ▶ Coarse mesh is 8×8 element and reference grid is 64×64 elements.



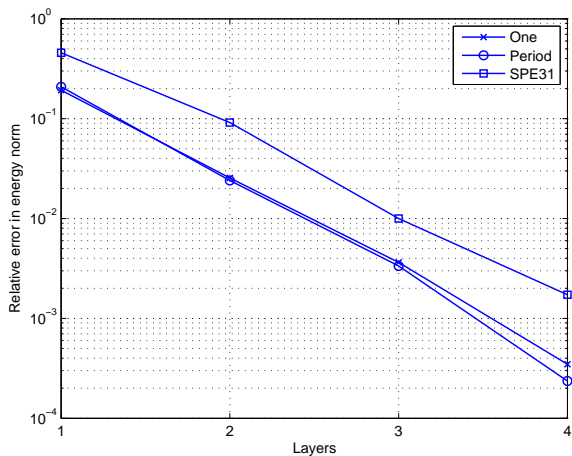


Figure: The error in relative error in broken energy norm with respect to the path size.

Convergence

Problem setting

- ▶ Consider the model problem (Poisson's equation)
- ▶ Keeping the refinement level constant and increasing the patch sizes $L = 1, \dots, N$ for all local problems.
- ▶ The coarse grid is 8×8 coarse elements.
- ▶ The reference solution U_{ref} is the DG solution computed on 64×64 elements.
- ▶ The right hand side is -1 in the lower left corner and 1 in the upper right.

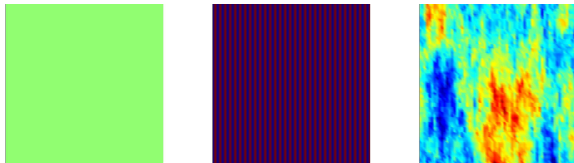


Figure: Permeabilities α .

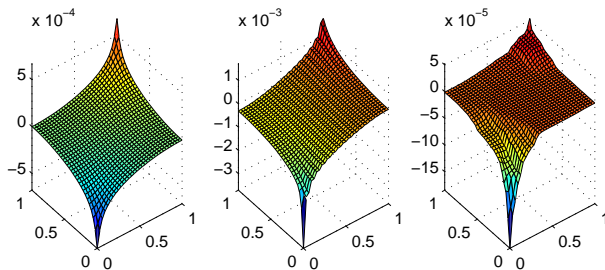


Figure: The reference solution to the model problem using the permeabilities *One*, *Period* and *SPE*

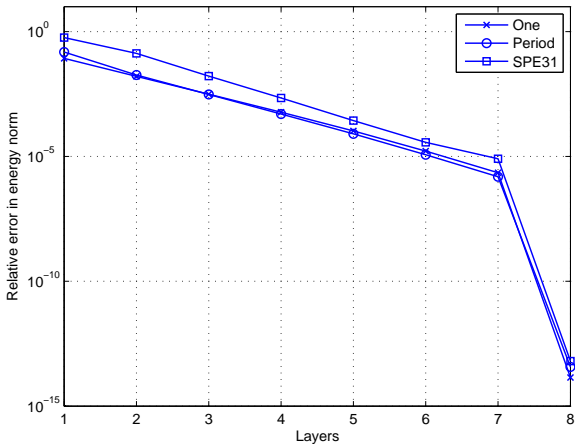


Figure: The relative error in broken energy norm with respect to the patch sizes.

Implementation

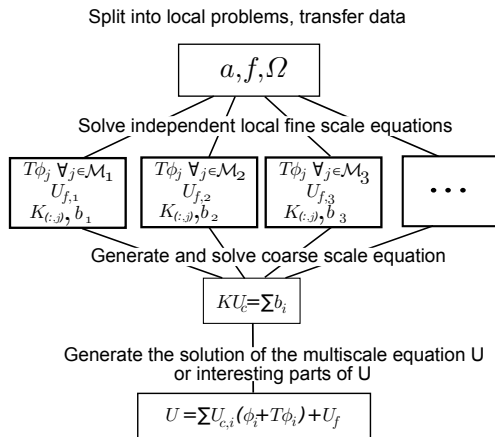


Figure: Scheme of the implementation.

Constraints on the fine scale equations

- ▶ The condition is realised using Lagrangian multiplier.
- ▶ Let ϕ be a coarse basis function and φ be a fine basis function.

Want so fined $\tilde{T}w \in \mathcal{V}_f(\omega_i^L)$

$$a_i(\tilde{T}w, v) = -a_i(w, v) \quad \forall v \in \mathcal{V}_f(\omega_i^L).$$

Algebraic problem reads:

$$\begin{pmatrix} K & P^T \\ P & 0 \end{pmatrix} \xi = \begin{pmatrix} b \\ 0 \end{pmatrix},$$

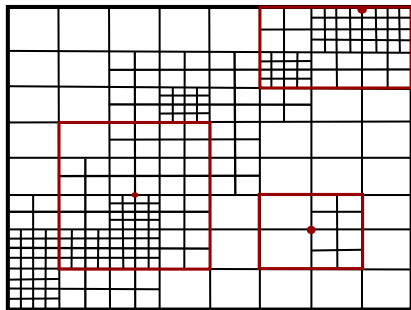
where $K_{k,l} = a_i(\varphi_k, \varphi_l)$, $b_k = -a_i(\phi_l, \varphi_k)$ and

$$P = \begin{pmatrix} (\phi_1, \varphi_1) & (\phi_1, \varphi_2) & \dots & (\phi_1, \varphi_N) \\ (\phi_2, \varphi_1) & (\phi_2, \varphi_2) & \dots & (\phi_2, \varphi_N) \\ \vdots & \vdots & \ddots & \vdots \\ (\phi_M, \varphi_1) & (\phi_M, \varphi_2) & \dots & (\phi_M, \varphi_N) \end{pmatrix}.$$

Adaptivity

Set up

- ▶ Need an a posteriori error estimate for the discontinuous Galerkin method.
- ▶ Use this in the framework for Multiscale methods to construct a a posteriori error estimate for the multiscale method.
- ▶ Construct a adaptive algorithm to automatically tune the critical parameters.



Theorem (A posteriori error estimate for DG method)

- ▶ Let u , u_h be given by the exact solution respectively the DG solution.
- ▶ Let also $\chi \in \mathcal{V}_h \cap H^1(\Omega)$
- ▶ Moreover, let $\mathcal{E} := \mathcal{E}_c + \mathcal{E}_d$ where $\mathcal{E}_c := u - \chi$ and $\mathcal{E}_d := \chi - u_h$.

Then,

$$\sum_{K \in \mathcal{K}} \|\sqrt{\alpha} \nabla \mathcal{E}_c\|_{L^2(K)}^2 \lesssim \sum_{K \in \mathcal{K}} (\varrho_K(u_h) + \zeta_K(u_h, \chi))^2,$$

where

$$\begin{aligned} \varrho_K(u_h) &= \frac{h_K}{\sqrt{\alpha_0}} \|f + \nabla \cdot \alpha \nabla u_h\|_{L^2(K)}, \\ &+ \sqrt{\frac{h_K}{\alpha_0}} \left(\|(1 - w_{K(e)}) n \cdot [\alpha \nabla u_h]\|_{L^2(\partial K)} + \left\| \frac{\sigma_e \gamma_e}{h_e} [u_h] \right\|_{L^2(\partial K \setminus \Gamma^B)} \right), \end{aligned}$$

$$\zeta_K(u_h, \chi) = \|\sqrt{\alpha} \nabla (u_h - \chi)\|_{L^2(K)}.$$

Treatment of $\zeta_K(u_h, \chi)$

1. First, χ has to be chosen in a clever way.
2. Second, $\zeta_K(u_h, \chi)$ can either be estimated or evaluated.
 - ▶ One possible choice is $\chi = \mathcal{I}_{Os} u_h$.
 - ▶ Under certain assumption on α , $\zeta_K(u_h, \chi)$ can be evaluated and hidden in $\varrho_K(u_h)$.

Lemma (Oswald interpolation operator)

- ▶ Let $\mathcal{I}_{Os} : \mathcal{V}_h \rightarrow \mathcal{V}_h \cap H^1$

$$\mathcal{I}_{Os} v = \sum_{i \in \mathcal{N}} \left(\frac{1}{|\mathcal{M}_i|} \sum_{j \in \mathcal{M}_i} v_j(x_i) \varphi_j \right).$$

Then,

$$\|\sqrt{\alpha} \nabla(v - \mathcal{I}_{Os} v)\|^2 \lesssim \alpha^0 \left\| \frac{1}{\sqrt{h_e}} [v] \right\|_{L^2(\partial K \setminus \Gamma^B)}^2.$$

Sketch of proof

We have

$$\sum_{K \in \mathcal{K}} \|\sqrt{\alpha} \nabla \mathcal{E}_c\|_{L^2(K)}^2 = a(\mathcal{E}_c, \mathcal{E}_c) = a(\mathcal{E}, \mathcal{E}_c) - a(\mathcal{E}_d, \mathcal{E}_c),$$

where

$$\begin{aligned} a(\mathcal{E}, \mathcal{E}_c) &= a(u, \mathcal{E}_c) - a(u_h, \mathcal{E}_c) = l(\mathcal{E}_c) - a(u_h, \mathcal{E}_c), \\ &= l(\eta) - a(u_h, \eta), \end{aligned}$$

where $\eta = \mathcal{E}_c - \pi_0 \mathcal{E}_c$.

- First integration by parts $l(\eta) - a(u_h, \eta)$ element wise and using the identity $[vz] = \{v\}_w[z] + \{v\}_{\bar{w}}[z]$.

We get,

$$\begin{aligned}
 & l(\eta) - a(u_h, \eta) \\
 &= \sum_{K \in \mathcal{K}} (f + \nabla \cdot \alpha \nabla u_h, \eta)_{L^2(K)} + \sum_{e \in \Gamma^I} \left(- (n \cdot [\alpha \nabla u_h], \{\eta\}_{\bar{w}})_{L^2(e)} \right. \\
 & \left. + (n \cdot \{\alpha \Pi \nabla \eta\}_w, [u_h])_{L^2(e)} - \sigma \gamma_e h_e^{-1} ([u_h], [\eta])_{L^2(e)} \right) + \sum_{e \in \Gamma^B} (n \cdot \alpha \nabla u_h, \eta)_{L^2(e)}.
 \end{aligned}$$

1. Then, using the inequalities and stability for the piecewise constant L^2 -projection.

$$\|v - \pi_0 v\|_{L^2(K)} \lesssim \frac{h_K}{\sqrt{\alpha_0}} \|\sqrt{\alpha} \nabla v\|_{L^2(K)}, \quad \forall v \in H^1(K),$$

$$\|v - \pi_0 v\|_{L^2(\partial K)} \lesssim \sqrt{\frac{h_K}{\alpha_0}} \|\sqrt{\alpha} \nabla v\|_{L^2(K)} \quad \forall v \in H^1(K).$$

2. For $a(\mathcal{E}_d, \mathcal{E}_c)$ use the Lemma (Oswald interpolation operator).

Theorem (A posteriori error estimate for ADG-MS)

- ▶ Let u, U be the exact solution respectively the multiscale solution.
- ▶ Let $X = \mathcal{I}_{O_s} U \in H^1(\Omega)$.
- ▶ Set $\mathcal{E} := \mathcal{E}_c + \mathcal{E}_d$ where $\mathcal{E}_c := u - X$ and $\mathcal{E}_d := X - U$.
- ▶ $U_i := \sum_{j \in \mathcal{M}_i} U_{c,j}(\phi_j + \tilde{\mathcal{T}}\phi_j) + U_{f,i}$, where $U_{c,j}$ are the nodal values.

Then,

$$\sum_{K \in \mathcal{K}} \|\sqrt{\alpha} \nabla \mathcal{E}_c\|_{L^2(K)}^2 \lesssim \sum_{K_c \in \mathcal{K}_c} \rho_{h,K_c}^2 + \sum_{i \in \mathcal{N}} \rho_{L,\omega_i^t}^2,$$

where

$$\rho_{L,\omega_i^t}^2 = \sum_{e \in \Gamma^B(\omega_i^t) \setminus \Gamma^B} \rho_{L,\omega_i^t,e}^2,$$
$$\rho_{L,\omega_i^t,e} = \frac{H_{\omega_i^t}}{\sqrt{h_K \alpha_0}} \left(\|n \cdot \{\alpha \nabla U_i\}_w\|_{L^2(e)} + \frac{\sigma_e \gamma_e}{h_e} \| [U_i] \|_{L^2(e)} \right),$$

measures the effect of the truncated patches.

Also

$$\rho_{h,K_c}^2 = \sum_{K \in K_c} (\varrho_K(U) + \zeta_K(U, X))^2,$$

with ϱ_K and ζ_K as in previous theorem.

Comments

- ▶ $\rho_{L,\omega_i^t}^2$ measure the effect of the truncated patches.
- ▶ $\rho_{h,K}^2$ measure the effect of the refinement level.
- ▶ $\frac{1}{\sqrt{h_K h_e}} \|[U_i]\|_{L^2(e)}$ behave as $h^{-3/2} e^{-L} \sim 1 \Rightarrow L \sim \frac{3}{2} \log h^{-1}$
- ▶ Another possible choice is a weighted Oswald-type interpolation operator with the weights depending on the diffusion tensor.

Sketch of proof

We have

$$a(\mathcal{E}_c, \mathcal{E}_c) = a(\mathcal{E}, \mathcal{E}_c) - a(\mathcal{E}_d, \mathcal{E}_c),$$

where

$$\begin{aligned} a(\mathcal{E}, \mathcal{E}_c) &= a(u, \mathcal{E}_c) - a(U, \mathcal{E}_c), \\ &= l(\mathcal{E}_c) - a(U, \mathcal{E}_c), \\ &= l(\mathcal{E}_c - v_c) - a(U, \mathcal{E}_c - v_c), \\ &= \sum_{i \in \mathcal{N}} \left(l_i(\mathcal{E}_c - v_c - v_f) - a(U_i, \mathcal{E}_c - v_c) + a_i(U_i, v_f) \right). \end{aligned}$$

here $v_c \in \mathcal{V}_c$ and $v_f \in \mathcal{V}_f$.

Notice that

$$\begin{aligned} a_i(\mathcal{E}, \mathcal{E}_c) &= a(\mathcal{E}, \mathcal{E}_c) + \sum_{e \in \Gamma^B(\omega_i^L) \Gamma^B} \left((n \cdot \{\alpha \nabla U_i\}_w, [v_f])_{L^2(e)} \right. \\ &\quad \left. + (n \cdot \{\alpha \nabla U_i\}_w, [v_f])_{L^2(e)} - \frac{\sigma_e \gamma_e}{h_e} ([U_i], [v_f])_{L^2(e)} \right) \end{aligned}$$

Then,

$$\begin{aligned} a(\mathcal{E}, \mathcal{E}_c) &= \left(I(\mathcal{E}_c - v_c - v_f) - a(\mathcal{E}, \mathcal{E}_c - v_c - v_f) \right) \\ &+ \sum_{i \in \mathcal{N}} \sum_{e \in \Gamma^B(\omega_i^L) \Gamma^B} \left((n \cdot \{\alpha \nabla U_i\}_w, [v_f])_{L^2(e)} + (n \cdot \{\alpha \nabla U_i\}_w, [v_f])_{L^2(e)} \right) \\ &- \frac{\sigma_e \gamma_e}{h_e} ([U_i], [v_f])_{L^2(e)} \\ &=: I + II. \end{aligned}$$

1. The first term (I), is bounded by the a apoteriori error estimate for DG.
2. To bound the second term (II),
 - ▶ Select v_c and v_f as the piecewise constant L^2 -projection onto V_c and V_f , respectively.
 - ▶ Then using a trace inequality, a interpolation estimate and L^2 -stability of π_f , II is bounded.

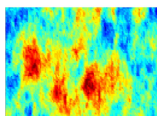
Adaptive algorithm

Algorithm 1 Adaptive Discontinuous Galerkin Multiscale Method

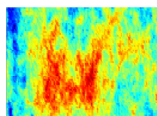
- 1: Initialize the coarse mesh with mesh size H .
 - 2: Let the fine mesh size be $h_K = H/2$ for all $K_c \in \mathcal{K}_c$ and $L(\omega_i) = 2$ for all $i \in \mathcal{N}$
 - 3: **while** $\sum_{i \in \mathcal{N}} (\rho_{h, \omega_i}^2 + \rho_{L, \omega_i}^2) > TOL$ **do**
 - 4: **for** $i \in \mathcal{N}$ **do**
 - 5: **if** $\rho_{L, \omega_i}^2 > TOL/(2\mathcal{N})$ **then**
 - 6: $L(\omega_i) := L(\omega_i) + 1$
 - 7: **end if**
 - 8: **end for**
 - 9: **for** $K_c \in \mathcal{K}_c$ **do**
 - 10: **if** $\rho_{h, K}^2 > TOL/(2|\mathcal{K}_c|)$ **then**
 - 11: $h_K := h_K/2$
 - 12: **end if**
 - 13: **end for**
 - 14: **end while**
-

Adaptivity

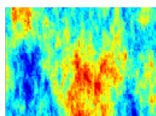
- ▶ Consider the model problem
- ▶ Using the a posteriori error estimate to construct an adaptive algorithm.
- ▶ Start with one refinement and 2 layers patches everywhere.
- ▶ Refine 30% of the coarse elements and increase 30% of the patch sizes in each iteration.
- ▶ Coarse mesh is 32×32 element and reference grid is 256×256 elements.
- ▶ The right hand side is -1 in the lower left corner and 1 in the upper right.



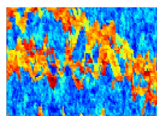
(a) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(b) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(c) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^5$



(d) $\frac{\alpha_{max}}{\alpha_{min}} \sim 10^6$

Figure: Permeabilities α projection in log scale.

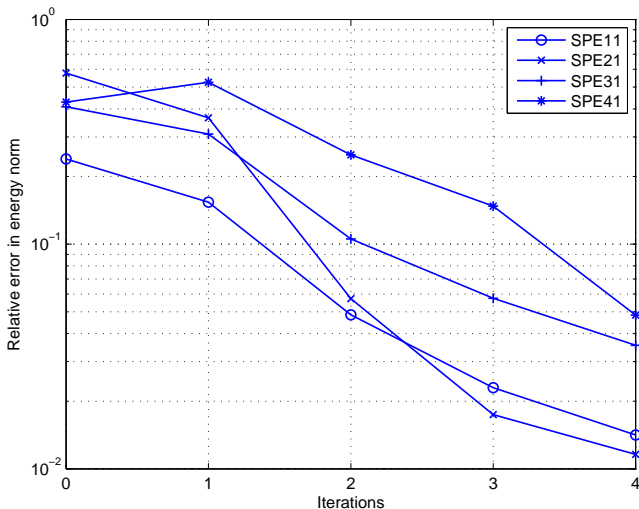
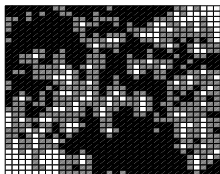


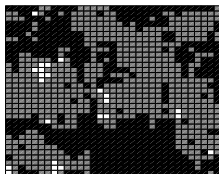
Figure: The relative error in broken energy norm with respect to number of iterations. Iteration 0 corresponds to the standard DG solution and iteration 1 the start values in the adaptive algorithm.

Figure (a) and (b) illustrates where the adaptive algorithm puts most effort

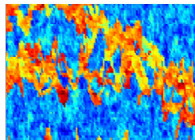
- ▶ Figure (a) corresponds to the refinements
- ▶ Figure (b) corresponds to the patch sizes.
- ▶ Figure (c) is the permeability α .



(a) Refine h_K



(b) Layers, L



(c) Layers, L

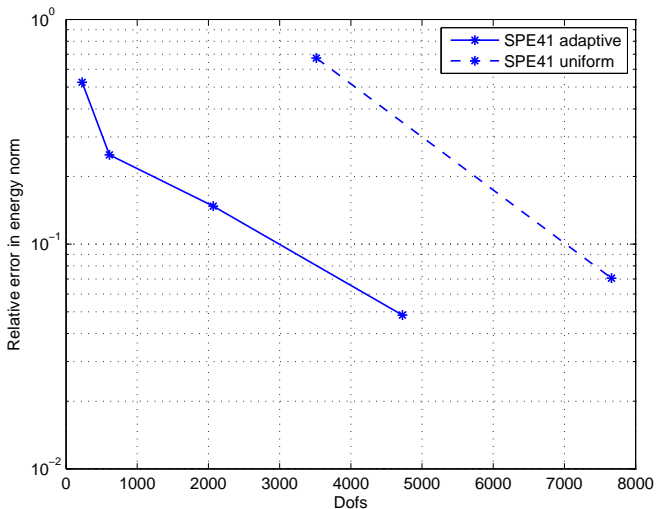


Figure: The relative error in broken energy norm with respect to the mean value of the degrees of freedom for the fine scale problems.

Conclusions

Advantage

- ▶ The fine scale problems are perfectly parallelizable.
- ▶ The exponential decay in the fine scale solution allows small patches.
- ▶ The error estimate and the adaptivity algorithm focus computational effort in critical areas.
- ▶ Very high aspect ratio in α can be solved.
- ▶ Possible to construct a conservative flux on the coarse scale.

Future work

- ▶ A priori analysis of the discontinuous Galerkin multiscale method.
- ▶ Hybrid method using DG on coarse scale and CG on fine scale.
- ▶ Extend the method and analysis to diffusion-convection problems.
- ▶ Investigate to sensibility in input data (uncertainty in the permeability α)
- ▶ Extend the implementation to triangular meshed to allow for complicated geometries.
- ▶ 3D implementation.



Questions