



# **Adaptive Variational Multiscale Method: Basic A Posteriori Error Estimation Framework**

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# Goal

- We want to find computational methods for solving multiscale problems in a Galerkin finite element setting.
- We need an a posteriori estimation framework to measure the reliability of our solution.
- We also want to use the error bounds for adaptivity.
- We start with two scales in two dimensions.

# Outline

- Model Problem
- Variational Multiscale Method
- Choice of Coarse and Fine Spaces
- The Basic Idea of our Method
- Error Estimates
- Adaptive Strategy
- Numerical Examples
- Future Work

# Model Problem

**Poisson Equation.** Find  $u \in H_0^1(\Omega)$  such that

$$-\nabla \cdot a \nabla u = f \quad \text{in } \Omega, \quad u = 0 \text{ on } \partial\Omega.$$

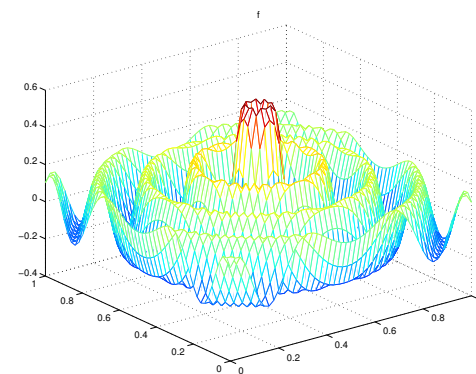
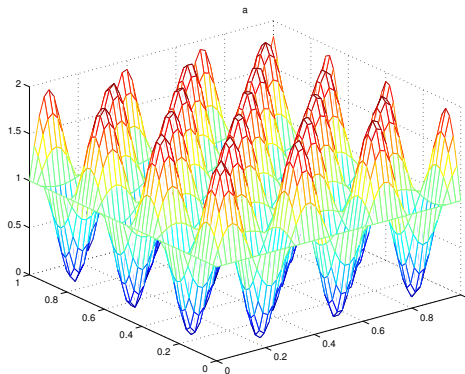
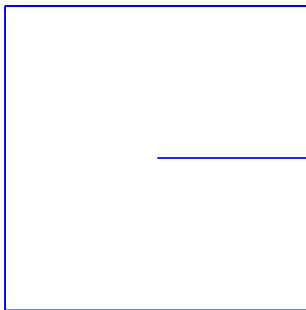
where  $f \in H^{-1}(\Omega)$ ,  $a > 0$  bounded, and  $\Omega$  is a domain in  $\mathbb{R}^d$ ,  $d = 1, 2, 3$ .

**Weak form.** Find  $u \in H_0^1(\Omega)$  such that

$$(a \nabla u, \nabla v) = (f, v) \quad \text{for all } v \in H_0^1(\Omega).$$

# Multiscale Problems

Below are three examples of multiscale problems.



The first one represents difficulties in the domain (cracks, holes, ...) the second one oscillations in  $a$  and the third one oscillations in  $f$ .

# Variational Multiscale Method

- See for instance T.J.R. Hughes (1995).
- $H_0^1 = V_c \oplus V_f$ ,  $u = u_c + u_f$ , and  $v = v_c + v_f$ .

Find  $u_c \in V_c$  and  $u_f \in V_f$  such that

$$\begin{aligned}(a \nabla u_c, \nabla v_c) + (a \nabla u_f, \nabla v_c) &= (f, v_c) \quad \text{for all } v_c \in V_c, \\ (a \nabla u_f, \nabla v_f) &= (f, v_f) - (a \nabla u_c, \nabla v_f) \\ &:= (R(u_c), v_f) \quad \text{for all } v_f \in V_f.\end{aligned}$$

# Variational Multiscale Method

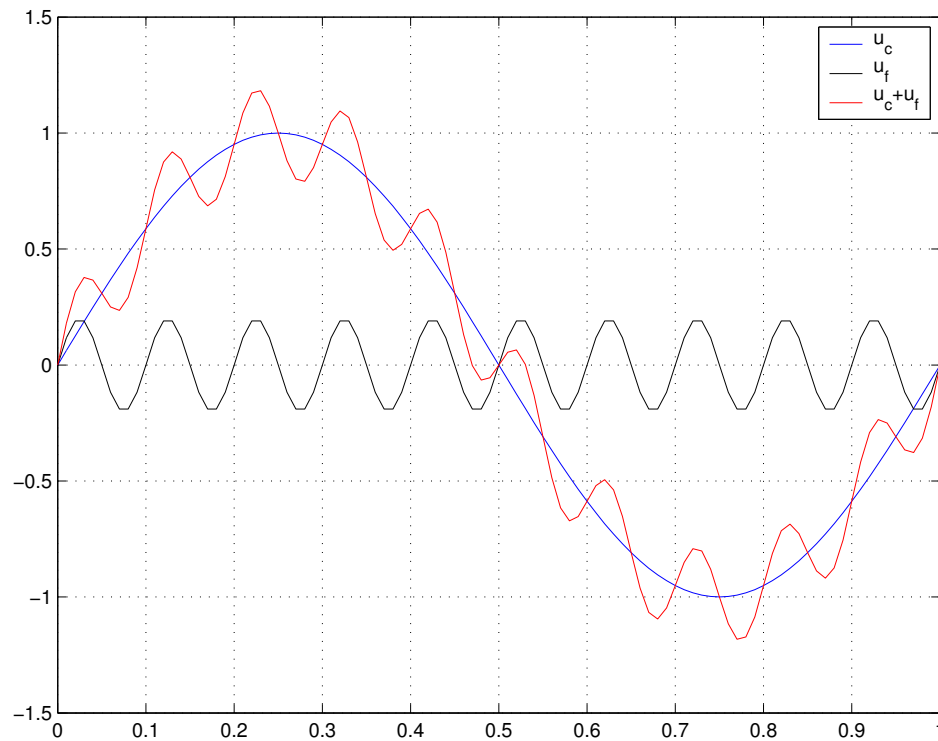


Figure 1:  $u_c$ ,  $u_f$ , and  $u_c + u_f$ .

# Variational Multiscale Method

- The fine scale is driven by the coarse scale residual.
- Approximation to fine scale solution solved on each element analytically (Green's functions).
- Fine scale information is then used to modify the coarse scale equation.

$$(a \nabla u_c, \nabla v_c) + (a \nabla \hat{A}_f^{-1} R(U_c), \nabla v_c) = (f, v_c) \quad \forall v_c \in V_c.$$



# Choice of $V_c$ and $V_f$

We use the splits proposed by Vassilevski-Wang (1998) and also used by Aksoylu-Holst (2004).

- Hierarchical basis, HB.
- Wavelet modified hierarchical basis, WHB.

The aim with WHB is to make  $V_f$  more  $L^2(\Omega)$  orthogonal to  $V_c$  than in ordinary HB.

$$(Q_c^a v, w) = (v, w), \quad \text{for all } w \in V_c.$$

$$\varphi_{WHB} = (I - Q_c^a)\varphi_{HB}.$$

# Choice of $V_c$ and $V_f$

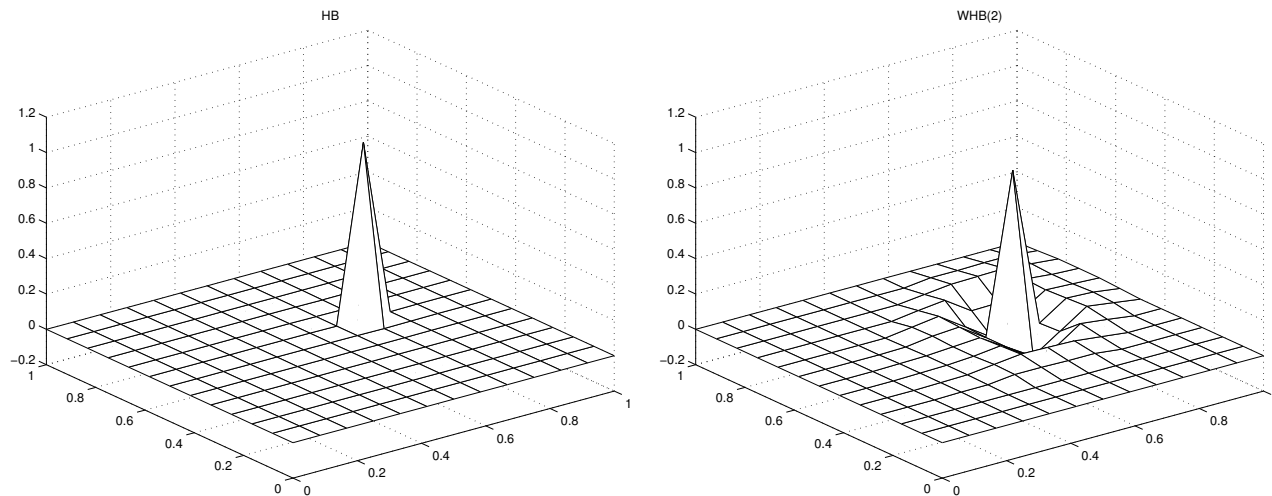


Figure 2: HB-function and WHB-function with two Jacobi iterations.

# Our Basic Idea

- Discretization of  $V_f$  by (W)HB-functions ( $V_f^h$ ).
- Solve localized fine scale problems for each coarse node (or some coarse nodes).
- Possibility to do this in parallel.
- A posteriori error estimation framework.
- Adaptive strategy for this setting.

# Decouple Fine Scale Equations

Remember the fine scale equations:

$$(a \nabla U_f, \nabla v_f) = (R(U_c), v_f), \quad \text{for all } v_f \in V_f^h.$$

Include a partition of unity,

$$(a \nabla U_f, \nabla v_f) = (R(U_c), v_f) = \sum_{i=1}^n (R(U_c), \varphi_i v_f),$$

let  $U_f = \sum_i^n U_{f,i}$  where

$$(a \nabla U_{f,i}, \nabla v_f) = (R(U_c), \varphi_i v_f).$$

# Approximate Solution

Find  $U_c \in V_c$  and  $U_f = \sum_i^n U_{f,i}$  where  $U_{f,i} \in V_f^h(\omega_i)$  such that

$$(a \nabla U_c, \nabla v_c) + (a \nabla U_f, \nabla v_c) = (f, v_c) \quad \text{for all } v_c \in V_c,$$
$$(a \nabla U_{f,i}, \nabla v_f) = (R(U_c), \varphi_i v_f) \quad \text{for all } v_f \in V_f^h(\omega_i).$$

- Since  $\varphi_i$  has support on a star  $S_i^1$  in node  $i$  we solve the fine scale equations approximately on  $\omega_i$  with  $U_{f,i} = 0$  on  $\partial\omega_i$ .

# Refinement and Layers

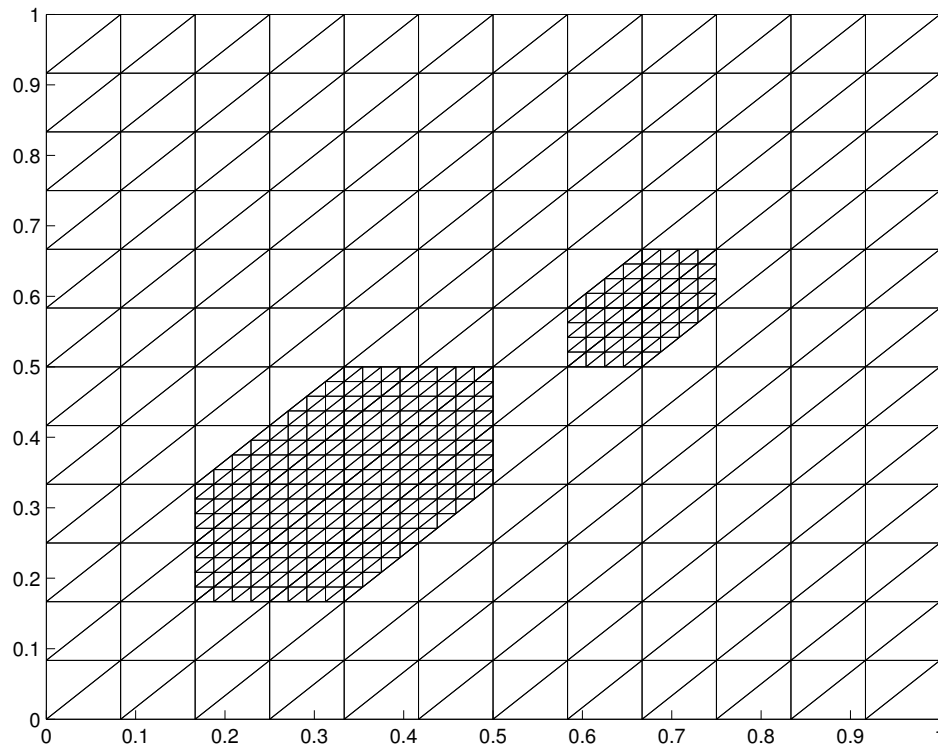


Figure 3: One,  $S_i^1$ , and two,  $S_i^2$ , layer stars.

# Iterative or Direct

**Iterative**  $U_{f,i}^0 = 0,$

$$(a \nabla U_c^k, \nabla v_c) = (f, v_c) - (a \nabla U_f^{k-1}, \nabla v_c),$$

$$(a \nabla U_{f,i}^k, \nabla v_f) = (R(U_c^k), \varphi_i v_f),$$

or in matrix form,

$$A_c U_c^k = b_c(U_f^{k-1})$$

$$\hat{A}_f U_{f,i}^k = b_f(U_c^k)$$

# Iterative or Direct

## Direct

$$(a\nabla U_c, \nabla v_c) + (\nabla \hat{A}_f^{-1} R(U_c), \nabla v_c) = (f, v_c)$$

or in matrix form,

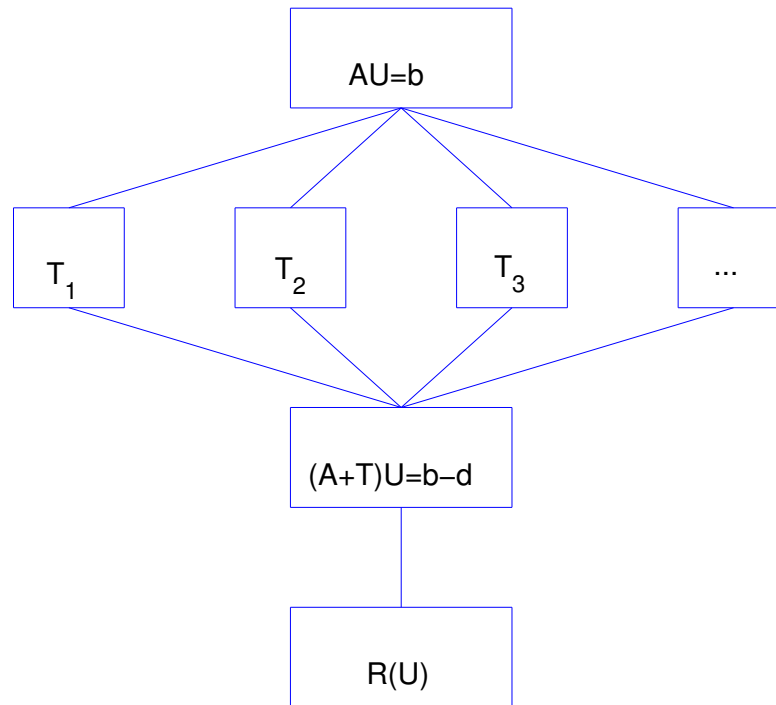
$$(A_c + T)U_c = b - d,$$

where  $b_j = (f, \varphi_j)$ ,

$$T_{ij}\varphi_j + d_i = (\nabla \hat{A}_f^{-1}(R(\varphi_i)), \nabla \varphi_j).$$



# Algorithm



# Error Estimation

We let  $e = u - U = u_c + \sum_{i=1}^n u_{f,i} - U_c - \sum_{i=1}^n U_{f,i}$  denote the error. We further let  $e_c = u_c - U_c$  and  $e_{f,i} = u_{f,i} - U_{f,i}$ .

- Energy norm estimate for primal solution,  $\|\nabla e\|$ , in the case when  $a = 1$ .
- Application on the dual problem.

# Standard Energy Norm Estimate

Weak form for the exact solution,

$$(\nabla u, \nabla v) = (f, v), \quad \text{for all } v \in H_0^1(\Omega).$$

Weak form for the finite element solution,

$$(\nabla U, \nabla v) = (f, v), \quad \text{for all } v \in V_h.$$

This gives us the error equation,

$$(\nabla e, \nabla v) = 0, \quad \text{for all } v \in V_h.$$

# Standard Energy Norm Estimate

Standard energy estimate

$$\begin{aligned}\|\nabla e\|^2 &= (\nabla e, \nabla e) = (\nabla e, \nabla(e - \pi e)) \\ &= (-\Delta e, e - \pi e) \leq C \|hR(U)\| \|\nabla e\|,\end{aligned}$$

where  $\pi e \in V_h$ . We get

$$\|\nabla e\| \leq C \|hR(U)\|.$$

# Energy Norm Estimate

$$\begin{aligned} \|\nabla e\| \leq & C \|hR(U_c + U_f)\| + C \sum_{\text{coarse}} \|HR(U_c)\|_{S_i^1} \\ & + C\sqrt{H} \sum_{\text{fine}} \|\Sigma_i\|_{\partial\omega_i} \end{aligned}$$

- The first term is referred to as the truth mesh error (reference).
- The third term is the normal derivative of the fine scale solutions on  $\partial\omega_i$ .

# Dual Problem

The standard approach to get a bound of a linear functional of the error is to introduce a dual problem: find  $\phi \in H_0^1$  such that

$$-\Delta\phi = \psi.$$

We then get for  $\pi\phi \in V_h$ ,

$$(e, \psi) = (e, -\Delta\phi) = (\nabla e, \nabla\phi) = (\nabla e, \nabla\phi - \pi\phi).$$

And after integration by parts we get

$$(e, \psi) = (R(U), \phi - \pi\phi).$$

# Dual Problem

- The dual solution  $\phi$  need to be approximated but not in  $V$ .
- Regular refinement or higher order method allocate lots of memory.

Instead we solve the dual problem by local problems in each coarse node,

$$(e, \psi) = \sum_{i=1}^n (R(U), \Phi_{f,i}) + (R(U), \phi_f - \Phi_f).$$

# Dual Problem

The second term can be estimated in the following way,

$$\begin{aligned}(\nabla e, \nabla(\phi_f - \Phi_f)) &\leq \|\nabla e\| \|\nabla(\phi_f - \Phi_f)\| \\ &\leq \|\nabla e\| \|\nabla(\phi - (\Phi_c + \Phi_f))\|.\end{aligned}$$

And we get the energy norm of the error in the dual solution which can be estimated.



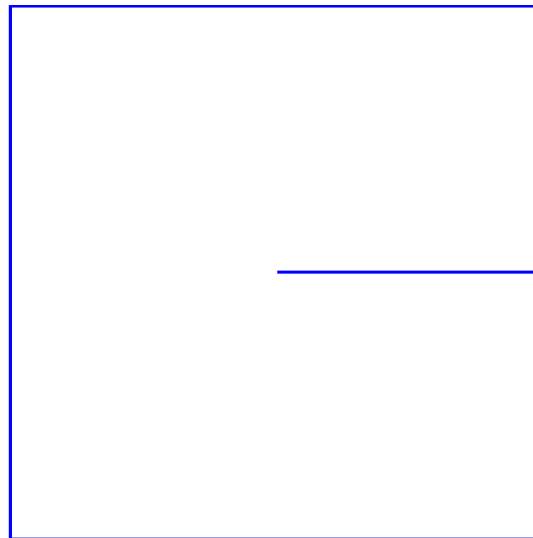
# Adaptive Strategy

$$\begin{aligned} \|\nabla e\| \leq & C \|hR(U_c + U_f)\| + C \sum_{\text{coarse}} \|HR(U_c)\|_{S_i^1} \\ & + C\sqrt{H} \sum_{\text{fine}} \|\Sigma_i\|_{\partial\omega_i} \end{aligned}$$

- We focus on the last two terms.
- We calculate these for each  $i \in \{\text{coarse fine}\}$ .
- Big values  $i \in \text{coarse} \rightarrow$  more local problems.
- Big values  $i \in \text{fine} \rightarrow$  more layers.

# Numerical Examples

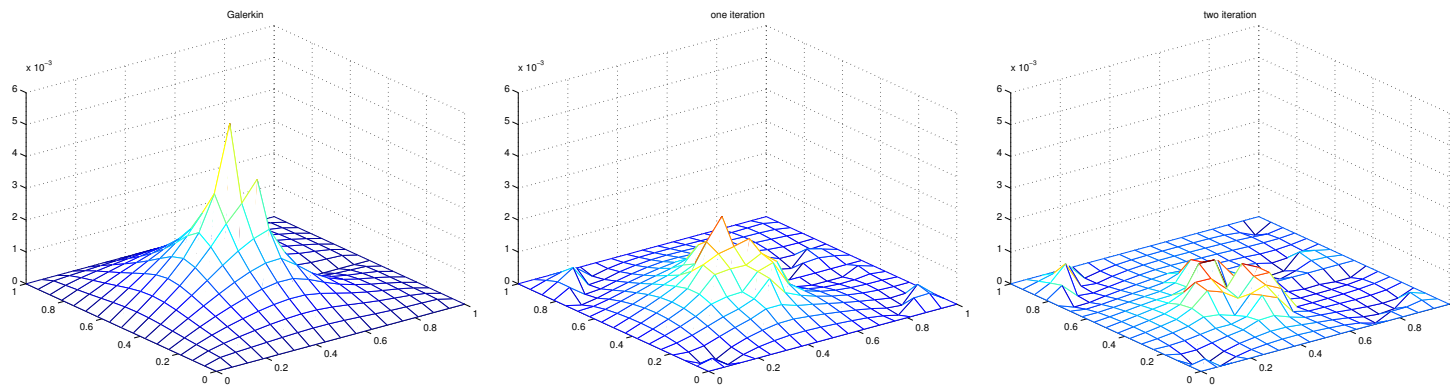
We start with a unit square containing a crack.



We let the coefficient  $a = 1$  and solve,  $-\Delta u = f$  with  $u = 0$  on the boundary including the crack.

# Numerical Examples

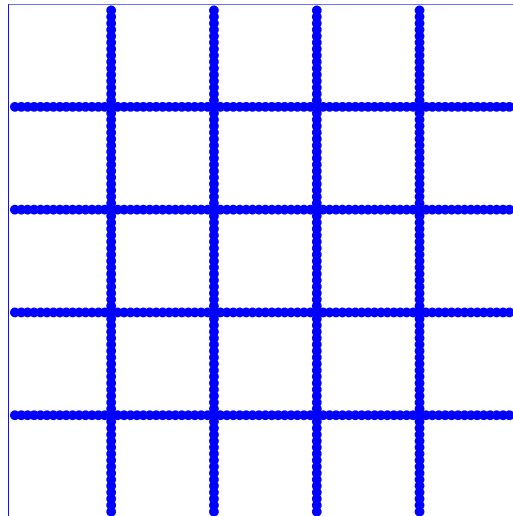
We solve the problem by using the adaptive algorithm with a refinement level of 10 % each iteration.



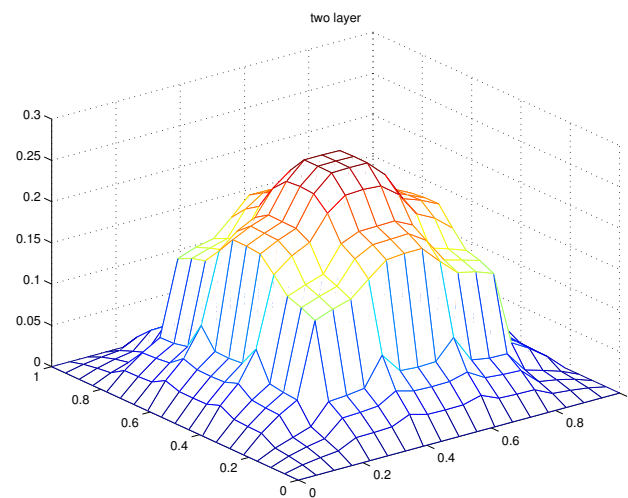
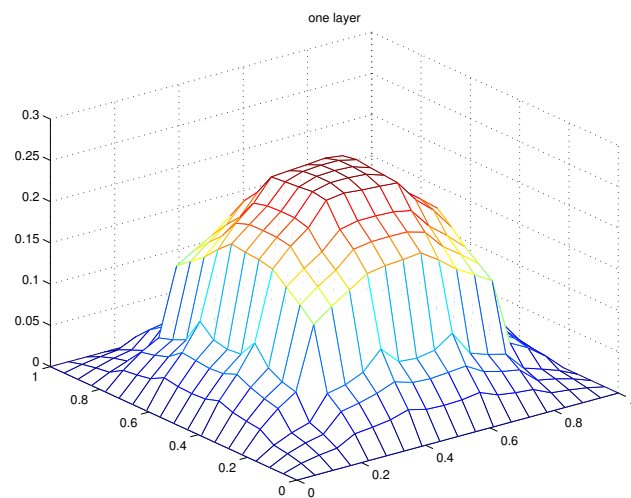
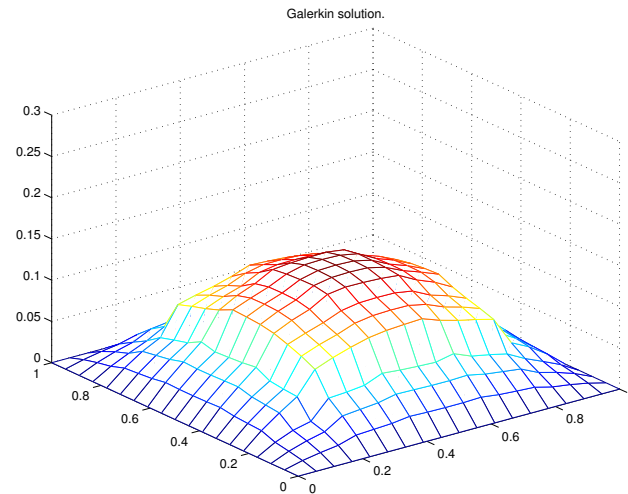
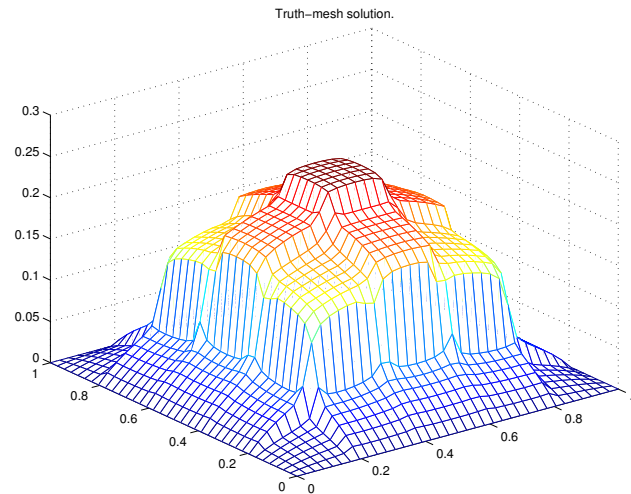
We plot the difference between our solution and a reference solution.

# Numerical Examples

In this example we study a discontinuous coefficient  $a$  in  $-\nabla \cdot a \nabla u = f$ .  $a = 1$  (white) and  $a = 0.05$  (blue).



# Numerical Examples



# Future Work

- Error estimates in the case when  $a \neq 1$ .
- Extended numerical tests in both 2D and 3D.
- More scales.
- Other equations (convection-diffusion, ...).
- Comparing results with classical Homogenization theory.

# References

## References

- [1] B. Aksoylu and M. Holst *An odyssey into local refinement and multilevel preconditioning II: stabilizing hierarchical basis methods*, SIAM J. Numer. Anal. in review
- [2] T. J.R. Hughes, *Multiscale phenomena: Green's functions, the Dirichlet-to-Neumann formulation ...* Comput. Methods Appl. Mech. Engrg. 127 (1995) 387-401.