

#### **Optimized Schwarz Methods for Problems with Discontinuous Coefficients**

#### **Olivier Dubois**

dubois@math.mcgill.ca

Department of Mathematics and Statistics

McGill University

http://www.math.mcgill.ca/~dubois



#### Motivation

Flow in heterogeneous media has many applications, for example

- oil recovery,
- earthquake prediction,
- underground disposal of nuclear waste.



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 $\Rightarrow$  This suggests a natural nonoverlapping domain decomposition.



#### Outline

- 1. Motivation
- 2. Introduction to the model problem
- 3. Schwarz iteration and optimal operators
- 4. Optimized transmission conditions
  - (a) one-sided Robin conditions (two versions)
  - (b) two-sided Robin conditions
  - (c) second order conditions
- 5. Asymptotic performance for strong heterogeneity
- 6. Numerical experiments
- 7. Conclusions and work in progress



#### **Some recent work**

- Y. Maday and F. Magoulès. Multilevel optimized Schwarz methods without overlap for highly heterogeneous media. Research report at Laboratoire Jacques-Louis Lions, 2005.
- Y. Maday and F. Magoulès. Improved ad hoc interface conditions for Schwarz solution procedure tuned to highly heterogeneous media. *Applied Mathematical Modelling*, 30(8):731-743, 2006.



#### The model problem

We consider a simple diffusion problem with a discontinuous coefficient

$$\begin{cases} -\nabla \cdot (\nu(x)\nabla u) = f \text{ on } \mathbb{R}^2, \\ u \text{ is bounded at infinity.} \end{cases}$$
(P)  
$$\overbrace{\qquad \Omega_1 \\ \nu(x) = \nu_1 \\ x = 0 \end{cases}$$

# A general Schwarz iteration

The solution to problem (P) satisfies the matching conditions

$$u(0^-, y) = u(0^+, y), \quad \nu_1 \frac{\partial u}{\partial n}(0^-, y) = \nu_2 \frac{\partial u}{\partial n}(0^+, y).$$

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Consider a general Schwarz iteration of the form

$$\begin{cases} -\nabla \cdot (\nu_1 \nabla u_1^{n+1}) &= f & \text{on } \Omega_1 = (-\infty, 0) \times \mathbb{R}, \\ \left(\nu_1 \frac{\partial}{\partial x} + S_1\right) u_1^{n+1} &= \left(\nu_2 \frac{\partial}{\partial x} + S_1\right) u_2^n & \text{at } x = 0, \end{cases}$$
$$\begin{cases} -\nabla \cdot (\nu_2 \nabla u_2^{n+1}) &= f & \text{on } \Omega_2 = (0, \infty) \times \mathbb{R}, \\ \left(-\nu_2 \frac{\partial}{\partial x} + S_2\right) u_2^{n+1} &= \left(-\nu_1 \frac{\partial}{\partial x} + S_2\right) u_1^n & \text{at } x = 0. \end{cases}$$

- $u_i^n = approximate solution in subdomain <math>\Omega_i$ , at iteration n.
- $\boldsymbol{\mathcal{S}}_i$  are linear boundary operators acting in y only



Fourier transform in *y*:

$$\mathcal{F}_y[u(x,y)] = \hat{u}(x,k) := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} u(x,y) e^{-iyk} dy$$



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Fourier symbols for the transmission operators  $S_i$ :

$$\mathcal{F}_y[\mathcal{S}_i u(x,y)] = \sigma_i(k)\hat{u}(x,k), \quad \text{for } i = 1, 2.$$



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Convergence factor of the Schwarz iteration in Fourier space:

$$\rho(k,\sigma_1,\sigma_2) := \left| \frac{\hat{u}_i^{n+1}(0,k)}{\hat{u}_i^{n-1}(0,k)} \right| = \left| \frac{(\sigma_1 - \nu_2|k|)(\sigma_2 - \nu_1|k|)}{(\sigma_1 + \nu_1|k|)(\sigma_2 + \nu_2|k|)} \right|.$$



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Optimal choice of operators:

$$\sigma_1^{opt}(k) = \nu_2 |k|, \quad \sigma_2^{opt}(k) = \nu_1 |k|.$$

# **Optimized Schwarz methods**

Find the "best" transmission conditions from a class of local operators C,

$$\min_{\sigma_1,\sigma_2\in\mathcal{C}} \left( \max_{k_1\leq k\leq k_2} \rho(k,\sigma_1,\sigma_2) \right).$$

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Equioscillation principle: often, the solution of this min-max problem is characterized by equioscillation of the convergence factor  $\rho$  at the local maxima, e.g.



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Convergence factor:

$$\rho(k, \mathbf{p}) = \left| \frac{(\mathbf{p} - \nu_1 |k|)(\mathbf{p} - \nu_2 |k|)}{(\mathbf{p} + \nu_1 |k|)(\mathbf{p} + \nu_2 |k|)} \right|$$

Uniform minimization of the convergence factor:

$$\min_{\boldsymbol{p}\in\mathbb{R}} \left( \max_{k_1 \le k \le k_2} \rho(k, \boldsymbol{p}) \right). \tag{M_1}$$

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We will state our result in terms of

$$\mu := \frac{\max(\nu_1, \nu_2)}{\min(\nu_1, \nu_2)}, \quad k_r = \frac{k_2}{k_1}$$

Theorem 1. Solution of the min-max problem  $(M_1)$ . Let

$$f(\mu) := \frac{(\mu+1)^2 + (\mu-1)\sqrt{\mu^2 + 6\mu + 1}}{4\mu}.$$

If  $k_r \ge f(\mu)$ , then one minimizer of  $(M_1)$  is  $p^* = \sqrt{\nu_1 \nu_2 k_1 k_2}$ . This minimizer  $p^*$  is unique when

$$\rho(k_1, \boldsymbol{p^*}) \ge \rho\left(\frac{\boldsymbol{p^*}}{\sqrt{\nu_1\nu_2}}, \boldsymbol{p^*}\right).$$

Otherwise, the minimum is also attained for any p chosen in some closed interval containing  $p^*$ .

If  $k_r < f(\mu)$ , then there are two minimizers given by the two positive roots of

$$p^{4} + \left[\nu_{1}\nu_{2}(k_{1}^{2} + k_{2}^{2}) - k_{1}k_{2}(\nu_{1} + \nu_{2})^{2}\right]p^{2} + (\nu_{1}\nu_{2}k_{1}k_{2})^{2}.$$

Both of these two values yield equioscillation, i.e.  $\rho(k_1, p^*) = \rho(k_2, p^*)$ .



Recall the optimal symbols are

$$\sigma_1^{opt}(k) = \nu_2 |k|, \quad \sigma_2^{opt}(k) = \nu_1 |k|.$$

This suggests a different scaling in the Robin conditions,

$$\sigma_1(k) = \nu_2 p, \quad \sigma_2(k) = \nu_1 p, \quad \text{for } p \in \mathbb{R}.$$

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Convergence factor:

$$\rho(k, p) = \frac{(p-k)^2}{(p+\mu k)(p+k/\mu)}.$$

Uniform minimization of the convergence factor:

$$\min_{\boldsymbol{p}\in\mathbb{R}} \left( \max_{k_1 \leq k \leq k_2} \rho(k, \boldsymbol{p}) \right).$$

 $(M_2)$ 

Theorem 2. Solution of the min-max problem  $(M_2)$ .

The optimization problem  $\left( M_{2} 
ight)$  has a unique minimizer, given by

$$p^* = \sqrt{k_1 k_2}.$$

This value always gives the equioscillation  $\rho(k_1, p^*) = \rho(k_2, p^*)$ .

#### **Optimized Robin conditions**

Comparison of optimized convergence factors for version 1 and 2:



 $\mu = 10$ 

 $\mu = 100$ 

## **Asymptotic performance**

**Theorem 3.** When  $k_2 = \frac{\pi}{h}$ , and as  $h \to 0$  (keeping  $\nu_1$  and  $\nu_2$  constant), we find the asymptotic expansions:

*Optimized Robin conditions, v. 1* 

$$\max_{k_1 \le k \le \pi/h} |\rho(k, \mathbf{p}^*)| = 1 - 2\left(\sqrt{\mu} + \frac{1}{\sqrt{\mu}}\right) \left(\frac{k_1 h}{\pi}\right)^{\frac{1}{2}} + O(h)$$

Deptimized Robin conditions, v. 2

$$\max_{k_1 \le k \le \pi/h} |\rho(k, \mathbf{p}^*)| = 1 - \frac{(\mu + 1)^2}{\mu} \left(\frac{k_1 h}{\pi}\right)^{\frac{1}{2}} + O(h)$$

We now consider two-sided Robin conditions, with two free parameters

$$\sigma_1(k) = \nu_2 p, \quad \sigma_2(k) = \nu_1 q, \quad \text{for } p, q \in \mathbb{R}.$$

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Convergence factor:

$$\rho(k, \boldsymbol{p}, \boldsymbol{q}) = \left| \frac{(\boldsymbol{p} - k)(\boldsymbol{q} - k)}{(\boldsymbol{p} + \frac{\nu_1}{\nu_2}k)(\boldsymbol{q} + \frac{\nu_2}{\nu_1}k)} \right|$$

Uniform minimization of the convergence factor:

$$\min_{\boldsymbol{p},\boldsymbol{q}\in\mathbb{R}} \left( \max_{k_1 \le k \le k_2} \rho(k,\boldsymbol{p},\boldsymbol{q}) \right). \tag{M_3}$$

Let 
$$\lambda := \frac{\nu_1}{\nu_2}$$

#### Theorem 4. Solution of the min-max problem $(M_3)$ .

When  $\lambda ≥ 1$ , the unique minimizing pair  $(p^*, q^*)$  of problem  $(M_3)$  is the unique solution to the system of equations

$$p^*q^* = k_1k_2, (E_1)$$

$$|\rho(k_1, p^*, q^*)| = |\rho(\sqrt{p^*q^*}, p^*, q^*)|, \qquad (E_2)$$

*satisfying*  $k_1 < p^* < q^*$ *.* 

■ When  $\lambda \leq 1$ , the unique minimizing pair  $(p^*, q^*)$  of  $(M_3)$  is the solution of the above equations  $(E_1)$ - $(E_2)$  satisfying  $k_1 < q^* < p^*$  instead.

The optimized convergence factor always has the equioscillation property at the frequencies  $k_1$ ,  $k_2$  and  $k_c = \sqrt{pq}$ .

Computing  $p^*$  reduces to finding the unique real root of the quartic

$$(\mathbf{p} + \lambda k_1)(\mathbf{p} + \lambda k_2)(\sqrt{k_1k_2} - \mathbf{p})^2 - (\mathbf{p} - k_1)(k_2 - \mathbf{p})(\mathbf{p} + \lambda\sqrt{k_1k_2})^2$$

in the interval  $(k_1, \sqrt{k_1k_2})$  if  $\lambda \ge 1$ , or in the interval  $(\sqrt{k_1, k_2}, k_2)$  if  $\lambda \le 1$ .

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in the interval  $(k_1, \sqrt{k_1k_2})$  if  $\lambda \ge 1$ , or in the interval  $(\sqrt{k_1, k_2}, k_2)$  if  $\lambda \le 1$ .

**Theorem 5.** When  $k_2 = \frac{\pi}{h}$ , and as h approaches 0, the asymptotic performance of the optimized two-sided Robin conditions is

$$\max_{k_1 \le k \le \pi/h} \rho(k, \boldsymbol{p^*}, \boldsymbol{q^*}) = \frac{1}{\mu} - \frac{4(\mu+1)}{\mu(\mu-1)} \sqrt{\frac{k_1}{\pi}} h^{\frac{1}{2}} + O(h).$$

# **Optimized 2nd order conditions**

We can also consider transmission conditions that are second order in the tangential direction

$$\begin{bmatrix} \nu_1 \frac{\partial}{\partial x} + \nu_2 \left( p - q \frac{\partial^2}{\partial y^2} \right) \end{bmatrix} u_1^{n+1} = \begin{bmatrix} \nu_2 \frac{\partial}{\partial x} + \nu_2 \left( p - q \frac{\partial^2}{\partial y^2} \right) \end{bmatrix} u_2^n,$$
$$\begin{bmatrix} -\nu_2 \frac{\partial}{\partial x} + \nu_1 \left( p - q \frac{\partial^2}{\partial y^2} \right) \end{bmatrix} u_2^{n+1} = \begin{bmatrix} -\nu_1 \frac{\partial}{\partial x} + \nu_1 \left( p - q \frac{\partial^2}{\partial y^2} \right) \end{bmatrix} u_1^n.$$

The corresponding symbols in Fourier space are

$$\sigma_1(k) = \nu_2(\mathbf{p} + \mathbf{q}k^2), \quad \sigma_2(k) = \nu_1(\mathbf{p} + \mathbf{q}k^2).$$

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$$\sigma_1(k) = \nu_2(\mathbf{p} + \mathbf{q}k^2), \quad \sigma_2(k) = \nu_1(\mathbf{p} + \mathbf{q}k^2).$$

This leads to the min-max problem

$$\min_{\mathbf{p},\mathbf{q}\in\mathbb{R}} \left( \max_{k_1 \le k \le k_2} \left| \frac{(\mathbf{p} + \mathbf{q}k^2 - k)^2}{(\mathbf{p} + \mathbf{q}k^2 + \mu k)(\mathbf{p} + \mathbf{q}k^2 + k/\mu)} \right| \right). \tag{M_4}$$

# Optimized 2nd order conditions

Conjecture 1. Solution of the min-max problem  $(M_4)$ .

The min-max problem  $(M_4)$  has a unique solution, given by the equioscillation of the convergence factor at the frequencies  $k_1$ ,  $k_2$  and  $k_c = \sqrt{\frac{p}{q}}$ . This gives the formulas

$$p^* = \frac{(k_1k_2)^{\frac{3}{4}}}{\sqrt{2(k_1+k_2)}}, \quad q^* = \frac{1}{\sqrt{2(k_1+k_2)(k_1k_2)^{\frac{1}{4}}}}$$

When  $k_2 = \frac{\pi}{h}$  and h tends to 0, we find the asymptotic performance

$$\max_{k_1 \le k \le \pi/h} \rho(k, \boldsymbol{p^*}, \boldsymbol{q^*}) = 1 - \sqrt{2} \left( 2 + \mu + \frac{1}{\mu} \right) \left( \frac{k_1}{\pi} \right)^{\frac{1}{4}} h^{\frac{1}{4}} + O(h^{\frac{1}{2}}).$$

#### **Comparison of convergence factors**

 $\mu = 100$  $\mu = 1000$ 0.18 0.02 Opt. Robin v.2 Opt. 2–sided Robin 0.16 0.018 Opt. 2nd order 0.016 0.14 0.014 0.12 0.012 0.1 ٩ 0.01 ρ 0.08 0.008 Opt. Robin v.2 0.06 Opt. 2-sided Robin 0.006 Opt. 2nd order 0.04 0.004 0.02 0.002 0<sup>L</sup> 0 0.5 0.7 0.5 0.1 0.2 0.3 0.4 0.6 0.8 0.9 0.1 0.2 0.3 0.4 0.6 0.7 0.8 0.9 k k

### **C** Asymptotics for strong heterogeneity

Let us now consider the case when h is small but held fixed, and  $\mu$  is large

$$\mu = \frac{\max(\nu_1, \nu_2)}{\min(\nu_1, \nu_2)} >> 1.$$

The asymptotics become

Optimized Robin conditions, v. 1

$$\max_{k_1 \le k \le \pi/h} |\rho(k, p^*)| = 1 - 2\sqrt{\frac{k_1 h}{\pi}} + \dots - O\left(\frac{h^{-\frac{1}{2}}}{\mu}\right).$$

Optimized Robin conditions, v. 2

$$\max_{k_1 \le k \le \pi/h} |\rho(k, \boldsymbol{p^*})| = \sqrt{\frac{\pi}{k_1 h}} \left(\frac{1}{\mu}\right) - O\left(\frac{1}{\mu}\right).$$

#### **Asymptotics for strong heterogeneity**

Optimized two-sided Robin conditions

$$\max_{k_1 \le k \le \pi/h} |\rho(k, p^*, q^*)| = \frac{1}{\mu} - O\left(\frac{h^{\frac{1}{2}}}{\mu}\right)$$

Optimized second order conditions

$$\max_{k_1 \le k \le \pi/h} |\rho(k, \boldsymbol{p^*}, \boldsymbol{q^*})| = \left(\frac{\pi}{4k_1h}\right)^{\frac{1}{4}} \left(\frac{1}{\mu}\right) - O\left(\frac{1}{\mu}\right).$$



#### **Numerical experiments**

$$\begin{cases} -\nabla \cdot (\nu(x)\nabla u(x)) &= 1 \quad \text{on } \Omega = (0,\pi)^2, \\ u &= 0 \quad \text{on } \partial \Omega. \end{cases}$$

The domain is divided into two symmetric subdomains and a finite volume discretization is used. For  $h = \frac{\pi}{300}$ :





### **Numerical experiments**

Here, we take  $\mu = 10$  and vary the grid size *h*. The table shows the number of iterations to reach a tolerance of  $10^{-6}$ .

h	Opt. Robin v.1	Opt. Robin v.2	Opt. 2-sided Robin
$\frac{\pi}{50}$	24	24	12
$\frac{\pi}{100}$	30	26	12
$\frac{\pi}{200}$	54	42	14
$\frac{\pi}{300}$	62	48	14



#### **Numerical experiments**

Now, let us fix  $h = \frac{\pi}{300}$  and vary the heterogeneity ratio  $\mu$ .

$\mu$	Opt. Robin v.1	Opt. Robin v.2	Opt. 2-sided Robin
$10^{1}$	62	48	14
$10^{2}$	72	16	10
$10^{3}$	180	10	8
$10^{4}$	204	8	6
$10^{5}$	202	6	6



#### Conclusions

We completely solved the min-max problem for

- optimized one-sided Robin conditions (2 versions),
- optimized two-sided Robin conditions.
- For one-sided Robin conditions, we showed that the second version leads to much better performance, particularly when the jump in the diffusion coefficient is large.
- For two-sided Robin conditions, we obtain an optimal asymptotic performance of  $\rho = O(1/\mu)$ , independent of *h*.
- For almost all optimized transmission conditions we considered, the convergence is improved as we increase the jump in the diffusion coefficient.



## Work in progress

- Design of an efficient coarse space correction when using many subdomains.
- Optimized conditions for the advection-diffusion equation in 2D,

$$-\nabla \cdot (\nu(x)\nabla u(x)) + \mathbf{a}(x) \cdot \nabla u(x) + c(x)u(x) = f(x),$$

with discontinuous coefficients.