Substructuring for multiscale problems

Clemens Pechstein



Johannes Kepler University

Linz (A)



jointly with

Rob Scheichl

Marcus Sarkis

Clark Dohrmann

DD 21, Rennes, June 2012



- 2 Weighted Poincaré Inequalities (WPI)
- 3 FETI Basics
- 4 Multiscale Toolkit





1 Introduction

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- 5 TFETI



Model Problem

Find
$$u \in H^1(\Omega), \ u_{|\Gamma_D} = 0$$
$$\int_{\Omega} \alpha \nabla u \cdot \nabla v \ dx = f($$

strongly varying coefficient

 $oldsymbol{lpha} \in L^\infty_+(\Omega)$ (uniformly positive)

Conditioning of FE system:

~
$$\sup_{\substack{x,y\in\Omega\\\mathcal{O}(10^3)-\mathcal{O}(10^{10})}} h^{-2}$$

Goal:

iterative solvers, robust in h and α

 $\Omega \subset \mathbb{R}^d \quad d = 2, 3$

 $(v) \quad \forall v \in H^1(\Omega), \ v_{|\Gamma_D} = 0$

towards applications:

porous media flows oil reservoir simulation elasticitiy with heterogeneous material nonlinear magnetostatics



Spectral properties – I

unit square uniform mesh h = 1/32pure Neumann problem

 $\sigma(K_{h,\alpha})$





spectrum many_large_islands.dat -- H/h=32 B=I



spectrum many_large_islands_qm.dat -- H/h=32 B=I



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Spectral properties – II

unit square uniform mesh h = 1/32pure Neumann problem

$$\sigma(\operatorname{diag}(K_{h,\alpha})^{-1}K_{h,\alpha})$$





spectrum many_large_islands.dat -- H/h=32 B=Kdiag



spectrum many_large_islands_qm.dat -- H/h=32 B=Kdiag



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Spectral properties – III

unit square uniform mesh h = 1/32pure Neumann problem

 $\sigma(M_{h,\alpha}^{-1}K_{h,\alpha})$





spectrum many_large_islands.dat -- H/h=32 B=Malpha







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Substructuring for multiscale problems

Weighted Poincaré Inequality (WPI)

domain D, coefficient $\alpha \in L^{\infty}_{+}(D)$

Definition

 $C_{P,\alpha}(D)$ smallest constant such that $\forall u \in H^1(D)$:

$$\inf_{\boldsymbol{c}\in\mathbb{R}} \|\boldsymbol{u}-\boldsymbol{c}\|_{L^{2}(D),\boldsymbol{\alpha}} \leq \underbrace{\underbrace{C_{\boldsymbol{P},\boldsymbol{\alpha}}(\boldsymbol{D})\operatorname{diam}(\boldsymbol{D})}_{\approx \lambda_{2}(\boldsymbol{M}_{h,\boldsymbol{\alpha}}^{-1}\boldsymbol{K}_{h,\boldsymbol{\alpha}})^{-1/2}} |\boldsymbol{u}|_{H^{1}(D),\boldsymbol{\alpha}}$$

where [Chua] [Chua & Wheeden] [Veeser & Verfürth] [Zikhov] ... $\|v\|_{L^2(D),\alpha}^2 = \int_D \alpha |v|^2 dx, \qquad |v|_{H^1(D),\alpha}^2 = \int_D \alpha |\nabla v|^2 dx$ infimum attained at weighted average: $c = \overline{u}^{D,\alpha} = \frac{\int_D \alpha \, u \, dx}{\int_D \alpha \, dx}$

For quasi-monotone coefficients (details soon):

$C_{P, \alpha}(D) \leq C$ independently of contrast

P. & Scheichl '11, '12*] early results by [Galvis & Efendiev '10]

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Preconditioners for multiscale problems



Minisymposium M13 (Tu 16.30, We 10.30) links to upscaling

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Quasi-monotone coefficients

Let $\alpha \in L^{\infty}_{+}(D)$ be **piecewise constant** w.r.t. partition

 $\{Y_k\}_{k=1}^n$ (*Y_k* connected Lipschitz) $\alpha_k := \alpha_{|Y_k|} = \text{const}$

Y₁... subregion with largest coefficient

Definition

α is quasi-monotone on D iff

for each k there exists a **path** P_{kL} from Y_k to Y_L :

- subsequent subregions share common (d-1)-dim. facet
- coefficient does not decrease along the path



"path inequalities ~> weighted inequality"

α quasi-monotone P_{kL} corresp. path from Y_k to Y_L $X^* ⊂ \overline{Y}_L$

Definition

ckL best constant:

$$\|u - \overline{u}^{X^*}\|_{L^2(Y_k)}^2 \leq c_{kL} \operatorname{diam}(D)^2 |u|_{H^1(P_{kL})}^2 \qquad \forall u \in H^1(P_{kL})$$

Lemma

$$\|u - \overline{u}^{X^*}\|_{L^2(D),\alpha}^2 \leq \left(\sum_{k=1}^n c_{kL}\right) \operatorname{diam}(D)^2 |u|_{H^1(D),\alpha}^2 \quad \forall u \in H^1(D)$$

$$\implies C_{P,\alpha}(D)^2 \leq \left(\sum_{k=1}^n c_{kL}\right) \quad independent \ of \ constrast!$$

Explicit dependence on geometric scales



Lemma

 $\begin{aligned} P_{kL} \dots \text{ path from } Y_k \text{ to } Y_L & X_i \dots \text{ interfaces or } X^* \\ c_{kL} \leq 2 \sum_{Y_\ell \subset P_{kL}} \sum_{X_i \subset \partial Y_\ell} \frac{|Y_k|}{|Y_\ell|} \frac{\text{diam}(Y_\ell)^2}{\text{diam}(D)^2} \, C_P(Y_\ell, X_i) \end{aligned}$

Lemma

 $\begin{array}{l} \substack{\boldsymbol{\alpha} \text{ quasi-monotone on } D \\ \{Y_{\ell}\} \text{ form shape regular partition, not extremely long paths} \\ \\ \substack{\boldsymbol{\mathcal{C}_{P, \boldsymbol{\alpha}}(D)^2 \\ \lesssim s_{\max} \frac{\operatorname{meas}(D)}{\operatorname{diam}(D)^2 \eta_{\min}^{d-2}} \\ \end{array} \lesssim \left(\frac{\operatorname{diam}(D)}{\eta_{\min}}\right)^{d-1} \end{array}$

WPI for FE functions

Definition

 α type-*m* quasi-monotone on *D* iff for each *k* there exists a path P_{kL} from Y_k to Y_L :

- subsequent subregions share common *m*-dim. facet
- weights within the path do not decrease



Analogous statements hold true on FE space $V^h(D)$, but constants depend on *h*:

m = d - 2: $1 + \log(\eta/h)$ More details:m = d - 3: η/h [P. & Scheichl: IMAJNA 2012 (to appear)]

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FETI algorithms – I

Finite Element Tearing and Interconnecting

FETI: [Farhat & Roux] FETI-DP: [Farhat, Lesoinne, Le Tallec, Pierson, Rixen] TFETI: [Dostál, Horák, Kučera], [Of, Steinbach]





FETI algorithms – II



Definition of jump operators

jump operators:

B: W ightarrow U	Wtorn FE space
$B_D: W \to U$	ULagrange multiplier space

 x^h ... interface DOF (node) on $\partial \Omega_i \cap \partial \Omega_k$:

$$(B w)_{ik}(x^h) = w_i(x^h) - w_k(x^h)$$



Definition of jump operators

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B: W ightarrow U	Wtorn FE space
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 x^h ... interface DOF (node) on $\partial \Omega_i \cap \partial \Omega_k$:

$$(\boldsymbol{B}_{D}\boldsymbol{w})_{ik}(\boldsymbol{x}^{h}) = \frac{1}{\sum\limits_{j \in \mathcal{N}_{\boldsymbol{x}^{h}}} \rho_{j}(\boldsymbol{x}^{h})} \left[\rho_{k}(\boldsymbol{x}^{h}) \, \boldsymbol{w}_{i}(\boldsymbol{x}^{h}) - \rho_{i}(\boldsymbol{x}^{h}) \, \boldsymbol{w}_{k}(\boldsymbol{x}^{h}) \right]$$

scalings $\rho_j(x^h)$

possible choices discussed later on

- per subdomain j
- per interface DOF x^h

partition of unity property averaging property: $I - B_D^{\top}B = E_D : W \to \widehat{W}$

Analysis framework

For FETI, FETI-DP (also BDD, BDDC):

• $\lambda_{\min} \geq 1$ • $\lambda_{\max} \leq \sup_{w \in W_{sub}} \frac{\frac{P_D}{|B_D^\top B w|_S^2}}{|w|_S^2}$

 $W_{sub} \subset W$ suitable

 $|\cdot|_{S}$...Schur complement norm on W

Common technique:

- splitting into subdomain face, edge, vertex contributions using cut-off functions
- transfer operators (Sobolev extension + Scott-Zhang)
- Poincaré inequality

[Mandel & Tezaur] [Klawonn & Widlund] [Klawonn, Widlund, Dryja] [Dohrmann, Mandel, Tezaur] [Mandel & Sousedik] [Klawonn, Rheinbach, Widlund]

Theory vs. Implementation



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Boundary layers & patch decompositions

subdomain Ω_i



Assumptions througout:

- patch decomposition \rightsquigarrow globally conforming mesh \mathcal{T}^η
- patch decomposition $\mathcal{T}^{\eta}(\Omega_{i,\eta})$ quasi-uniform
- variation on single patch $\leq c_{noise}$ small

Boundary layers & patch decompositions

boundary layer $\Omega_{i,\eta}$



Assumptions througout:

- ullet patch decomposition \leadsto globally conforming mesh \mathcal{T}^η
- patch decomposition $\mathcal{T}^{\eta}(\Omega_{i,\eta})$ quasi-uniform
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Boundary layers & patch decompositions



Assumptions througout:

- patch decomposition \rightsquigarrow globally conforming mesh \mathcal{T}^η
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Need good scalings

$\rho_j(x^h)$	Theory	Implementation
(a)	1	1 (multiplicity scaling)
(b)	$lpha_{\Omega_j}^{\sf max}$	$\max(K_j^{\text{diag}})$
(C)	$\max_{\tau \subset \Omega_j: x^h \in \overline{\tau}} \alpha_{ \tau}$	$K_j^{\text{diag}}(x^h)$ (stiffness scaling)
(d)	$\max_{\substack{Y_j^{(k)}: x^h \in \overline{Y}_j^{(k)}}} \alpha_{Y_j^{(k)}}^{\max}$	processed stiffness scaling ?
(e)	new promising te	echnique, C. Dohrmann's talk, Mo, M13

Choice (a): does not work for certain jumps across interfaces

Choice (b): κ may depend on $\max_{i} \frac{\alpha_{\Omega_{i}}^{\max}}{\alpha_{\Omega_{i}}^{\min}}$

Choice (c): κ may depend on oscillations of α or deteriorate for ragged interfaces!

Intermediate result

Lemma (P. & Scheichl '11, P. '12*)

For (theoretical) choice (d), $\rho_j(x^h) = \max_{\substack{Y_i^{(k)}: x^h \in \overline{Y}_i^{(k)}}} \alpha_{Y_i^{(k)}}^{\max}$

 $\forall w \in W$:

$$\begin{split} | \mathbf{P}_{\mathbf{D}} \mathbf{w} |_{\mathbf{S}}^{2} &\leq \mathbf{C} \, \mathbf{C}_{\textit{noise}} \, \sum_{j=1}^{N} \left[(1 + \log(\frac{\eta}{h}))^{2} \, |\mathbf{w}_{j}|_{H^{1}(\Omega_{j,\eta}), \mathbf{\alpha}}^{2} \right. \\ &+ \left(1 + \log(\frac{\eta}{h}) \right) \, \eta^{-2} \, \|\mathbf{w}_{j}\|_{L^{2}(\Omega_{j,\eta}), \mathbf{\alpha}}^{2} \end{split}$$

Proof:

- finer splitting into patch globs
- transfer operators (carefully!)
- conventional cut-off estimates

(d) difficult to mimic in practice (edge detection, TV minimization)



Assumption:

- "nice" subdomains Ω_i
- paths not extremely long

Then:

•
$$C_{P,\alpha}(\Omega_j)^2 \lesssim \left(\frac{H}{\eta}\right)^{d-1}$$

• $C_{P,\alpha}(\Omega_{j,\eta})^2 \lesssim \left(\frac{H}{\eta}\right)^{d-2}$

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Condition number bounds – I

TFETI,
$$\mathbf{Q} = \mathbf{B}_D \, \mathbf{S} \, \mathbf{B}_D^{\top}$$

[P. & Scheichl '08, '11, P. '12]

Theorem

 α constant in boundary layers $\Omega_{i,\eta}$

$$\kappa \leq C c_{noise} \left(\frac{H}{\eta}\right)^2 \left(1 + \log\left(\frac{\eta}{h}\right)\right)^2$$

improves to H/η if α inside *larger* than in $\Omega_{i,\eta}$

Theorem

 $oldsymbol{lpha}$ quasi-monotone in boundary layers $\Omega_{i,\eta}$

$$\kappa \leq \boldsymbol{C} \boldsymbol{c}_{\textit{noise}} \left(\frac{H}{\eta} \right)^{d} \left(1 + \log \left(\frac{\eta}{h} \right) \right)^{2}$$

Similar: $oldsymbol{Q} = oldsymbol{Q}_{\mathsf{diag}}$; type-m quasi-monotone

Condition number bounds – I

TFETI,
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Similar: $Q = Q_{\text{diag}}$; type-*m* quasi-monotone

Condition number bounds – II

artificial coefficient α^{art} :

 $oldsymbol{lpha}^{\mathsf{art}} = oldsymbol{lpha} \quad ext{in } \Omega_{j,\eta} \ oldsymbol{lpha}^{\mathsf{art}} \leq oldsymbol{lpha} \quad ext{elsewhere}$

Enough to have

$$\inf_{\boldsymbol{c} \in \mathbb{R}} \|\boldsymbol{w} - \boldsymbol{c}\|_{L^2(\Omega_{j,\eta}),\boldsymbol{\alpha}}^2 \leq \frac{C_j^* H^2 \|\boldsymbol{w}\|_{H^1(\Omega_j),\boldsymbol{\alpha}}^2}{2}$$

Theorem

 α^{art} quasi-monotone on Ω_i :

$$\kappa \leq \boldsymbol{C} \boldsymbol{c}_{\textit{noise}} \left(\frac{H}{\eta} \right)^{d+1} \left(1 + \log \left(\frac{\eta}{h} \right) \right)^2$$

[P. & Scheichl '08,'11] [Dryja & Sarkis '11] [Gippert, Klawonn, Rheinbach '12*]





Q: is quasi-monotonicity is necessary for the robust of TFETI ?

Conjecture

TFETI is robust iff for each subdomain, there exists a quasi-monotone artificial coefficient.

Q: is quasi-monotonicity is necessary for the robust of TFETI ?

Conjecture

TFETI is robust iff for each subdomain, there exists a quasi-monotone artificial coefficient.

That's wrong!

TFETI has more robustness properties!

Inclusion features – I

Docking inclusions can be elminated!



Theoretical reason:

[work in progress]

We actually have to estimate

$$\int_{\Omega_{i,\eta}} \min(\boldsymbol{\alpha}_i, \, \boldsymbol{\alpha}_{\text{neighbors}}) \, |...|^2 \, dx$$

Inclusion features – II

Even docking channels can be elminated!



Theoretical reason:

[work in progress]

We actually have to estimate

$$\int_{\Omega_{i,\eta}} \min(\boldsymbol{\alpha}_i, \, \boldsymbol{\alpha}_{\text{neighbors}}) \, |...|^2 \, dx$$

Face inclusions can be elminated!



Theoretical reasons:

[work in progress]

Special cut-off function

[Graham, Lechner, Scheichl '07]

• Special transfer operator, L^2_{α} , H^1_{α} -stable for binary media

What's bad...

E.g. two vertex inclusions or long channels





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Choice of primal DOFs

Choice of primal DOFs should

- ensure that \widetilde{K} is invertible
- make WPI applicable (on neighborhoods of edges/faces)

With tools from above this will allow for robustness analysis

Adapted tool: WPI with weighted averages [P., Sarkis, Scheichl, DD20 proc

 $\|u - \overline{u}^{X,\widehat{lpha}}\|_{L^2(D), lpha} \ \le \ \mathcal{C}_{\mathcal{P}, lpha}(D, X, \widehat{lpha}) \operatorname{diam}(D) |u|_{H^1(D), lpha}$

Essential requirements:

- α quasi-monotone
- averaging manifold X must see largest coefficient
- $\widehat{\alpha}$ and α have to match

Geometry dependence can again be made explicit

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 $\|u - \overline{u}^{X,\widehat{\alpha}}\|_{L^{2}(D),\alpha} \leq C_{P,\alpha}(D,X,\widehat{\alpha}) \operatorname{diam}(D) |u|_{H^{1}(D),\alpha}$

Essential requirements:

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Geometry dependence can again be made explicit

Simple examples:



Primal DOF associated to edge E:

$$\overline{u}^{E,\widehat{\alpha}} := \frac{\int_{E} \widehat{\alpha} \, u \, ds}{\int_{E} \widehat{\alpha} \, ds} \quad \text{or} \quad \overline{u}^{E,\widehat{\alpha},\text{alg}} := \frac{\sum_{x^h \in \overline{E}} \widehat{\alpha}(x^h) \, u(x^h)}{\sum_{x^h \in \overline{E}} \widehat{\alpha}(x^h)}$$

[Klawonn & Rheinbach, 2006]

FETI-DP coarse space

What to do here?





Problems:

- Primal DOF should have same weight for both sides
- Usage of multiple primal DOFs \rightsquigarrow dead end?

Recall: for edge *E*, we actually have to bound

 $\min(\alpha_i, \alpha_k) |...|^2 dx$

If minimum coefficient has quasi-monotone extensions then

$$\widehat{\boldsymbol{\alpha}}(x^h) = \min(K_i^{\text{diag}}(x^h), K_k^{\text{diag}}(x^h))$$

(algebraic average) does the job, at least robust w.r.t. contrast!

FETI-DP coarse space

What to do here?





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Recall: for edge E, we actually have to bound

$$\prod_{\Omega_{i,\eta}\cap U_E} \min(\alpha_i, \alpha_k) |...|^2 dx$$

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Numerical test



FETI-DP, edge crosspoint example, H/h=32

Summary

In this talk:

- Weighted Poincaré Inequalities
- Robustness theory for TFETI
- Robustness theory for FETI-DP (yet to be completed)

Theory and testing guides to more insight / new methods!

Open problems:

- Choice of scalings
- Choice of primal constraints
- Incorporation of eigensolves

monograph: Finite and Boundary Element Tearing and Interconnecting Solvers for Multiscale Problems Springer LNCSE series, to appear

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Interconnecting...





THANKS FOR YOUR ATTENTION

Interconnecting...





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