

Espaces grossiers adaptatifs pour les méthodes de décomposition de domaine

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- 1 Recall on GenEO for SPD problems
- 2 Extension of GenEO to Saddle Point problem
- 3 Numerical Results for the Extension to Saddle Point
- 4 HPDDM and FreeFem DSL
- 5 Spectral Coarse Space for General Sparse problems

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(Recall) An introduction to DDM I

Consider the discretized Poisson problem: $Au = f \in \mathbb{R}^n$.

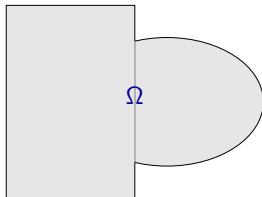
Given a decomposition of $[[1; n]]$, $(\mathcal{N}_1, \mathcal{N}_2)$, define:

- the restriction operator R_i from $\mathbb{R}^{[[1; n]]}$ into $\mathbb{R}^{\mathcal{N}_i}$,
- R_i^T as the extension by 0 from $\mathbb{R}^{\mathcal{N}_i}$ into $\mathbb{R}^{[[1; n]]}$.

$u^m \rightarrow u^{m+1}$ by solving concurrently:

$$u_1^{m+1} = u_1^m + A_1^{-1} R_1(f - Au^m) \quad u_2^{m+1} = u_2^m + A_2^{-1} R_2(f - Au^m)$$

where $u_i^m = R_i u^m$ and $A_i := R_i A R_i^T$.

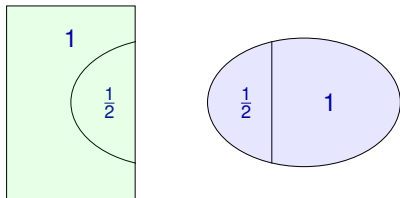


An introduction to DDM II

We have effectively *divided*, but we have yet to *conquer*.

Duplicated unknowns coupled via a *partition of unity*:

$$I = \sum_{i=1}^N R_i^T D_i R_i.$$



Then, $u^{m+1} = \sum_{i=1}^N R_i^T D_i u_i^{m+1}$.

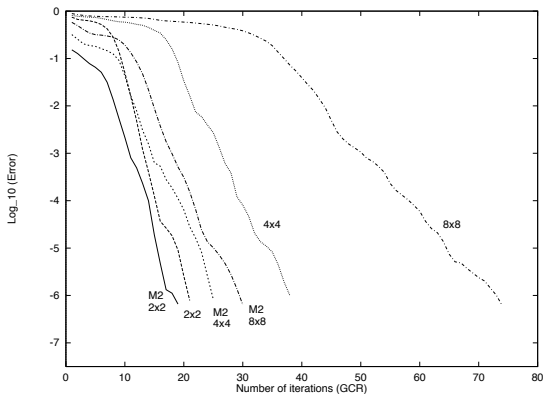
$$M_{RAS}^{-1} = \sum_{i=1}^N R_i^T D_i A_i^{-1} R_i$$

+ Krylov acceleration \Rightarrow RAS algorithm (Cai & Sarkis, 1999).

Symmetrized version $M_{ASM}^{-1} := \sum_{i=1}^N R_i^T A_i^{-1} R_i$ as a preconditioner.

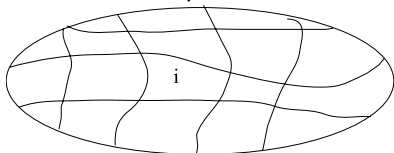
Convergence curves- more subdomains

Plateaus appear in the convergence of the Krylov methods.



Stagnation corresponds to a few very low eigenvalues in the spectrum of the preconditioned problem. They are due to the lack of a global exchange of information in the preconditioner.

$$\begin{aligned} -\Delta u &= f \text{ in } \Omega \\ u &= 0 \text{ on } \partial\Omega \end{aligned}$$



The mean value of the solution in domain i depends on the value of f on all subdomains.

How to achieve scalability

Introduction of a **coarse problem** that couples all subdomains. This is closely related to **deflation technique** classical in linear algebra (see Y. Saad, J. Erhel, Nabben and Vuik) and multigrid techniques.

Strongly related to Multiscale FEM and Reduced Order Model.

Adding a coarse space

One level methods are not scalable: $M_{ASM}^{-1} := \sum_{i=1}^N R_i^T A_i^{-1} R_i$.

We add a coarse space correction (*aka* second level). Let V_H be the coarse space and Z be a basis, $V_H = \text{span } Z$, we define the two level preconditioner as:

$$M_{ASM,2}^{-1} := Z (Z^T A Z)^{-1} Z^T + \sum_{i=1}^N R_i^T A_i^{-1} R_i.$$

For constant per subdomain coefficients, **Nicolaides approach** (1987) is to use the near-kernel of the local operators to build the coarse space:

$$Z := (R_i^T D_i R_i \mathbf{1})_{1 \leq i \leq N}.$$

Key notion: **Stable splitting** (J. Xu, 1989)

Great for (locally) smooth problems

Theoretical convergence result

Theorem (Widlund, Dryija)

Let $M_{ASM,2}^{-1}$ be the two-level additive Schwarz method:

$$\kappa(M_{ASM,2}^{-1}A) \leq C \left(1 + \frac{H}{\delta}\right)$$

where δ is the size of the overlap between the subdomains and H the subdomain size.

This does indeed work very well

Number of subdomains	8	16	32	64
ASM	18	35	66	128
ASM + Nicolaides	20	27	28	27

Fails for highly heterogeneous problems
You need a larger and adaptive coarse space

Introduction to GenEO

Adaptive Coarse space for highly heterogeneous Darcy and (compressible) elasticity problems:

GenEO .EVP per subdomain:

Find $V_{j,k} \in \mathbb{R}^{\mathcal{N}_j}$ and $\lambda_{j,k} \geq 0$:

$$A_j^{Neu} V_{j,k} = \lambda_{j,k} D_j R_j A R_j^T D_j V_{j,k}$$

In the two-level ASM, let τ be a user chosen parameter in $[0, 1)$:

Choose eigenvectors $\lambda_{j,k} \leq \tau$ per subdomain:

$$Z := (R_j^T D_j V_{j,k})_{\substack{j=1,\dots,N \\ \lambda_{j,k} \leq \tau}}$$

This automatically includes Nicolaides CS made of Zero

Energy Modes.

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Energy Modes.

Under two technical assumptions:

Theorem (Spillane, Dolean, Hauret, N., Pechstein, Scheichl
(Num. Math. 2013))

If for all j : $0 < \lambda_{j,m_{j+1}} < \infty$:

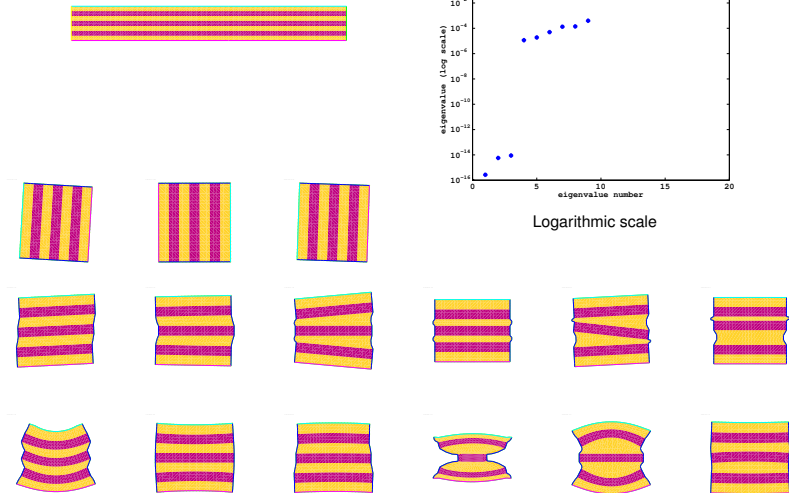
$$\kappa(M_{ASM,2}^{-1}A) \leq (1 + k_0) \left[2 + k_0 (2k_0 + 1) \left(1 + \frac{1}{\tau} \right) \right]$$

Possible criterion for picking τ : (used in our Numerics)

$$\tau := \max_{j=1,\dots,N} \frac{\delta_j}{H_j}$$

$H_j \dots$ subdomain diameter, $\delta_j \dots$ overlap

Eigenvalues and eigenvectors (Elasticity)

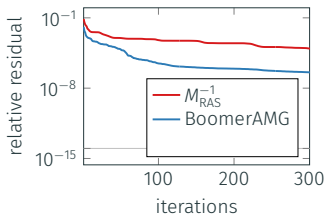
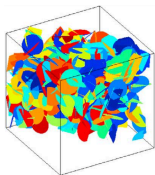
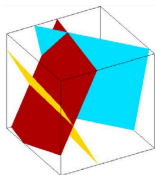


Multiscale FEM based locally *harmonic* spectral coarse space by Ma, Scheichl, Dodwell (2022) and Bénézech, Seelinger, Bastian, Butler, Dodwell, Ma, Scheichl, (2024)

Highly Heterogeneous diffusion problem

USE CASE FOR SPD SYSTEMS I

- HHO for Darcy's law [Ern et al. 2022]
- discrete fracture networks from Inria SERENA
- 140M d.o.f., 4k MPI processes



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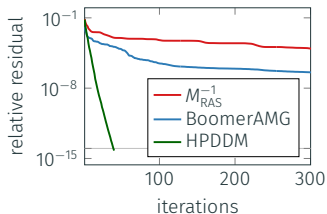
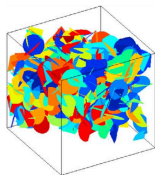
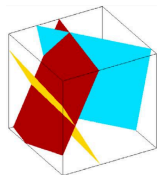
Joint work with P. Jolivet, M. Kern, G. Pichot, et al.

Tailoring multigrid method via a PhD: 3 years and is "fragile"

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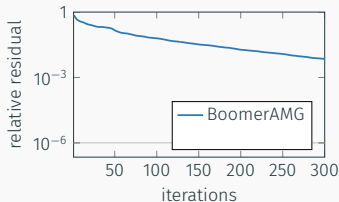
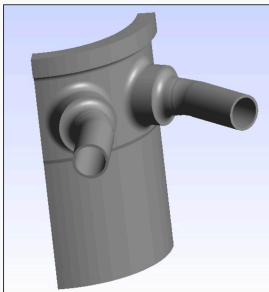
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USE CASE FOR SPD SYSTEMS II

- elasticity for stress corrosion cracking at EDF
- Code_Aster for the discretization
- 68M d.o.f., 1.2k MPI processes

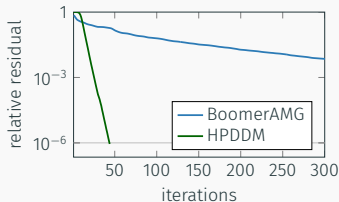
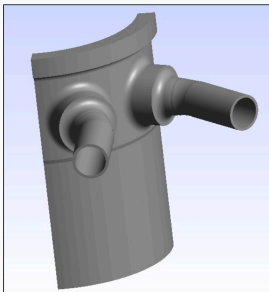


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By courtesy of EDF (N. Tardieu)

USE CASE FOR SPD SYSTEMS II

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Extension of GenEO to Saddle Point problem I

Inverting \mathcal{A} (e.g. Stokes, Nearly inc. elasticity):

$$\mathcal{A} := \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix} = \begin{pmatrix} I_d & 0 \\ BA^{-1} & I_d \end{pmatrix} \begin{pmatrix} A & 0 \\ 0 & -(C + BA^{-1}B^T) \end{pmatrix} \begin{pmatrix} I_d & A^{-1}B^T \\ 0 & I_d \end{pmatrix}$$

is equivalent to inverting A and $S := C + BA^{-1}B^T$. Starting with $A^{-1} \approx^1 M_{ASM2}^{-1}$ as above, we have

$$S \approx C + BM_{ASM2}^{-1}B^T = S_0 + \underbrace{\sum_{i=1}^N \tilde{R}_i^T (\tilde{C}_i + \tilde{B}_i (R_i A R_i^T)^{-1} \tilde{B}_i^T) \tilde{R}_i}_{S_1},$$

where $S_0 := B Z_{GenEO} (Z_{GenEO}^T A Z_{GenEO})^{-1} Z_{GenEO}^T B^T$.

The operator S_1 is itself preconditioned by a **Neumann-Neumann/FETI** preconditioner with a **GenEO** type correction.

¹ \approx means controlled provable spectral equivalence

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Nearly incompressible elasticity

The mechanical properties of a solid are characterized by its elastic energy:

$$\int_{\Omega} 2\mu \underline{\underline{\varepsilon}}(\mathbf{u}) : \underline{\underline{\varepsilon}}(\mathbf{u}) + \lambda |\operatorname{div}(\mathbf{u})|^2$$

where the Lamé coefficients λ and μ are defined in terms of the Young modulus E and Poisson ratio ν :

$$\lambda = \frac{E\nu}{(1+\nu)(1-2\nu)} \quad \text{and} \quad \mu = \frac{E}{2(1+\nu)}.$$

As ν is close to $1/2^-$, $\lambda \rightarrow \infty$ so that $\operatorname{div}(\mathbf{u}) \rightarrow 0$, but the pressure p :

$$p := \lambda \operatorname{div}(\mathbf{u}) \rightarrow p_{\text{incompressibility}}$$

and has thus to be introduced for stability, e.g. $\nu_{\text{rubber}} = 0.4999$.

The resulting discretized variational formulation reads:

$$\begin{cases} \int_{\Omega} 2\mu \underline{\underline{\varepsilon}}(\mathbf{u}_h) : \underline{\underline{\varepsilon}}(\mathbf{v}_h) dx & - \int_{\Omega} p_h \operatorname{div}(\mathbf{v}_h) dx = \int_{\Omega} \mathbf{f} \mathbf{v}_h dx \\ - \int_{\Omega} \operatorname{div}(\mathbf{u}_h) q_h dx & - \int_{\Omega} \frac{1}{\lambda} p_h q_h = 0. \end{cases} \quad (1)$$

where we take the lowest order Taylor-Hood finite element $C0P2 - C0P1$ so that the pressure p_h is continuous. In matrix form we have:

$$\begin{pmatrix} A & B^T \\ B & -C \end{pmatrix} \begin{pmatrix} \mathbf{u}_h \\ p_h \end{pmatrix} = \begin{pmatrix} \mathbf{F}_h \\ 0 \end{pmatrix}.$$

with an arbitrary domain decomposition .

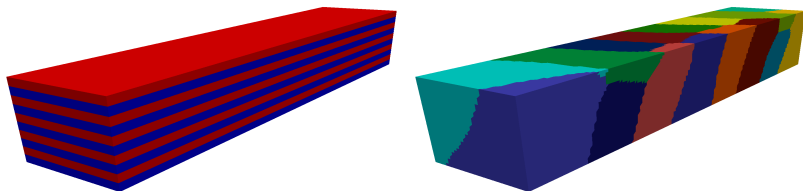


Figure: Heterogeneous beam of rubber and steel. Coefficient distribution (left) and mesh partitioning by the automatic graph partitioner *Metis* (right).

Rubber is nearly incompressible $\nu_{rubber} = 0.4999$ and soft $E_{rubber} = 0.01 \text{ GPa}$ whereas steel is compressible $\nu_{steel} = 0.35$ and hard $E_{steel} = 200. \text{ GPa}$.

#cores	n	$\dim(V_0)$	$\dim(\tilde{W}_0)$	setup(s)	#It	gmres(s)	total(s)	#It N_S^{-1}
262	15 987 380	5 383	3 319	710.7	24	631.6	1342.3	11
525	27 545 495	9 959	2 669	526.6	21	519.5	1046.1	12
1 050	64 982 431	17 837	4 587	675.2	22	665.9	1341.1	11
2 100	126 569 042	32 361	7 995	689.2	25	733.8	1423.0	10
4 200	218 337 384	59 704	13 912	593.0	27	705.4	1298.4	10
8 400	515 921 881	141 421	25 949	735.8	32	1152.5	1888.3	10
16 800	1 006 250 208	260 348	41 341	819.2	29	1717.9	2537.1	12

Table: Weak scaling experiment.

Reproducible script with the open source DSL **FreeFem**²
https://github.com/FreeFem/FreeFem-sources/blob/develop/examples/ffddm/elasticity_saddlepoint.edp

²Hecht since 1987

Comparison with AMG GAMG (PETSc)

Comparisons on the velocity (only) formulation since we were unable to run GAMG on the saddle point formulation.

525 cores	GAMG		DD solver				
	#It	total(s)	$\dim(V_0)$	setup(s)	#It	gmres(s)	total(s)
0.48	56	25.5	41 766	60.4	18	5.0	65.4
0.485	60	26.1	41 984	60.9	20	5.3	66.2
0.49	116	33.3	42 000	60.4	23	5.9	66.3
0.495	>2000	/	42 000	60.4	32	7.6	68.1
0.499	>2000	/	42 000	60.6	95	20.3	81.0

Table: GAMG (PETSc) versus standard GenEO for a homogeneous beam discretized with 7.9 million unknowns.

As ν gets close to 0.5, GAMG fails to compute a solution.

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- focused on domain decomposition [Jolivet, Hecht, Nataf and Prud'homme 2013]
- block Krylov methods [Jolivet and Tournier 2016]
- **POD (Plain Old Data) library, Open-sourced in 2014**
- Written in C++, interfaces in C, Python, and Fortran
- integrated in **PETSc: KSPHPDDM and PCHPDDM** [Jolivet, Roman, Zampini 2021]
- Running on GENCI machines (Irene Rome, A64FX, Jean Zay GPU), porting to Frontier in process
- Used in academia and the industry (FreeFem++, EDF PRISME and ERMES, Denso, ABB, MOOSE)
- **Thermomechanics with 300 millions de d.o.f. on 2000 cores for which multigrid fails.**

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In a 2024 Arxiv preprint with E. Parolin, we develop a general theory for:

$$M^{-1} := \sum_{i=1}^N R_i^T S_i R_i$$

(S_i is a local operator) applied to a well posed, real or complex valued, *sparse* problem $Ax = b$.

Examples:

- RAS algorithm: $S_i = D_i A_i^{-1}$
- RAS algorithm with inexact local solves: e.g.
 $S_i = D_i \text{ILU0}_i^{-1}$
- ASM algorithm: $S_i = A_i^{-1}$

Coarse Space for General Sparse problems

Let C be a Hermitian matrix such that

$$\rho := \|I - C^{-1}A\|_C < 1,$$

and $\tilde{L}_i := \tilde{D}_i - \tilde{S}_i \tilde{R}_i A \tilde{R}_i^*$, (tilde matrices are built on a subdomain enclosing subdomain i , oversampling subdomain), the coarse space is built from local generalized eigenvalues:

$$\tilde{L}_i^* \tilde{R}_i C \tilde{R}_i^* \tilde{L}_i \tilde{u}_{ik} = \lambda_{ik} \tilde{C}_i \tilde{u}_{ik}$$

Then, let $\tau > 0$ be some user-chosen threshold, we can define a two-level preconditioner M_2^{-1} with a coarse space built from the eigenvalues larger than τ such that:

$$\|I - M_2^{-1}A\|_C \leq \frac{\sqrt{k_0 k_1 \tau}}{1 - \rho}.$$

Key remark: if $S_j := D_j A_j^{-1}$ as in RAS, then the local generalized eigenvalues of:

$$\tilde{L}_j^* \tilde{R}_j C \tilde{R}_j^* \tilde{L}_j \tilde{u}_{ik} = \lambda_{ik} \tilde{C}_j \tilde{u}_{ik}$$

are harmonic in the subdomains as in

- *Ma, R. Scheichl, and T. Dodwell, 2022* for Multiscale FEM
- *Q. Hu and Z. Li, 2024* and *Strehlow, Ma, Scheichl, 2024* where the two-level RAS preconditioner with **exact local solves** is studied

Note that our theory applies to the RAS algorithm with **inexact local solves** as well as to other DD preconditioners:

- additive Schwarz Method (ASM): $S^j := A_j^{-1}$
- Symmetrized Optimized Restricted Additive Schwarz (SORAS): $S^j := D_j A_j^{-1} D_j$

- Smaller coarse space than the GenEO coarse space.
- Efficiency with Incomplete Cholevski (ICC)
- More on the way

"Past" works

- Spectral Coarse space for SPD problems
- Iterative solver for saddle point problem with highly heterogeneous linear elasticity, Stokes systems, ...
- Available via HPDDM with or without its interface to PETSc
- Available to FreeFem user

Recent works (2024) on coarse space constructions

- Nataf, Parolin, "Coarse spaces for non-symmetric two-level preconditioners based on local generalized eigenproblems", 2024
- Al Daas, Jolivet, Nataf, Tournier (2024). "A robust two-level Schwarz preconditioner for sparse matrices", submitted for publication. (not shown here)



See also *Heinlein, Smetana, 2022*

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-  F Nataf and P.-H. Tournier, "A GenEO Domain Decomposition method for Saddle Point problems", Comptes Rendus. Académie des Sciences, 2023
-  F Nataf and E. Parolin, "Coarse spaces for non-symmetric two-level preconditioners based on local generalized eigenproblems", Arxiv, 2024