

Un espace grossier de type GenEO pour les problèmes de point selle

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Numerical solutions for saddle point problems arising in Fluid dynamics
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- 1 Saddle Point Problem and Solvers
- 2 Recall on GenEO for SPD problems
- 3 Extension of GenEO to Saddle Point problem
- 4 Numerical Results and the ffdm script

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Solve

$$\mathcal{A} \begin{pmatrix} \mathbf{u}_h \\ p_h \end{pmatrix} = \begin{pmatrix} \mathbf{F}_h \\ G_h \end{pmatrix} \text{ with } \mathcal{A} := \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix}.$$

Pervasive in scientific computing:

- (nearly) incompressible fluids or solids \Rightarrow pressure formulation is usually mandatory.
- Multi Point Constraints (MPC) \Rightarrow Lagrange multipliers.

For small enough problems, direct solvers are the method of choice (MUMPS, PARDISO, SUPERLU, ...)

Comparison with a Direct solver *MUMPS* on a steel-rubber 3D beam

Timings are in seconds. OOM means: Out Of available Memory

n	#cores	MUMPS			DD saddle point solver			
		setup	solve	total	setup	#It	gmres	total
134 000	16	7.1	0.1	7.2	27.1	18	19.7	46.8
1 058 000	32	85.7	0.8	86.5	166.2	20	137.2	303.4
1 058 000	65	71.0	0.6	71.6	91.0	21	77.1	168.1
1 058 000	131	63.2	0.5	63.7	59.7	24	49.7	109.4
3 505 000	55	477.8	3.7	481.5	404.1	24	430.1	834.2
3 505 000	110	392.3	2.3	394.6	242.5	23	212.8	455.3
3 505 000	221	387.0	2.1	389.1	134.8	23	109.4	244.2
3 505 000	442	453.9	2.2	456.1	88.2	24	68.6	156.8
8 235 000	262	OOM	/	/	278.5	25	264.3	542.8
8 235 000	525	1622.1	6.1	1628.2	172.1	24	136.0	308.1
8 235 000	1050	1994.3	7.4	2001.7	136.5	25	99.7	236.2

Maximum problem size with direct solver is around 10 million unknowns.

The [GenEO domain decomposition solver](#) introduced here will solve a problem with 1 billion unknowns.

Difficulty: Matrix \mathcal{A} is symmetric but not positive. If it is made positive, symmetry is lost \Rightarrow issue for iterative solvers.

$$\mathcal{A} := \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix}.$$

Algebraic multigrid and Domain Decomposition solvers:

As problems get large, penalization and augmented Lagrangian techniques may enhance convergence but at the expense of approximation errors and round-off error issues.

For saddle point problems with 3D nearly incompressible elasticity and arbitrary high heterogeneities, existing iterative solvers seem not be so usable.

Here, we propose an

[Extension of the GenEO DDM to saddle point problems.](#)

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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, $V_H = \text{span } Z$, writing $R_0 = Z^T$ we define the two level preconditioner as:

$$M_{ASM,2}^{-1} := R_0^T (R_0 A R_0^T)^{-1} R_0 + \sum_{i=1}^N R_i^T A_i^{-1} R_i.$$

The **Nicolaides approach** (1987) is to use the near-kernel of the local operators to build the coarse space:

$$R_0^T Z := (R_i^T D_i R_i \mathbf{1})_{1 \leq i \leq N},$$

where D_i are chosen so that we have a partition of unity: $\sum_{i=1}^N R_i^T D_i R_i = Id$. Key notion: **Stable splitting** (J. Xu, 1989)

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

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

GenEO .EVP per subdomain:

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

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

In the two-level ASM, let τ be a user chosen parameter:

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

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

This automatically includes Nicolaides CS made of Zero Energy Modes.

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Two technical assumptions.

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

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

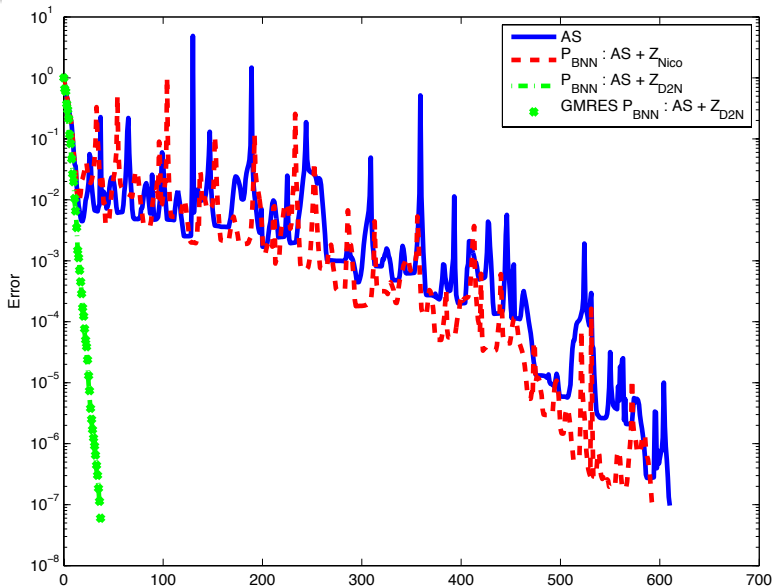
$$\kappa(M_{ASM,2}^{-1}A) \leq (1 + k_0) \left[2 + k_0 (2k_0 + 1) (1 + \tau) \right]$$

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

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

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

Convergence on a Highly Heterogeneous diffusion problem



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Extension of GenEO to Saddle Point problem

Preconditioning \mathcal{A} (e.g. Stokes, Nearly incompressible elasticity):

$$\mathcal{A} := \begin{pmatrix} A & B^T \\ B & -C \end{pmatrix}.$$

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

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

where $S_0 := B Z_{GenEO} (Z_{GenEO}^T A Z_{GenEO})^{-1} Z_{GenEO}^T B^T$.
The operator S_1 is dense and has to be preconditioned.

Extension of GenEO to Saddle Point problem

But as a sum of local Schur complements, S_1 can be preconditioned by a **Neumann-Neumann** preconditioner

$$M_{S_1, \text{one level}}^{-1} := \sum_{i=1}^N \tilde{R}_i^T \tilde{D}_i (\tilde{C}_i + \tilde{B}_i (R_i A R_i^T)^{-1} \tilde{B}_i^T)^\dagger \tilde{D}_i \tilde{R}_i.$$

made scalable and robust with a **GenEO** type correction :

$$M_{S_1}^{-1} := Z_{S_1} (Z_{S_1}^T S_1 Z_{S_1})^{-1} Z_{S_1}^T + \left(\sum_{i=1}^N \tilde{R}_i^T \tilde{D}_i (I_d - \xi_i) (\tilde{C}_i + \tilde{B}_i (R_i A R_i^T)^{-1} \tilde{B}_i^T)^\dagger (I_d - \xi_i^T) \tilde{D}_i \tilde{R}_i \right).$$

where Z_{S_1} is populated with weighted local eigenvectors corresponding to the largest eigenvalues of the following GEVP:

$$\tilde{D}_i \tilde{R}_i S_1 \tilde{R}_i^T \tilde{D}_i \tilde{\mathbf{P}}_{ik} = \mu_{ik} (\tilde{C}_i + \tilde{B}_i (R_i A R_i^T)^{-1} \tilde{B}_i) \tilde{\mathbf{P}}_{ik}, \quad (1)$$

and ξ_i denotes an orthogonal projection on the local contribution of the subdomain to the coarse space.

Two Stage Algorithm

Define N_S^{-1} a spectrally equivalent preconditioner to S :

$$N_S := S_0 + M_{S_1}.$$

The application of the preconditioner N_S^{-1} consists in solving:

$$N_S \mathbf{P} = \mathbf{G},$$

by a Krylov solver with $M_{S_1}^{-1}$ as a preconditioner.

Saddle point algorithm in three solves:

INPUT: $\begin{pmatrix} \mathbf{F}_U \\ \mathbf{F}_P \end{pmatrix} \in \mathbb{R}^{n+m}$ OUTPUT: $\begin{pmatrix} \mathbf{U} \\ \mathbf{P} \end{pmatrix}$ the solution.

1. Solve $\mathbf{A}\mathbf{G}_U = \mathbf{F}_U$ by a PCG with M_A^{-1} as a preconditioner
2. Compute $\mathbf{G}_P := \mathbf{F}_P - \mathbf{B}\mathbf{G}_U$
3. Solve $\mathbf{S}\mathbf{P} := (\mathbf{C} + \mathbf{B}\mathbf{A}^{-1}\mathbf{B}^T)\mathbf{P} = -\mathbf{G}_P$ by a PCG with N_S^{-1} as a preconditioner (nested loops).
4. Compute $\mathbf{G}_U := \mathbf{F}_U - \mathbf{B}^T\mathbf{P}$
5. Solve $\mathbf{A}\mathbf{U} = \mathbf{G}_U$ by a PCG with M_A^{-1} as a preconditioner

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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:

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

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 .

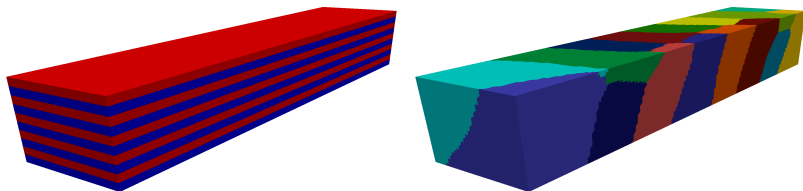


Figure: Heterogeneous beam of rubber and steel. Coefficient distribution (left) and mesh partitioning by the automatic graph partitioner *Metis* (right).

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	$\dim(V_0)$	$\dim(W_0)$	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

```
https://github.com/FreeFem/FreeFem-sources/  
blob/develop/examples/ffddm/elasticity\_  
saddlepoint.edp
```

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 ν	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.

Algorithm assessment on Stokes computations made for the minisymposium

#cores	n	$\dim(V_0)$	$\dim(\tilde{W}_0)$	setup(s)	#It	gmres(s)	total(s)	#It N_S^{-1}
4	717 837	93	4	420.0	11	107.2	527.2	6
8	717 837	151	8	197.9	11	56.2	254.1	6
16	717 838	267	16	115.2	12	35.6	150.8	7
64	717 838	616	65	44.0	14	15.2	59.2	9
32	717 842	420	36	71.5	13	21.9	93.4	8
8	2 867 499	327	8	792.7	11	236.5	1029.2	7
16	2 867 499	577	16	371.6	12	148.2	519.8	9
32	2 867 499	877	32	291.3	12	86.8	378.1	10
64	2 867 503	1 306	66	164.9	13	55.9	220.8	11
128	2 867 503	1 985	133	118.6	13	41.4	160.0	13
8	11 462 307	606	8	3365.4	11	1146.3	4511.7	8
16	11 462 307	1 133	16	1753.6	11	640.4	2394.0	11
32	11 462 307	1 827	32	1099.9	12	404.8	1504.7	13
64	11 462 307	2 760	64	628.0	12	213.9	841.9	13
128	11 462 307	4 124	134	438.5	13	162.1	600.6	15

Table: 2D Stokes bubbles experiments.



Going further: Comparisons for 3D flows with multigrid solvers, Cahouet-Chabard method on this or other problems, ...

- Iterative solver for saddle point problem with highly heterogeneous coefficients that works for linear elasticity, Stokes systems, ...
- Available to FreeFem users via https://github.com/FreeFem/FreeFem-sources/blob/develop/examples/ffddm/elasticity_saddlepoint.edp
- Preprint available on HAL:
 - 📄 F Nataf and P.-H. Tournier, "A GenEO Domain Decomposition method for Saddle Point problems", <https://hal.archives-ouvertes.fr/view/index/docid/3450974> , HAL Archive.
- Prospects
 - More than 2-level
 - Inclusion into HPDDM for PETSc users
 - Multiscale finite element for saddle point problem

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