

FREEFEM, PETSc, SLEPc, AND HPDDM

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https://joliv.et/FF_16th.pdf

FreeFEM days, 16th edition

INTRODUCTION

1. Introduction
2. Using PETSc
3. How to make my code go faster?
4. Advanced functionalities
5. Domain decomposition methods
6. Conclusion

INTRODUCTION

BASIC SEQUENTIAL FUNCTIONALITIES

Domain specific language for FE

- mesh structure
- associated finite element spaces
- vector and matrix assemblies
- algebraic operations

INTERFACE TO OTHER LIBRARIES

- installed by PETSc (`make petsc-slepc`)
 - MPICH (if no other MPI implementation)
 - CMake (if unavailable)
 - Bison
 - BLAS (if unavailable)
 - MUMPS (if Fortran available)
 - SuperLU_DIST (if Fortran unavailable)
 - SuperLU
 - SuiteSparse
 - hypre*
 - [Par]METIS
 - SCOTCH
 - TetGen
 - SLEPc
 - HPDDM (if shared libraries available)
 - ARPACK (if Fortran available)
 - [Par]Mmg
 - Htool

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⇒ sequential or parallel (with FreeFem++-mpi)

USING PETSc

Application Codes

**TS**

Time Steppers

Pseudo-Transient, Runge-Kutta, IMEX, SSP, ...
Local and Global Error Estimators
Adaptive Timestepping
Event Handling
Sensitivity via Adjoints



PDE-Constrained
Adjoint-Based
Derivative-Free

Levenberg-Marquardt
Newton's Method
Interior Point Methods

SNES

Nonlinear Solvers

Newton Line search Successive Substitution
Newton Trust Region Nonlinear CG
BFGS (Quasi-Newton) Active Set VI
Nonlinear Gauss-Seidel

KSP

Linear Solvers

CG, GMRES, BiCGStab, FGMRES, ...
Pipelined Krylov Methods
Hierarchical Krylov Methods

**PC**

Preconditioners

ILU/ICC
Additive Schwarz
Fieldsplit (Block Preconditioners)
PCMG (Geometric Multigrid)
GAMG (Algebraic Multigrid)

Vec

Vectors

IS

Index Sets

Mat

Linear Operators

AIJ (Compressed Sparse Row)
SAIJ (Symmetric)
BAIJ (Blocked)
Dense
GPU Matrices

PetscSF

Parallel Communication

Communication and Computational Kernels

MPI

BLAS/LAPACK

Kokkos

CUDA

...

SLEPc
Eigensolvers

SPECIFIC TYPE

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- `ObjectView` or `MatLoad`

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- `set(sparams = "...")` to fine-tune solvers
- `ObjectView` or `MatLoad`
- single exposed type but many implementations
 - `MatAIJ`
 - `MatSBAIJ`
 - `MatHtool`
 - `MatNest`
 - `MatShell`
 - `MatDense`
 - `MatComposite`
 - `MatIS`
 - `MatTranspose`

HOW TO MAKE MY CODE GO FASTER?

GOING PARALLEL

- no shared-memory parallelism
- distributed-memory paradigm (MPI)

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- no shared-memory parallelism
- distributed-memory paradigm (MPI)
- not fully transparent to users
- lightweight macros for workload distribution

FINITE ELEMENTS IN PARALLEL

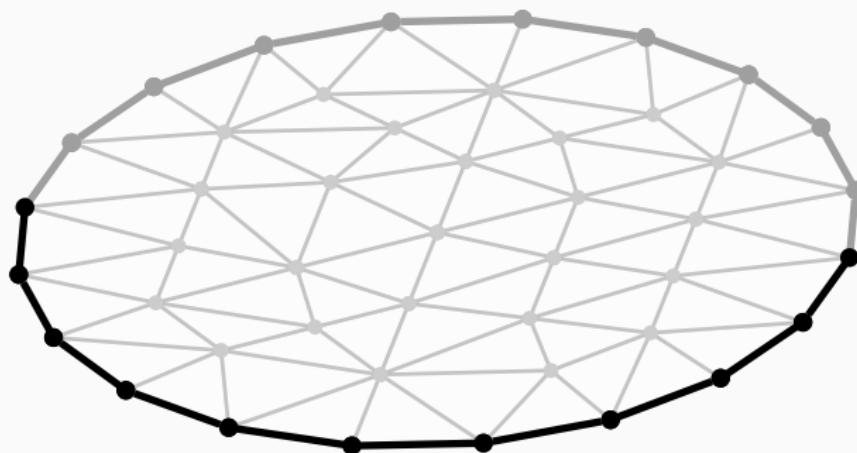
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- distribute data efficiently to minimize communication
- orthogonal to shared-memory parallelism (OpenMP)

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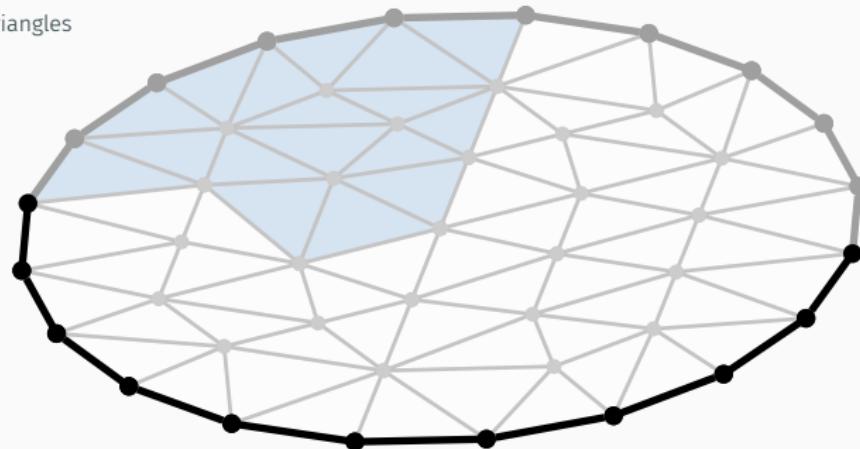


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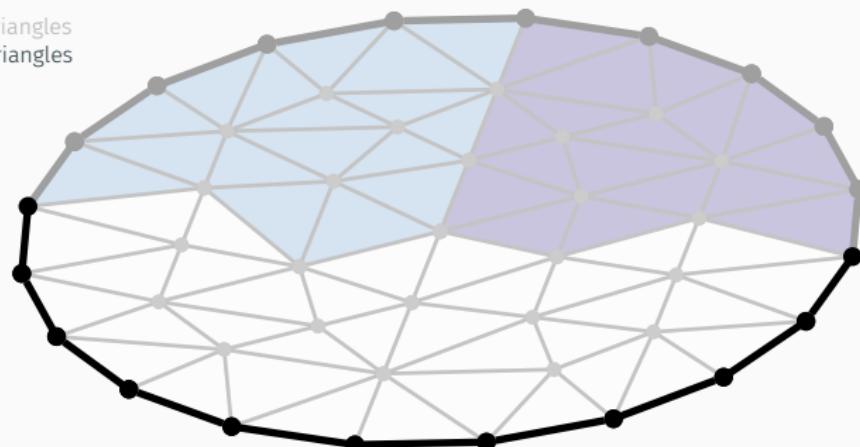


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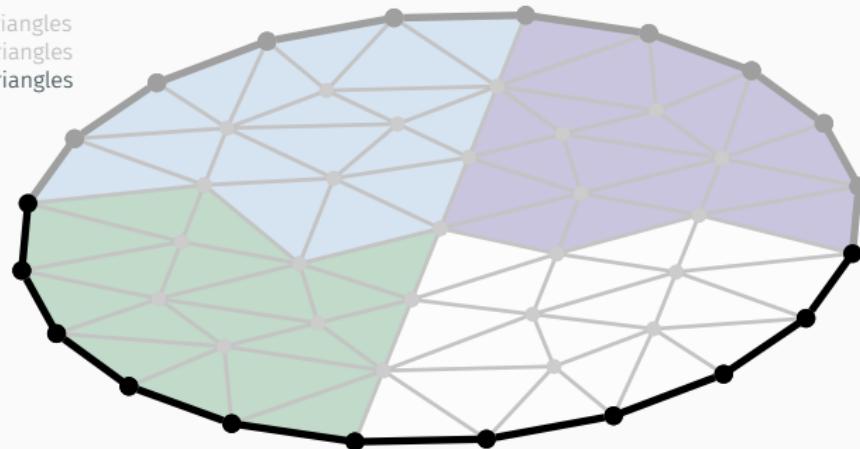


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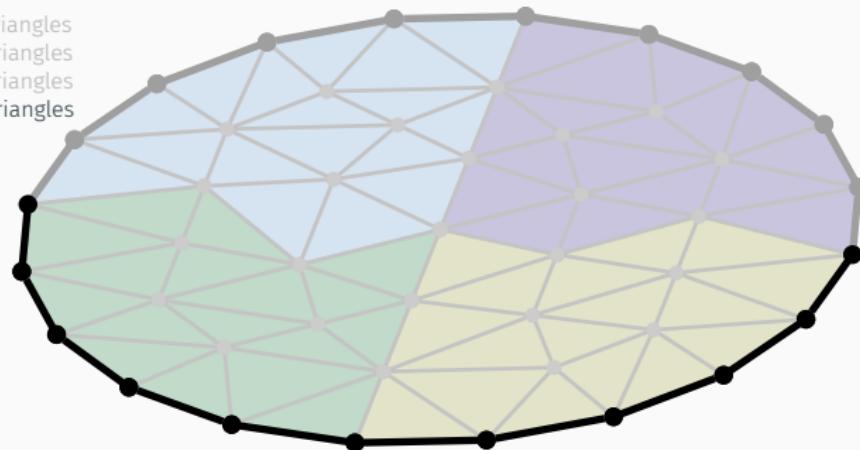


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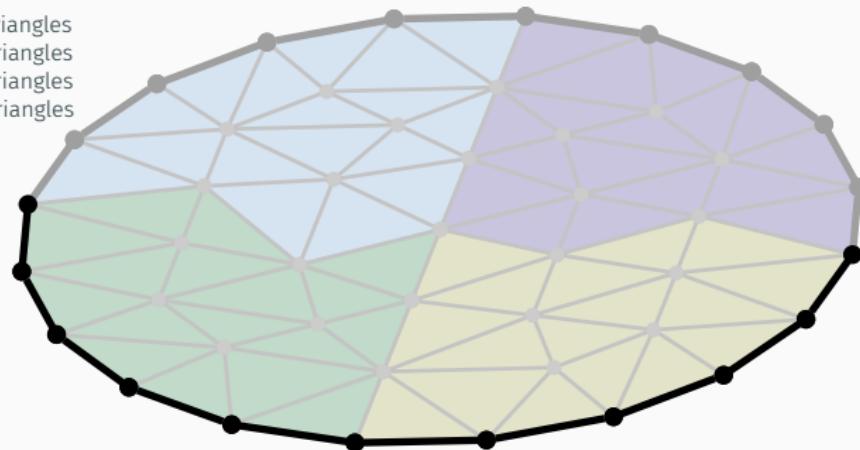


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THE END OF THE STORY?

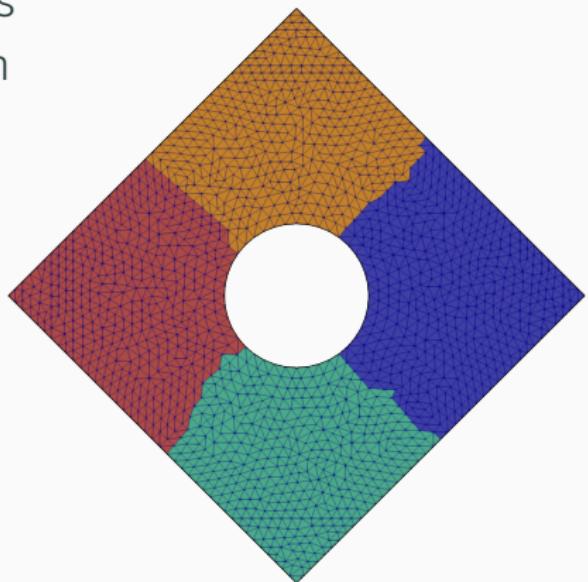
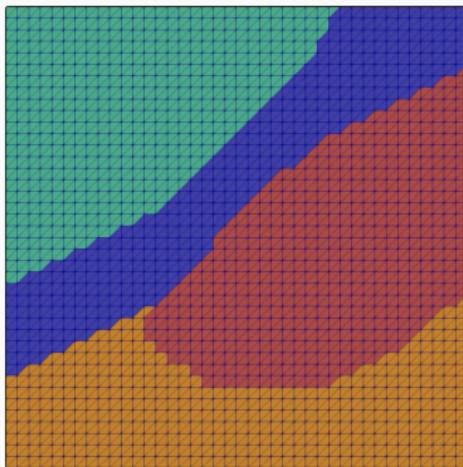
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THE END OF THE STORY?

- assemblies are now easy to parallelize
- many algorithms are still challenging
 - distance to a level-set
 - unstructured adaptive mesh refinement
 - nonlinear and linear solvers

ANOTHER EXAMPLE: INTERPOLATION

- different meshes
- different basis functions
- different decomposition



SMALL SNIPPET OF macro_ddm.idp

- DmeshSave
- DmeshLoad
- DmeshCreate
- MatCreate
- VecInterpolate
- MatInterpolate

SMALL SNIPPET OF `macro_ddm.idp`

- `DmeshSave`
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- examples, see `examples/hpddm/README.md`
- functionalities, see `CHANGELOG.md`
- <https://joliv.et/FreeFem-tutorial>

ADVANCED FUNCTIONALITIES

NONSTANDARD EIGEN SOLVERS

- nonlinear $T(\lambda)x = 0 \implies \text{NEPSolve}$
- polynomial $(\sum_{i=0}^N \lambda^i A_i)x = 0, N \geq 2 \implies \text{PEPSolve}$
- SVD $Av = \sigma u \implies \text{SVDSolve}$
- more in [Roman et al., 2023]

Array of PETSc matrices

As easy as FreeFEM matrices: `Mat[int] A(N)`

EASIER USE OF SHELLS FOR PRECONDITIONERS

Oseen equations: $A = \begin{bmatrix} F & B \\ B^T & C \end{bmatrix} \Rightarrow \text{Mat } A = [[F, B], [B^T, C]]$

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Mat Object: 4 MPI processes

type: nest

rows=37507, cols=37507

Matrix object:

type=nest, rows=2, cols=2

MatNest structure:

(0,0) : prefix="fieldsplit_0_", type=mpiaij...

(0,1) : type=mpiaij, rows=33282, cols=4225

(1,0) : type=transpose, rows=4225, cols=33282

(1,1) : prefix="fieldsplit_1_", type=mpiaij...

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Triangular preconditioner: $\tilde{A} = \begin{bmatrix} \tilde{F} & B \\ 0 & \tilde{S} \end{bmatrix}$ with $\tilde{S}^{-1} \approx -M_p^{-1}F_pL_p^{-1}$

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```
func real[int] PCD(real[int]& in) {
    real[int] out(in.n);
    KSPSolve(Lp,in,in); MatMult(Fp,in,out);
    KSPSolve(Mp,out,out); out *= -1.0;
    return out; // = -Mp \ Fp * Lp \ in
}
set(C, parent = A, precon = PCD,
    sparams = "-fieldsplit_1_pc_type shell");
```

FEM-BEM COUPLING

Solve the generalized eigenvalue problem for (u, λ) :

$$\begin{bmatrix} A_k & D'_k - \frac{1}{2}M^{\text{tr}} \\ M^{\text{tr}} & -S_k \end{bmatrix} u = \lambda \begin{bmatrix} M & 0 \\ 0 & 0 \end{bmatrix} u$$

```
Mat<complex> C;  
Sk *= -1.0;  
C = [[Ak , R*TDkM]  
     [F2B, Sk    ]];
```

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ObjectView(C);
```

Mat Object: 4 MPI processes
type=nest, rows=2, cols=2
MatNest structure:
(0,0) : type=mpiaij, rows=10410, cols=10410
(0,1) : type=composite, rows=10410, cols=...
(1,0) : type=mpiaij, rows=1156, cols=10410
(1,1) : type=htool, rows=1156, cols=1156

MORE ECONOMICAL MatNest

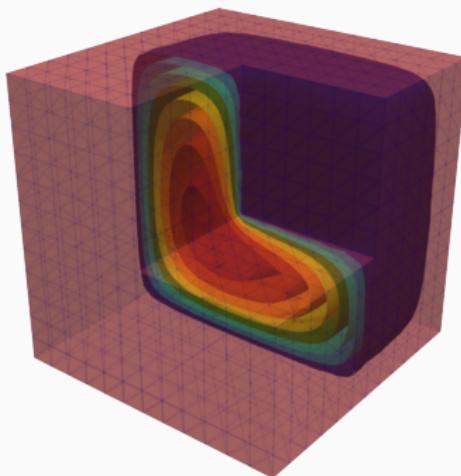
- support for dense blocks `real[int, int] d`
- pay the (memory) price twice with
`matrix A = [[a, d],
[d', 0]]`

MORE ECONOMICAL MatNest

- support for dense blocks `real[int, int] d`
- pay the (memory) price twice with
 - `matrix A = [[a, d],
[d', 0]]`
- no such overhead with
 - `Mat B = [[b, d],
[d', 0]]`
- the action of the transpose is computed implicitly
- `MATSOLVERMUMPS` handles `MatNest` since 3.20

PARALLEL MESH ADAPTATION

- Poisson equation on the Fischera corner
- algebraic system solved with GAMG
- 128 MPI processes of Irène@TGCC



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Iteration #2

# of tetrahedra	47k
ParMmg (sec)	2.6
transfer (sec)	2.8
GAMG (sec)	0.2

PARALLEL MESH ADAPTATION

- Poisson equation on the Fischera corner
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Iteration	#2	#3
# of tetrahedra	47k	91k
ParMmg (sec)	2.6	4.9
transfer (sec)	2.8	1.9
GAMG (sec)	0.2	0.2

PARALLEL MESH ADAPTATION

- Poisson equation on the Fischera corner
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Iteration	#2	#3	#4
# of tetrahedra	47k	91k	221k
ParMmg (sec)	2.6	4.9	16.2
transfer (sec)	2.8	1.9	1.4
GAMG (sec)	0.2	0.2	0.5

PARALLEL MESH ADAPTATION

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Iteration	#2	#3	#4	#5
# of tetrahedra	47k	91k	221k	686k
ParMmg (sec)	2.6	4.9	16.2	22.0
transfer (sec)	2.8	1.9	1.4	1.5
GAMG (sec)	0.2	0.2	0.5	0.6

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Iteration	#2	#3	#4	#5	#6
# of tetrahedra	47k	91k	221k	686k	2.2M
ParMmg (sec)	2.6	4.9	16.2	22.0	49.9
transfer (sec)	2.8	1.9	1.4	1.5	2.7
GAMG (sec)	0.2	0.2	0.5	0.6	0.9

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Iteration	#2	#3	#4	#5	#6	#7
# of tetrahedra	47k	91k	221k	686k	2.2M	7M
ParMmg (sec)	2.6	4.9	16.2	22.0	49.9	211
transfer (sec)	2.8	1.9	1.4	1.5	2.7	11.5
GAMG (sec)	0.2	0.2	0.5	0.6	0.9	2.8

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Iteration	#2	#3	#4	#5	#6	#7	#8
# of tetrahedra	47k	91k	221k	686k	2.2M	7M	22M
ParMmg (sec)	2.6	4.9	16.2	22.0	49.9	211	180
transfer (sec)	2.8	1.9	1.4	1.5	2.7	11.5	66
GAMG (sec)	0.2	0.2	0.5	0.6	0.9	2.8	4.2

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Iteration	#2	#3	#4	#5	#6	#7	#8	#9
# of tetrahedra	47k	91k	221k	686k	2.2M	7M	22M	62M
ParMmg (sec)	2.6	4.9	16.2	22.0	49.9	211	180	299
transfer (sec)	2.8	1.9	1.4	1.5	2.7	11.5	66	87
GAMG (sec)	0.2	0.2	0.5	0.6	0.9	2.8	4.2	6.0

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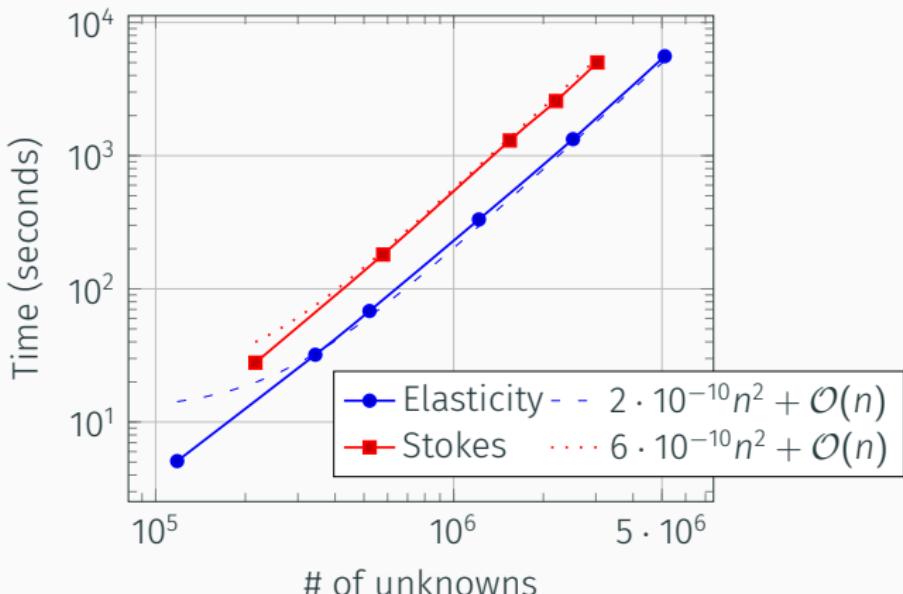
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Iteration	#2	#3	#4	#5	#6	#7	#8	#9	#10
# of tetrahedra	47k	91k	221k	686k	2.2M	7M	22M	62M	170M
ParMmg (sec)	2.6	4.9	16.2	22.0	49.9	211	180	299	697
transfer (sec)	2.8	1.9	1.4	1.5	2.7	11.5	66	87	297
GAMG (sec)	0.2	0.2	0.5	0.6	0.9	2.8	4.2	6.0	—

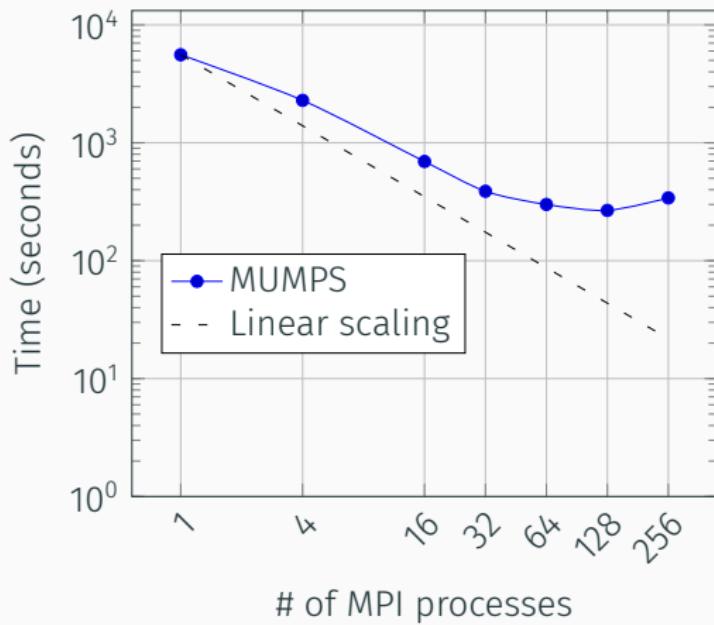
DOMAIN DECOMPOSITION METHODS

COMPLEXITY STUDY OF EXACT FACTORIZATION

- 3D problems, lowest-order FE
- sequential, double-precision, exact LDL^T factorization
- <https://mumps-solver.org>

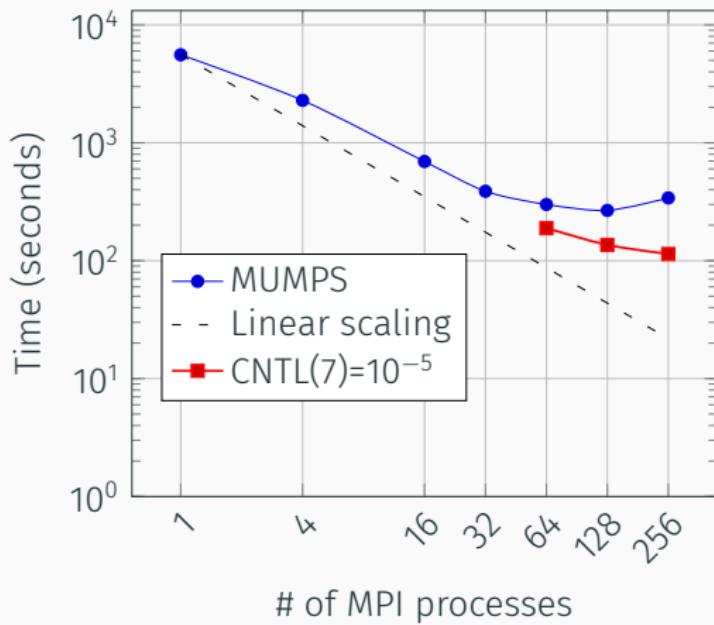


SCALABILITY FOR THE LARGEST (ELASTICITY) PROBLEM



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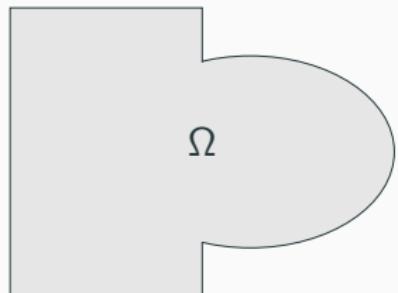
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- FGMRES with a 10^{-5} tolerance, 15 iterations

OVERLAPPING SCHWARZ METHODS

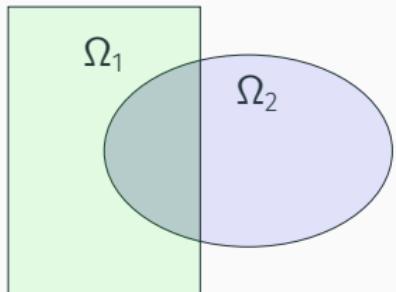
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Two-way decomposition of $\llbracket 1; n \rrbracket$ and restriction operators

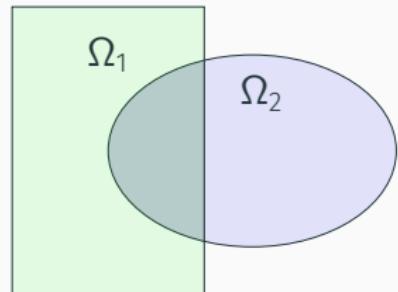


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Two-way decomposition of $\llbracket 1; n \rrbracket$ and restriction operators

$$M_{\text{ASM}}^{-1} = \sum_{i=1}^{N=2} R_i^T (R_i A R_i^T)^{-1} R_i$$



[Schwarz, 1870]

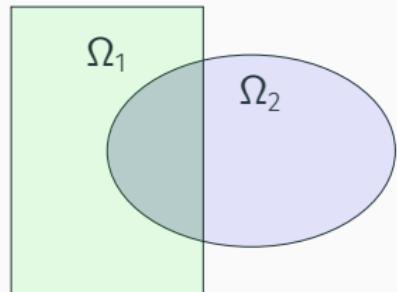
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[Schwarz, 1870] [Cai and Sarkis, 1999]

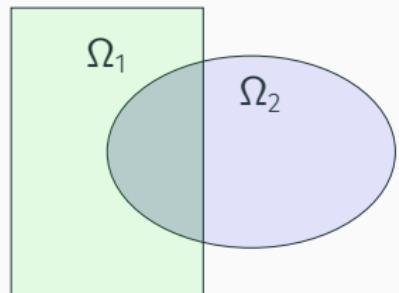
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[Schwarz, 1870] [Cai and Sarkis, 1999]

→ no guaranteed scalability for large N

COARSE CORRECTION FOR DOMAIN DECOMPOSITION

Abstract setting

- define locally $W_i = [\Lambda_1 \quad \cdots \quad \Lambda_{\gamma_i}]$

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COARSE CORRECTION FOR DOMAIN DECOMPOSITION

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How to choose γ_i and $\{\Lambda_j\}_{j=1,\dots,\gamma_i}$?

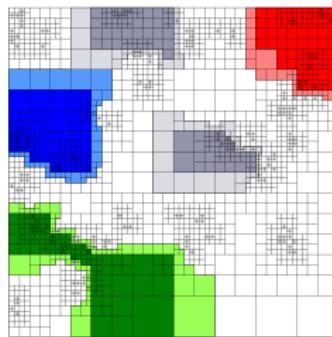
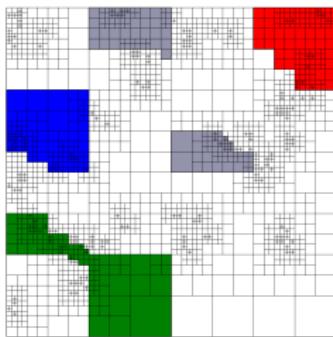
- aggressive coarsening
- capturing the “difficulties” of A
- at a moderate cost

HPDDM

- specific implementation [Jolivet, Hecht, Nataf, et al., 2013]

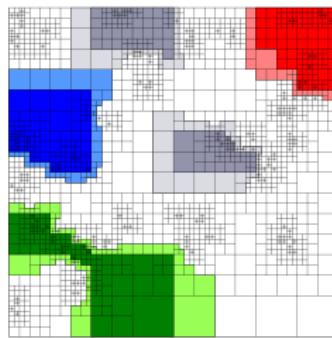
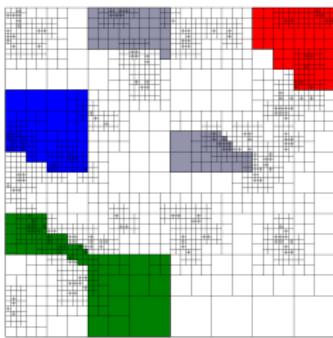
HPDDM + PETSc

- specific implementation [Jolivet, Hecht, Nataf, et al., 2013]
- interfaced in PETSc [Jolivet, Roman, and Zampini, 2021]
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HPDDM + PETSc

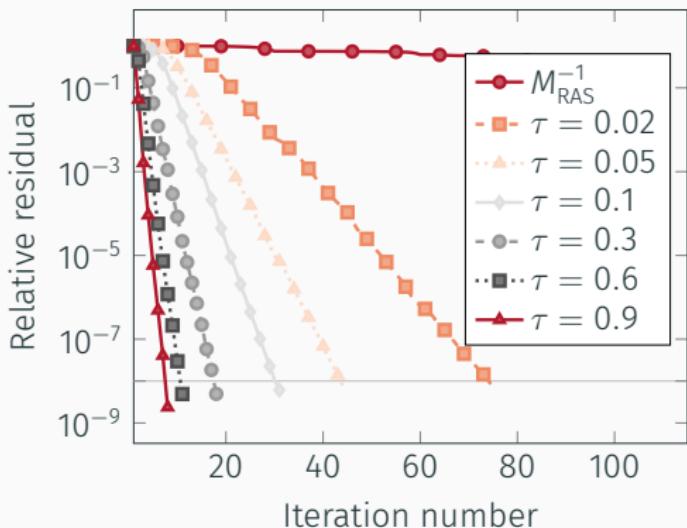
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→ leveraging the flexibility of PETSc for algebraic variants

ADAPTIVE PRECONDITIONING FOR GENERAL SYSTEMS

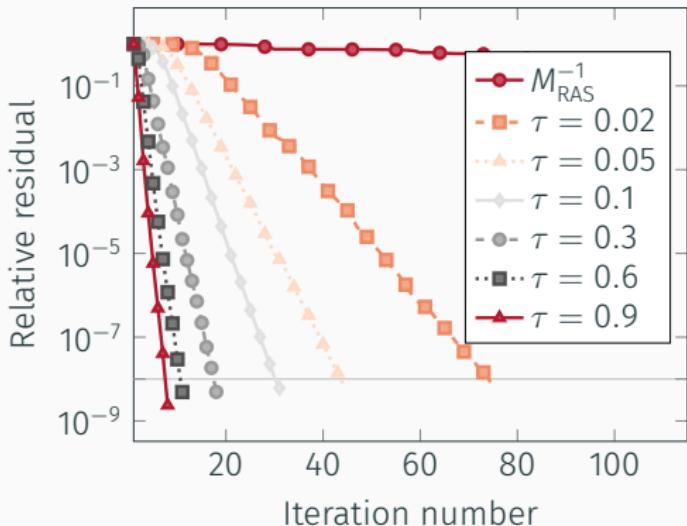
G3_circuit from SSMC on
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τ	n_C	Iterations
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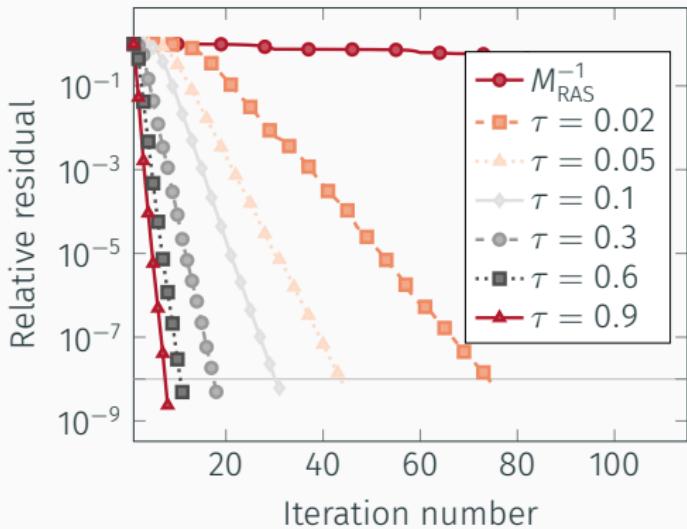


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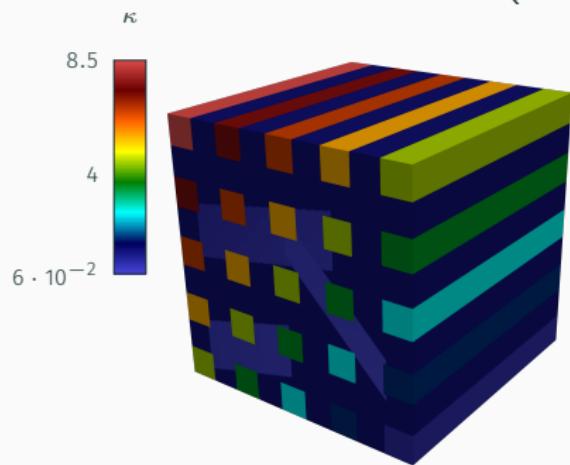
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Chevron4	5	†	†
consph	93	†	†
nxp1	20	†	†

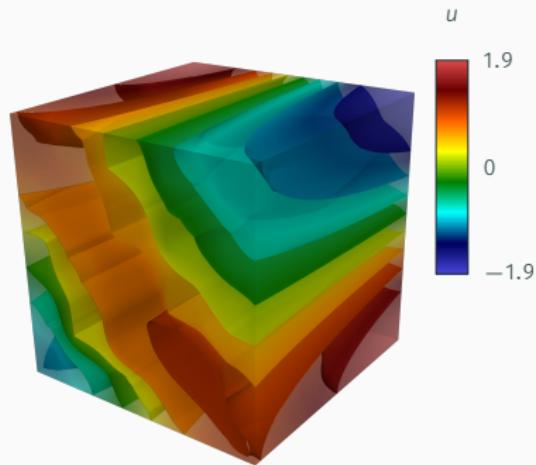
NONLINEAR PROBLEM

- automatic assembly of Neumann matrices in FreeFEM
- Newton method using SNES [Brune et al., 2015]

$$-\nabla \cdot (\kappa \nabla u) - 6.2e^u = 0$$



(a) Coefficient distribution

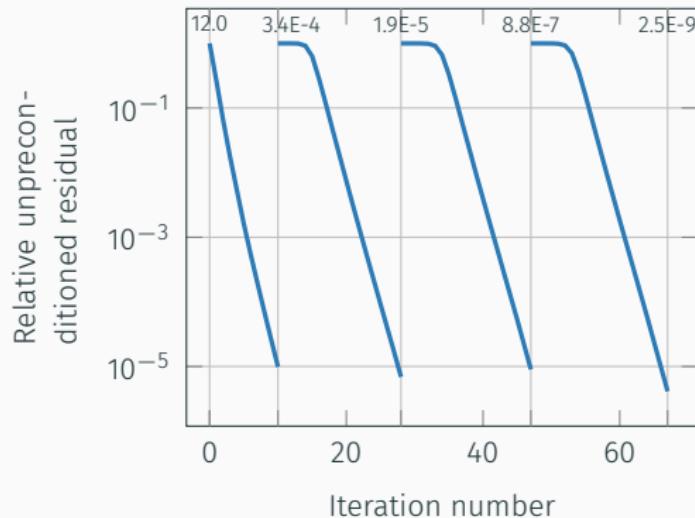


(b) Isosurfaces of the solution

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$$Ax = \begin{bmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \end{bmatrix} = \begin{bmatrix} b_0 \\ b_1 \end{bmatrix}$$

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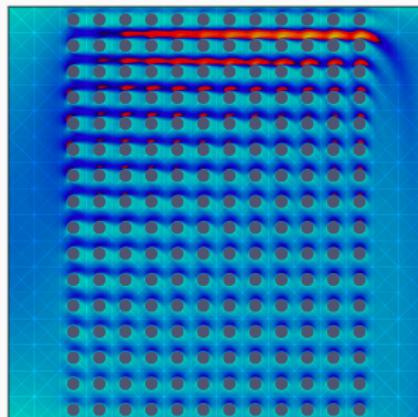
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- more at **stokes-block-hpddm-2d-PETSc.edp**

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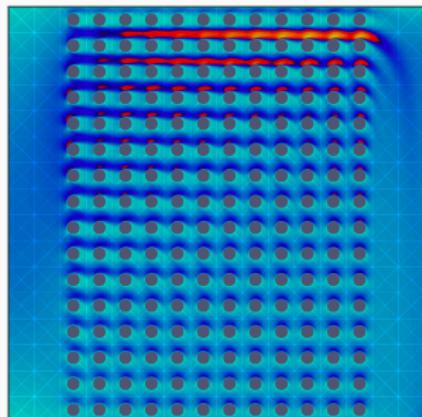
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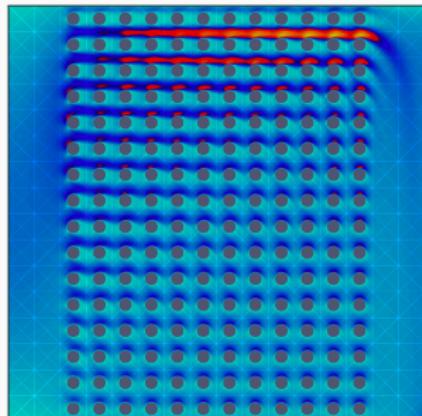
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harmonic	5	22	7

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5×10^{-4}	outer solves	A_{00} solves	S solves
GenEO	†	†	†
harmonic	6	14	6

CONCLUSION

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Thank you!

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