

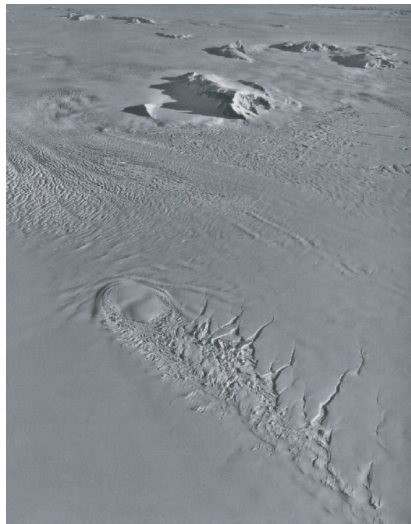
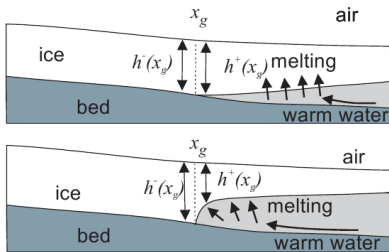
Scalable solvers for 3D Stokes problems

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Why do we need 3D Stokes?



Non-Newtonian Stokes system

- Strong form: Find $(\mathbf{u}, p) \in \mathcal{V}_D \times \mathcal{P}$ such that

$$-\nabla \cdot (\eta D\mathbf{u}) + \nabla p - \mathbf{f} = \mathbf{0}$$

$$\nabla \cdot \mathbf{u} = 0$$

where

$$D\mathbf{u} = \frac{1}{2} (\nabla \mathbf{u} + (\nabla \mathbf{u})^T)$$

$$\gamma(D\mathbf{u}) = \frac{1}{2} D\mathbf{u} : D\mathbf{u}$$

$$\eta(\gamma) = B(\Theta, \dots) (\epsilon + \gamma)^{\frac{p-2}{2}}, \quad p = 1 + \frac{1}{n} \approx \frac{4}{3}$$

with boundary conditions

$$(D\mathbf{u} - p\mathbf{1}) \cdot \mathbf{n} = \begin{cases} \mathbf{0} & \text{free surface} \\ -\rho_w z \mathbf{n} & \text{ice-ocean interface} \end{cases}$$

$$\mathbf{u} = \mathbf{0} \quad \text{frozen bed, } \Theta < \Theta_0$$

$$\left. \begin{aligned} \mathbf{u} \cdot \mathbf{n} &= \mathbf{g}_{\text{melt}}(T\mathbf{u}, \dots) \\ T(D\mathbf{u} - p\mathbf{1}) \cdot \mathbf{n} &= \mathbf{g}_{\text{slip}}(T\mathbf{u}, \dots) \end{aligned} \right\} \text{nonlinear slip, } \Theta \geq \Theta_0$$

Other forms

- Minimization form: Find $\mathbf{u} \in \mathcal{V}_D$ which minimizes

$$\mathcal{I}(\mathbf{u}) = \int_{\Omega} |D\mathbf{u}|^p - \mathbf{f} \cdot \mathbf{u}$$

subject to

$$\nabla \cdot \mathbf{u} = 0$$

- Weak form: Find $(\mathbf{u}, p) \in \mathcal{V}_D \times \mathcal{P}$ such that

$$\int_{\Omega} \eta D\mathbf{v} : D\mathbf{u} - p \nabla \cdot \mathbf{v} - q \nabla \cdot \mathbf{u} - \mathbf{f} \cdot \mathbf{v} \\ - \int_{\partial\Omega} \mathbf{g}(T\mathbf{u}) \cdot \mathbf{v} = 0 \quad \forall (\mathbf{v}, q) \in \mathcal{V}_0 \times \mathcal{P}$$

- Slip

$$\mathbf{g}_{\text{slip}}(T\mathbf{u}) = \beta_m(\dots) |T\mathbf{u}|^{m-1} T\mathbf{u}$$

Navier $m = 1$, Weertman $m \approx \frac{1}{3}$, Coulomb $m = 0$.

Newton iteration

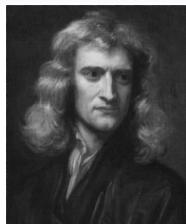
- Standard form of a nonlinear system

$$F(x) = 0$$

- Iteration

$$\text{Solve: } J(x^n)s^n = -F(x^n)$$

$$\text{Update: } x^{n+1} \leftarrow x^n + s^n$$



Stokes problem

$$F(\mathbf{u}, p) \sim \int_{\Omega} \eta D\mathbf{v} : D\mathbf{u} - p \nabla \cdot \mathbf{v} - q \nabla \cdot \mathbf{u} - \mathbf{f} \cdot \mathbf{v} = 0 \quad \forall (\mathbf{v}, q)$$

$$\begin{bmatrix} \mathbf{v} \\ q \end{bmatrix}^T J(\mathbf{w}) \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} \sim \int_{\Omega} (D\mathbf{v})^T [\eta \mathbf{1} + \eta' D\mathbf{w} \otimes D\mathbf{w}] D\mathbf{u} \\ - p \nabla \cdot \mathbf{v} - q \nabla \cdot \mathbf{u}$$

$$J(\mathbf{w}) = \begin{bmatrix} A(\mathbf{w}) & B^T \\ B & \end{bmatrix}$$

Matrices and Preconditioners

Definition (Matrix)

A **matrix** is a linear transformation between finite dimensional vector spaces.

Definition (Forming a matrix)

Forming or assembling a matrix means defining it's action in terms of entries (usually stored in a sparse format).

Definition (Preconditioner)

A preconditioner \mathcal{P} is a method for constructing a matrix (just a linear function, not assembled!) $P^{-1} = \mathcal{P}(\hat{J})$ using information \hat{J} , such that $P^{-1}J$ (or JP^{-1}) has favorable spectral properties.

Left preconditioning in a Krylov iteration

$$(P^{-1}J)x = P^{-1}b$$

$$\{P^{-1}b, (P^{-1}J)P^{-1}b, (P^{-1}J)^2P^{-1}b, \dots\}$$



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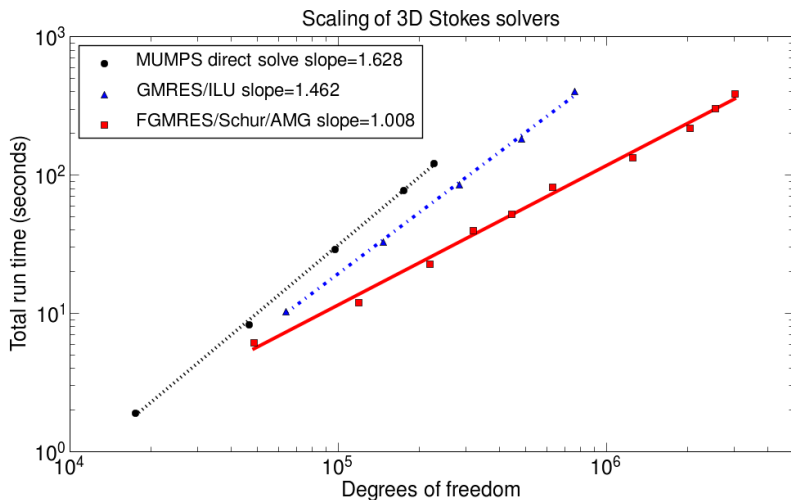
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Normal preconditioners fail for indefinite problems



Stokes

Weak form of the Newton step

Find (\mathbf{u}, p) such that

$$\int_{\Omega} (D\mathbf{v})^T [\eta \mathbf{1} + \eta' D\mathbf{w} \otimes D\mathbf{w}] D\mathbf{u} - p \nabla \cdot \mathbf{v} - q \nabla \cdot \mathbf{u} = -v \cdot F(\mathbf{w}) \quad \forall (\mathbf{v}, q)$$

Matrix

$$\begin{bmatrix} A(\mathbf{w}) & B^T \\ B & \end{bmatrix} \begin{pmatrix} u \\ p \end{pmatrix} = - \begin{pmatrix} F_u(\mathbf{w}) \\ 0 \end{pmatrix}$$

Block factorization

$$\begin{bmatrix} A & B^T \\ B & \end{bmatrix} = \begin{bmatrix} 1 & \\ BA^{-1} & 1 \end{bmatrix} \begin{bmatrix} A & B^T \\ & S \end{bmatrix} = \begin{bmatrix} A & \\ B & S \end{bmatrix} \begin{bmatrix} 1 & A^{-1}B^T \\ & 1 \end{bmatrix}$$

where the Schur complement is

$$S = -BA^{-1}B^T.$$

Properties of the Schur complement

Block factorization

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where

$$S = -BA^{-1}B^T.$$

- S is symmetric negative definite if A is SPD and B has full rank (discrete inf-sup condition)
- S is dense
- We only need to multiply B, B^T with vectors.
- We need preconditioners for A and S .
- Any definite preconditioner can be used for A .
- It's not obvious how to precondition S , more on that later.

Reduced factorizations are sufficient

Theorem (GMRES convergence)

GMRES applied to

$$Kx = b$$

converges in n steps for all right hand sides if the minimal polynomial of K has degree n .

(There exists a polynomial π_n such that $\pi_n(K) = 0$ and $\pi_n(0) = 1$.)

A lower-triangular preconditioner

Left precondition J :

$$\begin{aligned} K = P^{-1}J &= \begin{bmatrix} A & \\ B & S \end{bmatrix}^{-1} \begin{bmatrix} A & B^T \\ B & \end{bmatrix} \\ &= \begin{bmatrix} A^{-1} & \\ -S^{-1}BA^{-1} & S^{-1} \end{bmatrix} \begin{bmatrix} A & B^T \\ B & \end{bmatrix} = \begin{bmatrix} 1 & A^{-1}B^T \\ & 1 \end{bmatrix} \end{aligned}$$

Since $(K - 1)^2 = 0$, GMRES converges in at most 2 steps.

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Preserving symmetry for MINRES

P must be SPD

$$P^{-1} = \begin{bmatrix} A & \\ & -S \end{bmatrix}^{-1}$$

$$K = P^{-1}J = \begin{bmatrix} A^{-1} & \\ & -S^{-1} \end{bmatrix} \begin{bmatrix} A & B^T \\ B & \end{bmatrix} = \begin{bmatrix} 1 & A^{-1}B^T \\ -S^{-1}B & \end{bmatrix}$$

$$\left(K - \frac{1}{2}\right)^2 = \begin{bmatrix} \frac{1}{4} - A^{-1}B^T S^{-1}B & \\ & \frac{5}{4} \end{bmatrix}$$

$$\left(K - \frac{1}{2}\right)^2 - \frac{1}{4} = \begin{bmatrix} -A^{-1}B^T S^{-1}B & \\ & 1 \end{bmatrix}$$

Now $Q = -A^{-1}B^T S^{-1}B$ is a projector ($Q^2 = Q$) so

$$\left[\left(K - \frac{1}{2}\right)^2 - \frac{1}{4}\right]^2 = \left(K - \frac{1}{2}\right)^2 - \frac{1}{4}$$

Rearranging, $K(K-1)(K^2-K-1) = 0$. MINRES converges in at most 3 iterations.

Preconditioning the Schur complement

- $S = -BA^{-1}B^T$ is dense so we can't form it, we need S^{-1} .

Physics-based commutator: anisotropic pressure diffusion

$$\mathbf{v}^T A(\mathbf{w}) \mathbf{u} \sim \int (D\mathbf{v})^T [\eta \mathbf{1} + \eta' D\mathbf{w} \otimes D\mathbf{w}] D\mathbf{u}$$

- We would like to find an operator A_p such that

$$-S = BA^{-1}B^T \approx BB^T A_p^{-1} =: P_S$$

so that

$$P_S^{-1} = A_p(BB^T)^{-1}$$

- Note

$$BB^T \sim (-\nabla \cdot) \nabla = -\Delta$$

corresponds to a Laplacian in the pressure space (multigrid).

- If $\eta', \nabla \eta \ll 1$ then $A_p \sim -\eta \Delta$ so $P_S^{-1} = \eta \mathbf{1}$

Least squares commutator

- Schur complement

$$S = -BA^{-1}B^T$$

Suppose B is square and nonsingular. Then

$$S^{-1} = -B^{-T}AB^{-1}.$$

B is not square, replace B^{-1} with Moore-Penrose pseudoinverse

$$B^\dagger = B^T(BB^T)^{-1}, \quad (B^T)^\dagger = (BB^T)^{-1}B.$$

Then

$$P_S^{-1} = -(BB^T)^{-1}BAB^T(BB^T)^{-1}.$$

- Requires 2 Poisson preconditioners for $(BB^T)^{-1}$ per iteration
- Better with scaling, from mass matrices and effective viscosity (Elman et al. 2006, May & Moresi 2008)

Unsteady Navier-Stokes

Strong form

$$J(\mathbf{w}) \begin{bmatrix} \mathbf{u} \\ p \end{bmatrix} \sim \begin{cases} \rho(\alpha \mathbf{u} + \mathbf{w} \cdot \nabla \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{w}) - \eta \nabla^2 \mathbf{u} + \nabla p = -F(\mathbf{w}) \\ \nabla \cdot \mathbf{u} = 0 \end{cases}$$

Matrix form

$$\begin{bmatrix} A(\mathbf{w}) & B^T \\ B & \end{bmatrix} = \begin{bmatrix} 1 & \\ BA^{-1} & 1 \end{bmatrix} \begin{bmatrix} A & B^T \\ & S \end{bmatrix} \quad S = -BA^{-1}B^T$$

Define $A(\mathbf{w})$ in pressure space

- Want $P_S = (BB^T)A_p^{-1} \approx BA^{-1}B^T$, $P_S^{-1} = A_p(BB^T)^{-1}$
- $A_p \sim \rho(\alpha p + \mathbf{w} \cdot \nabla p + p \operatorname{tr}(\nabla \mathbf{w})) - \eta \nabla^2 p$
- $p \operatorname{tr}(\nabla \mathbf{w})$ term is questionable, not needed for Picard
- Almost mesh-independent, weak Reynolds number dependence

(Silvester, Elman, Kay, Wathen. *Efficient preconditioning of the linearized Navier-Stokes equations for incompressible flow*. 2001) (Elman et al. 2005-2008)

Artificial compressibility

$$\begin{bmatrix} A & B^T \\ B & -C \end{bmatrix} = \begin{bmatrix} 1 & -B^T C^{-1} \\ & 1 \end{bmatrix} \begin{bmatrix} S & \\ & -C \end{bmatrix}$$

where

$$S = A + B^T C^{-1} B.$$

- $C = \epsilon M_p$ corresponds to almost incompressible elasticity
- $B^T C^{-1} B \sim \epsilon^{-1} \nabla(\nabla \cdot \mathbf{u})$
- Must precondition grad-div added system S which becomes singular as $\epsilon \rightarrow 0$
- Some results show weaker Reynolds number dependence than former options

(Dohrmann and others. 2006,2007)

Unsplit schemes

Multigrid

- discretization-dependent smoothers and interpolation

Overlapping (2-level additive Schwarz)

- must ensure that subdomain problems are stable
- definition of coarse level
- Can perform better than split methods ¹

Non-overlapping (2-level BNN, BDDC, FETI-DP)

- More complicated
- Especially robust to jumps in coefficients
- Poorly developed for nonsymmetric problems
- Usual formulation involves exact subdomain and coarse solves

¹Klawonn and Pavarino, *Comparison of overlapping Schwarz methods and block preconditioners for saddle point problems*, 2000

Conclusions

- Indefinite preconditioning is *not* a solved problem
- Large jumps in coefficients present a difficulty
- Mesh-independence is attainable for nearly all classes
- All choices show at least weak Reynolds number dependence

How to take advantage of further advances?

- Provide discrete operators $(A_p, BB^T, \eta^{-1}M_p, \dots)$
- Libraries can abstract the matrix gymnastics
- PETSc's `PCFieldSplit`: generic interface to block relaxation and factorization where Schur complements are (optionally) reinterpreted physically
- Everything should be a runtime option