

A Review of Preconditioning Techniques for Steady Incompressible Flow

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Outline

- PDEs
- Review : 1984 – 2005
- Update : 2005 – 2009

Outline

- PDEs



$$\left. \begin{aligned} \vec{u} \cdot \nabla \vec{u} - \nu \nabla^2 \vec{u} + \nabla p &= 0 \\ \nabla \cdot \vec{u} &= 0 \end{aligned} \right\} \text{Navier--Stokes}$$

Outline

- PDEs



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$$\left. \begin{array}{l} -\nabla^2 \vec{u} + \nabla p = 0 \\ \nabla \cdot \vec{u} = 0 \end{array} \right\} \text{Stokes}$$

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- PDEs

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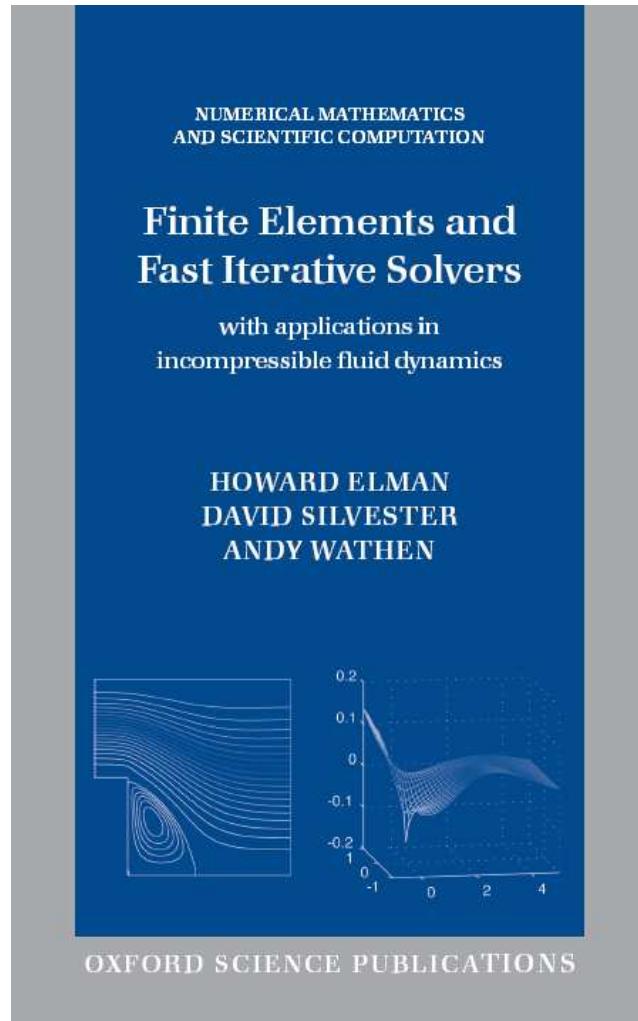
$$\left. \begin{array}{l} \mathcal{A}^{-1} \vec{u} + \nabla p = 0 \\ \nabla \cdot \vec{u} = 0 \end{array} \right\} \text{Darcy}$$

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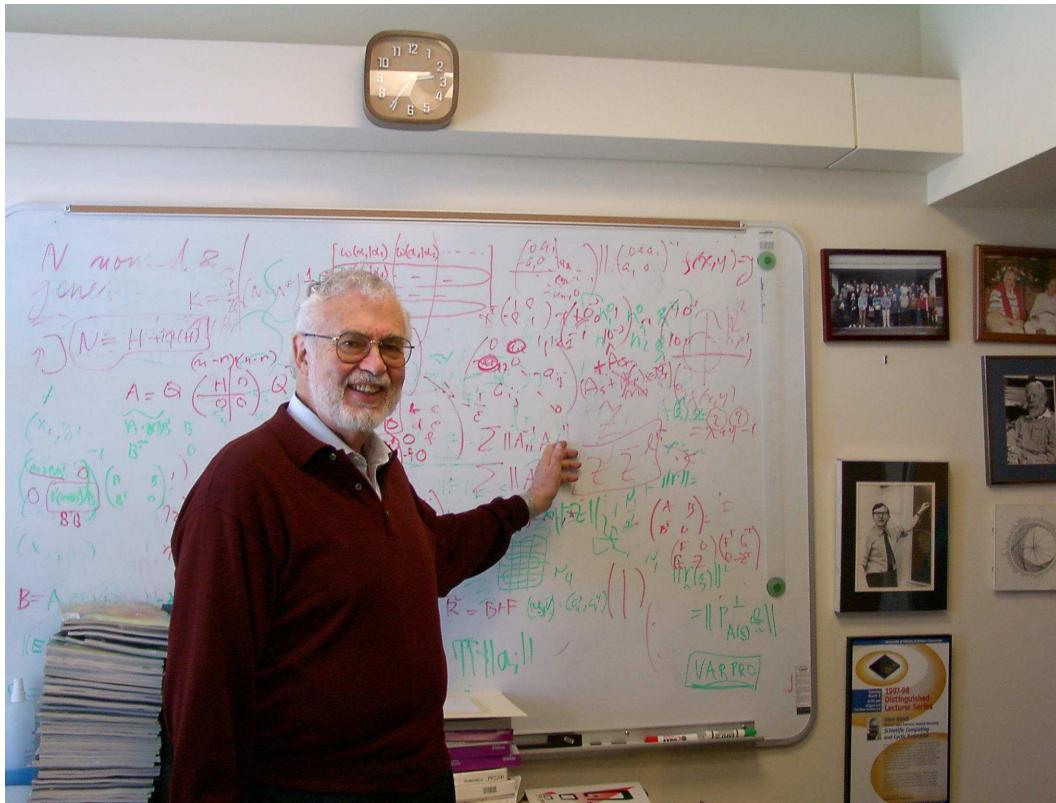
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Numerical solution of saddle point problems

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We dedicate this paper to Gil Strang on the occasion of his 70th birthday

Large linear systems of saddle point type arise in a wide variety of applications throughout computational science and engineering. Due to their indefiniteness and often poor spectral properties, such linear systems represent a significant challenge for solver developers. In recent years there has been a surge of interest in saddle point problems, and numerous solution techniques have been proposed for this type of system. The aim of this paper is to present and discuss a large selection of solution methods for linear systems in saddle point form, with an emphasis on iterative methods for large and sparse problems.

Steady-state Navier-Stokes equations

$$\begin{aligned}\vec{u} \cdot \nabla \vec{u} - \nu \nabla^2 \vec{u} + \nabla p &= 0 && \text{in } \Omega \\ \nabla \cdot \vec{u} &= 0 && \text{in } \Omega.\end{aligned}$$

Boundary conditions:

$$\vec{u} = \vec{w} \text{ on } \partial\Omega_D, \quad \nu \frac{\partial \vec{u}}{\partial n} - \vec{n}p = \vec{0} \text{ on } \partial\Omega_N.$$

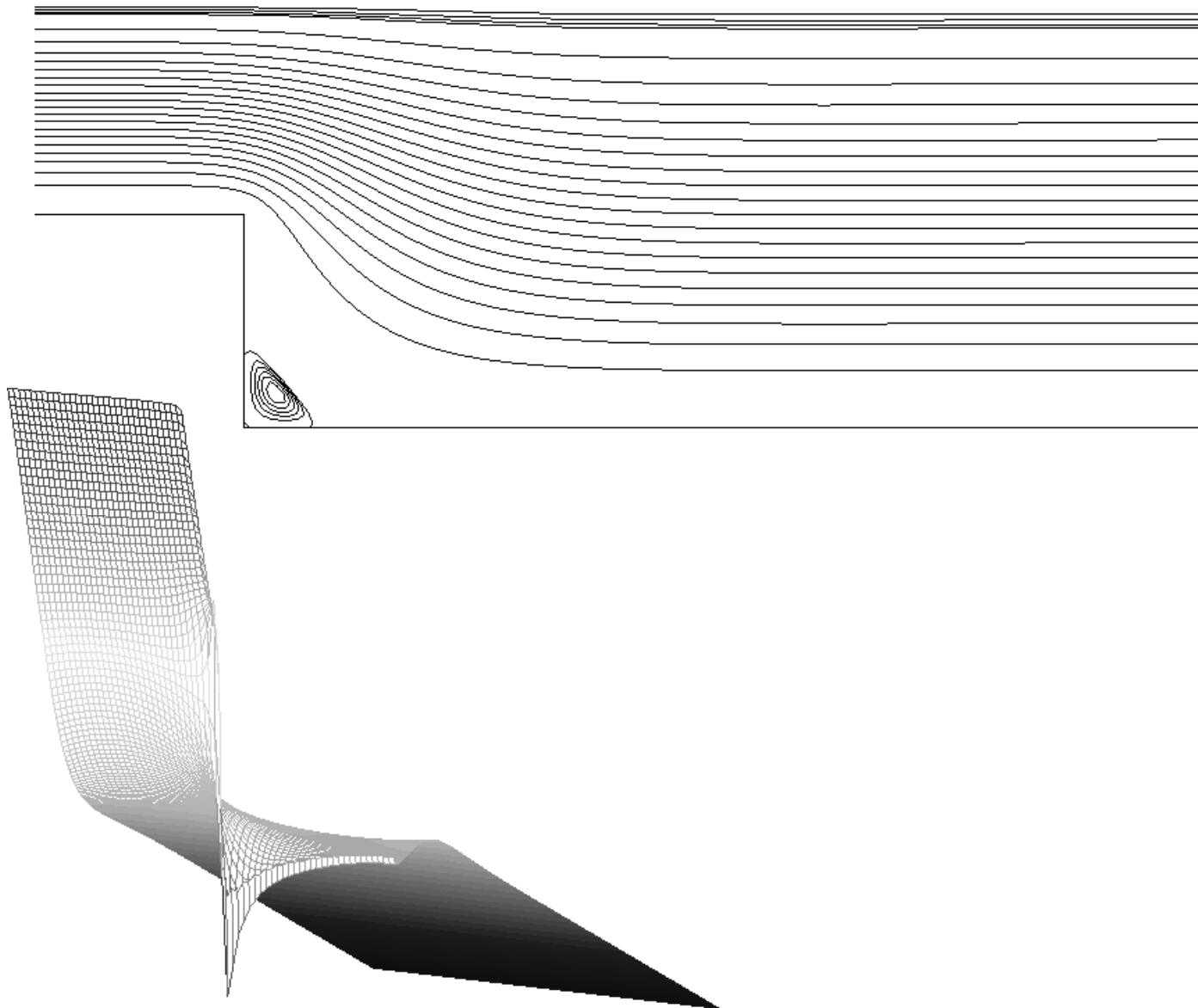
Picard linearization:

Given \vec{u}^0 , compute $\vec{u}^1, \vec{u}^2, \dots, \vec{u}^k$ via

$$\begin{aligned}\vec{u}^k \cdot \nabla \vec{u}^{k+1} - \nu \nabla^2 \vec{u}^{k+1} + \nabla p^{k+1} &= 0, \\ \nabla \cdot \vec{u}^{k+1} &= 0 && \text{in } \Omega\end{aligned}$$

together with appropriate boundary conditions.

Example: Flow over a Step



Finite element matrix formulation

Introducing the basis sets

$$\mathbf{X}_h = \text{span}\{\vec{\phi}_i\}_{i=1}^{n_u}, \quad \text{Velocity basis functions};$$
$$M_h = \text{span}\{\psi_j\}_{j=1}^{n_p}, \quad \text{Pressure basis functions}.$$

gives the discretized system:

$$\begin{pmatrix} N + \nu A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ p \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{g} \end{pmatrix},$$

with associated matrices

$$N_{ij} = (\mathbf{w}_h \cdot \nabla \vec{\phi}_i, \vec{\phi}_j), \quad \text{convection}$$

$$A_{ij} = (\nabla \vec{\phi}_i, \nabla \vec{\phi}_j), \quad \text{diffusion}$$

$$B_{ij} = -(\nabla \cdot \vec{\phi}_j, \psi_i), \quad \text{divergence} .$$

In the Stokes limit $\nu \rightarrow \infty$, we obtain the **saddle-point problem**

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

with $A \in \mathbb{R}^{n_u \times n_u}$, $B \in \mathbb{R}^{n_p \times n_u}$, and for a **stable** approximation

- $A = A^T$, $\mathbf{u}^T A \mathbf{u} > 0$, $\forall \mathbf{u} \neq 0$,
- $\text{rank}(B) = n_p$ if $\partial\Omega_N \neq \emptyset$, otherwise $\text{null}(B^T) = \{1\}$.

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- $\text{rank}(B) = n_p$ if $\partial\Omega_N \neq \emptyset$, otherwise $\text{null}(B^T) = \{1\}$.

Note that the Stokes system matrix is “highly indefinite”:

$$\mathcal{A} = \begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} = \begin{pmatrix} I & 0 \\ BA^{-1} & I \end{pmatrix} \begin{pmatrix} A & 0 \\ 0 & -BA^{-1}B^T \end{pmatrix} \begin{pmatrix} I & A^{-1}B^T \\ 0 & I \end{pmatrix}$$

where $S = BA^{-1}B^T$ is the **Schur complement** of A in \mathcal{A} .

A Simple Iterative Solver ...

Given scalar parameter α

Algorithm: Arrow, Hurwicz, Uzawa, 1958

for $k = 0, 1, \dots$

solve $A\mathbf{u}_{k+1} = \mathbf{f} - B^T \mathbf{p}_k$

compute $\mathbf{p}_{k+1} = \mathbf{p}_k + \alpha B\mathbf{u}_{k+1}$

end

A Simple Iterative Solver ...

Given scalar parameter α

Algorithm: Arrow, Hurwicz, Uzawa, 1958

```
for k = 0, 1, ...
    solve      Auk+1 = f - BTpk
    compute    pk+1 = pk + αBTuk+1
end
```

Eliminating u_{k+1} gives

$$p_{k+1} = p_k + \alpha(BA^{-1}f - BA^{-1}B^T p_k).$$

This is simple (first order) Richardson iteration for the Schur Complement problem

$$BA^{-1}B^T p = BA^{-1}f.$$

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984

Idea I : Introduce the pressure mass matrix $\mathcal{Q} \sim I$ and use an iterated penalty approach:

for $k = 0, 1, \dots$

$$A\mathbf{u}_{k+1} + B^T \mathbf{p}_{k+1} = \mathbf{f}$$

$$B\mathbf{u}_{k+1} - \epsilon \mathcal{Q} \mathbf{p}_{k+1} = -\epsilon \mathcal{Q} \mathbf{p}_k$$

end

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$$B\mathbf{u}_{k+1} - \epsilon \mathbf{Q} \mathbf{p}_{k+1} = -\epsilon \mathbf{Q} \mathbf{p}_k$$

end

Eliminating p_{k+1} gives

$$(A + \frac{1}{\epsilon} B^T \mathbf{Q}^{-1} B) \mathbf{u}_{k+1} = \mathbf{f} - B^T \mathbf{p}_k$$
$$\mathbf{p}_{k+1} = \mathbf{p}_k + \frac{1}{\epsilon} \mathbf{Q}^{-1} B \mathbf{u}_{k+1}$$

- Michel Fortin & Roland Glowinski
Augmented Lagrangian Methods
North-Holland, 1984.

Idea II : Apply CG to the Schur Complement system, but replace inversion of A with n_A multigrid V-cycles A_*^{-1} :

$$BA_*^{-1}B^T \mathbf{p} = BA^{-1}\mathbf{f}.$$

- Rudiger Verfürth
A Combined Conjugate Gradient-Multigrid algorithm for
the Numerical Solution of the Stokes Problem
IMA J. Numer. Anal., 4, 1984.

where it is established that this approach leads to “textbook”
(grid independent) convergence if n_A is sufficiently big.

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984
- 1991

Silvester & Wathen : A Fledgling-Fast Solver

We introduce the pressure mass matrix Q and solve the (symmetric-) system $\mathcal{A}x = f$

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix}$$

with the **Conjugate Residual** method and the diagonal preconditioning

$$\mathcal{P}^{-1} = \begin{pmatrix} D_A^{-1} & 0 \\ 0 & D_Q^{-1} \end{pmatrix}.$$

Analysis is elegant ...

Given

- ▷ “inf–sup” stability:

$$\gamma^2 \leq \frac{\mathbf{p}^T B A^{-1} B^T \mathbf{p}}{\mathbf{p}^T Q \mathbf{p}} \leq \Gamma^2 \quad \forall \mathbf{p} \in \mathbb{R}^{n_p}$$

- ▷ the mass matrix preconditioner satisfies

$$\theta^2 \leq \frac{\mathbf{p}^T Q \mathbf{p}}{\mathbf{p}^T D_Q \mathbf{p}} \leq \Theta^2, \quad \forall \mathbf{p} \in \mathbb{R}^{n_p}$$

- ▷ the vector Laplacian satisfies:

$$g(h) \leq \frac{\mathbf{u}^T A \mathbf{u}}{\mathbf{u}^T D_A \mathbf{u}} \leq 1 \quad \forall \mathbf{u} \in \mathbb{R}^{n_u}$$

with $g(h) \rightarrow 0$ as $h \rightarrow 0$.

if $g(h) \rightarrow 0$ as $h \rightarrow 0$, then the eigenvalues of the preconditioned Stokes operator lie in the union of two real intervals

$$\left[-\Gamma^2 \Theta^2, -\gamma \theta \sqrt{g(h)} \right] \cup \left[g(h), 1 + \Gamma^2 \Theta^2 \right]$$

In particular, using simple diagonal scaling $g(h) = O(h^2)$, and the eigenvalues lie in

$$[-a, -bh] \cup [ch^2, d].$$

- Andrew Wathen & David Silvester
Fast iterative solution of stabilised Stokes systems
Part I: Using simple diagonal preconditioners,
Report NA-91-04, Stanford University, 1991.
SIAM J. Numer. Anal., 30, 1993.

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984
- 1991
- 1992

Silvester & Wathen : A Fast Solver

We solve the (symmetric–) system $\mathcal{A}x = f$

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix}$$

using MINRES (due to Bernd Fischer) with the spectrally equivalent preconditioning

$$\mathcal{P}^{-1} = \begin{pmatrix} A_*^{-1} & 0 \\ 0 & D_Q^{-1} \end{pmatrix},$$

so that the vector Laplacian satisfies

$$\lambda^2 \leq \frac{\mathbf{u}^T A \mathbf{u}}{\mathbf{u}^T A_* \mathbf{u}} \leq 1 \quad \forall \mathbf{u} \in \mathbb{R}^{n_u}$$

with $\lambda \geq \lambda_* > 0$ as $h \rightarrow 0$.

In this case the eigenvalues of the preconditioned Stokes operator lie in the union of two real intervals

$$\left[-\Gamma^2 \Theta^2, \frac{1}{2}(\lambda - \sqrt{\lambda^2 + 4\gamma^2\theta^2\lambda}) \right] \cup [\lambda, 1 + \Gamma^2 \Theta^2]$$

giving a convergence rate which is bounded away from one independently of the grid.

- Torgeir Rusten & Ragner Winther
[A Preconditioned Iterative Method for Saddle-Point Problems](#), SIAM J. Matrix. Anal. Appl., 13, **1992**.
- Andrew Wathen & David Silvester
[Fast iterative solution of stabilised Stokes systems Part II: Using general block preconditioners](#), Report NA-**92**-xx, Stanford University.
SIAM J. Numer. Anal., 31, **1994**.

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984
- 1991
- 1992
- 1993

Gene's Idea: Use efficient inexact solve (e.g. multigrid V-cycle)

for $k = 0, 1, \dots$

$$A\mathbf{u}_{k+1} = \mathbf{f} - B^T \mathbf{p}_k + \delta_k$$

$$\mathbf{p}_{k+1} = \mathbf{p}_k + \alpha B\mathbf{u}_{k+1}$$

end

with stopping criterion $\|\delta_k\| \leq \tau \|B\mathbf{u}_k\|$.

- Howard Elman & Gene Golub
Inexact and Preconditioned Uzawa Algorithms for
Saddle Point Problems
SIAM J. Numer. Anal., 31, 1994.

where it is established that this approach leads to
“textbook” (grid independent) convergence.

$$\begin{pmatrix} \textcolor{red}{F} & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984
- 1991
- 1992
- 1993
- **1994**

Elman & Silvester : A Fledgling-Fast Solver

Now for the Navier-Stokes Equations. We solve the nonsymmetric system $\mathcal{F}x = f$

$$\begin{pmatrix} N + \nu A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ p \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ 0 \end{pmatrix}$$

using GMRES with the spectrally equivalent preconditioning

$$\mathcal{P} = \begin{pmatrix} F_* & 0 \\ 0 & D_Q \end{pmatrix},$$

so that the vector convection-diffusion operator F_* satisfies

$$\lambda^2 \leq \frac{\mathbf{u}^T (N + \nu A) \mathbf{u}}{\mathbf{u}^T F_* \mathbf{u}} \leq \Lambda^2 \quad \forall \mathbf{u} \in \mathbb{R}^{n_u}.$$

In this case the eigenvalues of the preconditioned Oseen operator lie in two boxes on either side of the imaginary axis in the complex plane.

- Gives a convergence rate which is bounded away from one **independently of the grid**.

In this case the eigenvalues of the preconditioned Oseen operator lie in two boxes on either side of the imaginary axis in the complex plane.

- Gives a convergence rate which is bounded away from one **independently of the grid**.
- Convergence deteriorates as $\nu \rightarrow 0$ because of low-quality Schur complement approximation

$$B(N + \nu A)^{-1} B^T \sim D_Q.$$

- Howard Elman & David Silvester
Fast Nonsymmetric Iterations and Preconditioning for Navier-Stokes Equations,
Report UMIACS-TR-94-66, University of Maryland.
SIAM J. Sci. Comput, 17, 1996.

Block triangular preconditioning

$$\begin{pmatrix} F & B^T \\ B & 0 \end{pmatrix} \mathcal{P}^{-1} \quad \mathcal{P} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix}$$

A **perfect** preconditioner is given by

$$\begin{pmatrix} F & B^T \\ B & 0 \end{pmatrix} \underbrace{\begin{pmatrix} F^{-1} & F^{-1}B^T S^{-1} \\ 0 & -S^{-1} \end{pmatrix}}_{\mathcal{P}^{-1}} = \begin{pmatrix} I & 0 \\ BF^{-1} & I \end{pmatrix}$$

Here $F = N + \nu A$ and $S = BF^{-1}B^T$.

Block triangular preconditioning

$$\begin{pmatrix} F & B^T \\ B & 0 \end{pmatrix} \mathcal{P}^{-1} \quad \mathcal{P} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix}$$

A **perfect** preconditioner is given by

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Here $F = N + \nu A$ and $S = BF^{-1}B^T$. Note that

$$\underbrace{\begin{pmatrix} F^{-1} & F^{-1}B^T S^{-1} \\ 0 & -S^{-1} \end{pmatrix}}_{\mathcal{P}^{-1}} \underbrace{\begin{pmatrix} F & B^T \\ 0 & -S \end{pmatrix}}_{\mathcal{P}} = \begin{pmatrix} I & 0 \\ 0 & I \end{pmatrix}$$

$$\begin{pmatrix} \textcolor{red}{F} & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

- 1984
- 1991
- 1992
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- 1999

A Fast Solver

Given

$$\begin{pmatrix} N + \nu A & B^T \\ B & 0 \end{pmatrix} \mathcal{P}^{-1} \quad \mathcal{P} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{g} \end{pmatrix}$$

With discrete matrices

$$N_{ij} = (\vec{w}_h \cdot \nabla \vec{\phi}_i, \vec{\phi}_j), \quad \text{convection}$$

$$A_{ij} = (\nabla \vec{\phi}_i, \nabla \vec{\phi}_j), \quad \text{diffusion}$$

$$B_{ij} = -(\nabla \cdot \vec{\phi}_j, \psi_i), \quad \text{divergence}$$

For an **efficient** block diagonal (or **triangular**) preconditioner \mathcal{P} we require a sparse approximation to the “exact” Schur complement

$$\mathcal{S}^{-1} = (B(N + \nu A)^{-1}B^T)^{-1} =: (BF^{-1}B^T)^{-1}$$

Schur complement approximation – I

Introducing associated pressure matrices

$$A_p \sim (\nabla \psi_i, \nabla \psi_j), \quad \text{diffusion}$$

$$N_p \sim (\vec{w}_h \cdot \nabla \psi_i, \psi_j), \quad \text{convection}$$

$$\mathcal{F}_p = \nu A_p + N_p, \quad \text{convection-diffusion}$$

gives the “pressure convection-diffusion preconditioner”:

$$(BF^{-1}B^T)^{-1} \approx Q^{-1} \underbrace{\mathcal{F}_p}_{\text{AMG}} \underbrace{A_p^{-1}}_{\text{AMG}}$$

- David Kay & Daniel Loghin (& Andy Wathen)
A Green’s function preconditioner for the steady-state
Navier-Stokes equations
Report NA-99/06, Oxford University Computing Lab.
SIAM J. Sci. Comput, 24, 2002.

Schur complement approximation – II

Introducing the diagonal of the velocity mass matrix

$$M_* \sim M_{ij} = (\vec{\phi}_i, \vec{\phi}_j),$$

gives the “least-squares commutator preconditioner”:

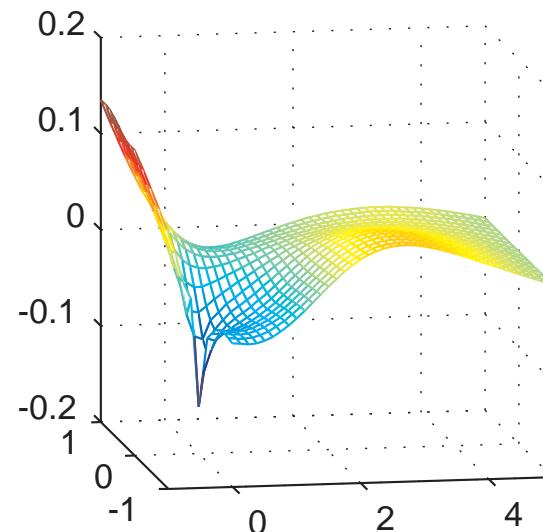
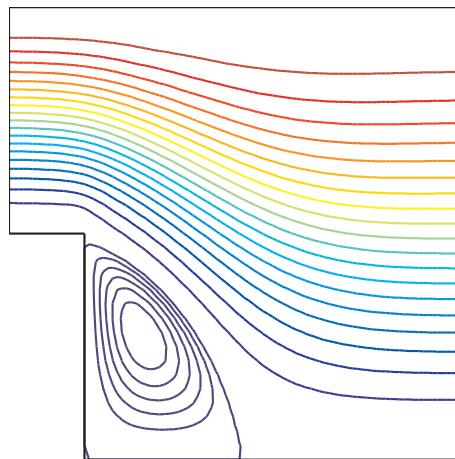
$$(BF^{-1}B^T)^{-1} \approx \underbrace{(BM_*^{-1}B^T)^{-1}}_{\text{AMG}} (BM_*^{-1}FM_*^{-1}B^T) \underbrace{(BM_*^{-1}B^T)^{-1}}_{\text{AMG}}$$

- Howard Elman (& Ray Tuminaro et al.)
Preconditioning for the steady-state Navier-Stokes equations with low viscosity,
SIAM J. Sci. Comput, 20, 1999.
Block preconditioners based on approximate commutators,
SIAM J. Sci. Comput, 27, 2006.

$$\begin{pmatrix} \textcolor{red}{F} & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{f} \\ \mathbf{0} \end{pmatrix},$$

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IFISS2.2 computational results



Final step of Oseen iteration : $R = 2/\nu$

GMRES iterations using Q_2-Q_1 (Q_1-P_0) — tol = 10^{-6}
Exact Least Squares Commutator

$1/h$	$R = 10$	$R = 100$	$R = 200$
5	15 (15)	17 (16)	
6	19 (21)	21 (22)	29 (32)
7	23 (31)	29 (32)	29 (30)

- Update : 2005 – 2009
- Howard Elman & Alison Ramage & David Silvester,
Algorithm 866 : IFISS, a Matlab toolbox for modelling
incompressible flow,
ACM Trans. Math. Soft. 33, 2007.
- Howard Elman (& Silvester et al.)
Least squares preconditioners for stabilized
discretizations of the Navier-Stokes equations,
SIAM J. Sci. Comput, 30, 2007.
- Howard Elman & Ray Tuminaro
Boundary conditions in approximate commutator
preconditioners for the Navier-Stokes equations,
Report UMIACS-TR-2009-02, University of Maryland.

Darcy Flow Equations

Given boundary “pressure head” data g , and an isotropic permeability matrix $\mathcal{A} = \mu I_2$, such that

$$0 < \mu_* \leq \mu(\vec{x}) \leq \mu^* < \infty \quad \forall \vec{x} \in D \subset \mathbb{R}^2 :$$

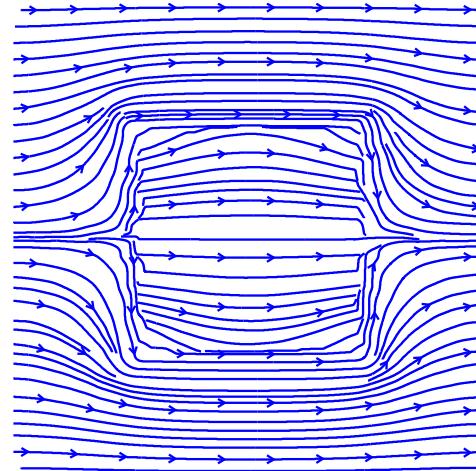
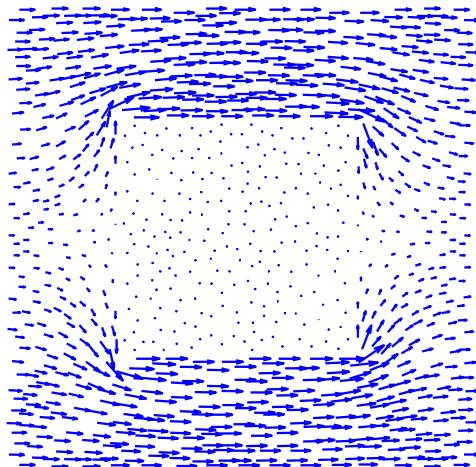
we want to compute the pair (\vec{u}, p) such that

$$\mathcal{A}^{-1} \vec{u} + \nabla p = 0 \quad \text{in } D,$$

$$\nabla \cdot \vec{u} = 0 \quad \text{in } D,$$

$$p = g \quad \text{on } \Gamma_D; \quad \vec{u} \cdot \vec{n} = 0 \quad \text{on } \Gamma_N.$$

Example: Groundwater Flow



Finite element matrix formulation

Introducing the basis sets

$$\mathbf{X}_h = \text{span}\{\vec{\phi}_i\}_{i=1}^{n_u}, \quad \text{Flux basis functions;}$$

$$M_h = \text{span}\{\psi_j\}_{j=1}^{n_p}; \quad \text{Pressure basis functions.}$$

gives the discretized system

$$\begin{pmatrix} M & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{g} \\ \mathbf{0} \end{pmatrix},$$

with associated matrices:

$$M_{ij} = (\mu^{-1} \vec{\phi}_i, \vec{\phi}_j), \quad \text{local mass}$$

$$B_{ij} = -(\nabla \cdot \vec{\phi}_j, \psi_i), \quad \text{divergence}$$

Fast Saddle-Point Solver

We solve the (symmetric–) system $\mathcal{L}x = f$

$$\begin{pmatrix} M & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \begin{pmatrix} \mathbf{g} \\ \mathbf{0} \end{pmatrix}$$

with $M \in \mathbb{R}^{n_u \times n_u}$, and $B \in \mathbb{R}^{n_p \times n_u}$, using MINRES with the block diagonal preconditioning

$$\mathcal{P}^{-1} = \begin{pmatrix} M_*^{-1} & 0 \\ 0 & Q_*^{-1} \end{pmatrix}.$$

Given that the blocks M_* (mass matrix diagonal) and Q_* (via [AMG](#)) satisfy

$$\gamma^2 \leq \frac{\mathbf{u}^T M \mathbf{u}}{\mathbf{u}^T M_* \mathbf{u}} \leq \Gamma^2 \quad \forall \mathbf{u} \in \mathbb{R}^{n_u},$$

$$\theta^2 \leq \frac{\mathbf{p}^T B M_*^{-1} B^T \mathbf{p}}{\mathbf{p}^T Q_* \mathbf{p}} \leq \Theta^2 \quad \forall \mathbf{p} \in \mathbb{R}^{n_p},$$

then the eigenvalues of the preconditioned problem,

$$\begin{pmatrix} M & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix} = \lambda \begin{pmatrix} M_* & 0 \\ 0 & Q_* \end{pmatrix} \begin{pmatrix} \mathbf{u} \\ \mathbf{p} \end{pmatrix},$$

lie in the union of intervals that are bounded away from zero and $\pm\infty$, independently of h and μ .

- Catherine Powell & David Silvester
Optimal Preconditioning for Raviart-Thomas Mixed
Formulation of Second-Order Elliptic Problems
SIAM J. Matrix Anal. Appl., 25, 2004.
- David Silvester & Catherine Powell
PIFISS Potential (Incompressible) Flow & Iterative
Solution Software guide,
MIMS Eprint 2007.14.

- Catherine Powell & David Silvester
Optimal Preconditioning for Raviart-Thomas Mixed
Formulation of Second-Order Elliptic Problems
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