

Parallel System Solvers for the Navier-Stokes Equations

Murat Manguoglu

Ahmed Sameh

Faisal Saied

Purdue University

In collaboration with Tayfun Tezduyar, Rice University

PMAA , IRISA, Rennes, France, September 7- 9, 2006

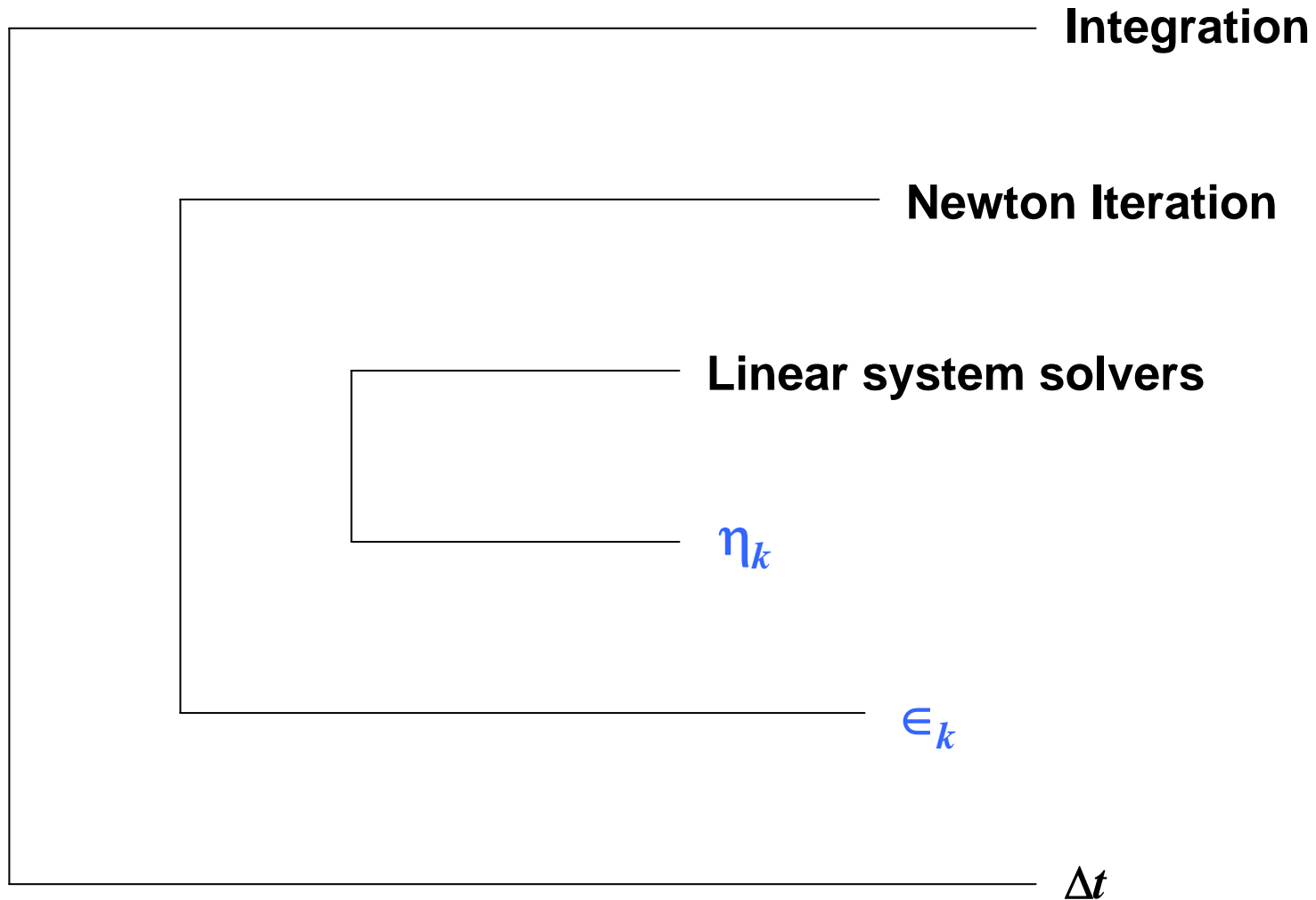
Outline

1. Present preconditioned schemes for solving the resulting “**segregated**” and “**reordered**” linear systems.
2. Use for the stabilized and unstabilized finite element formulations.
3. Effectiveness of solvers demonstrated on model problems in –
 - *time-accurate solutions in fluid-structure/solid interaction*
 - *steady-state case*
4. Concluding remarks.

Applications

- **Space-time flow computation:**
 - **a simplified parachute model**
 - **flow past a hemisphere.**
- **Particulate flow**
- **Incompressible fluid flow within a “leaky” lid-driven cavity problem steady-state case.**

Adaptive Tolerances



Linear Systems

$$\begin{bmatrix} A & B \\ C^T & G \end{bmatrix} \begin{bmatrix} u \\ p \end{bmatrix} = \begin{bmatrix} f \\ g \end{bmatrix}$$

- $A \neq A^T \quad := (n \times n), \quad \text{rank} = n$
- $B, C \quad := (n \times m), \quad \text{maximal col. rank} \leq m$
- $G \quad := \text{symmetric +ve semi-definite}$
- $m < n$

Preconditioned Scheme

Solve
$$\begin{pmatrix} A & B \\ C^T & G \end{pmatrix} \begin{pmatrix} u \\ p \end{pmatrix} = \begin{pmatrix} f \\ g \end{pmatrix}$$

via a Krylov subspace method

Preconditioner

$$M = \begin{pmatrix} E & B \\ C^T & G \end{pmatrix}$$

Solve $Mz = r$ via Richardson scheme

$$z_{k+1} = z_k + \hat{M}^{-1}(r - Mz_k)$$

where,

$$\hat{M} = \begin{pmatrix} \hat{A} & B \\ C^T & G \end{pmatrix}$$

Solving $\hat{M}u = c$

$$\text{Solve } \begin{pmatrix} \hat{A} & B \\ C^T & G \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \mathbf{w} \end{pmatrix} = \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix}$$

Since

$$\begin{pmatrix} \hat{A} & B \\ C^T & G \end{pmatrix} = \begin{pmatrix} \hat{A} & 0 \\ C^T & I \end{pmatrix} \begin{pmatrix} \hat{A}^{-1} & 0 \\ 0 & G - C^T \hat{A}^{-1} B \end{pmatrix} \begin{pmatrix} \hat{A} & B \\ 0 & I \end{pmatrix}$$

Then solution scheme is given by,

- solve $\hat{A}\mathbf{v} = \mathbf{a}$
- solve $(G - C^T \hat{A}^{-1} B)\mathbf{w} = \mathbf{b} - C^T \mathbf{v}$
- solve $\hat{A}\mathbf{v} = \mathbf{a} - B\mathbf{w}$

-
- **Solve** $\hat{A}v = h$
 - \hat{A} = approximate *LU* factorization of E , or
 - \hat{A}^{-1} : **approximate sparse inverse of E**
e.g., $\min \left\| I - E \hat{A}^{-1} \right\|_F$ for a given sparsity structure of \hat{A}^{-1} .
 - *Action of \hat{A}^{-1} is realized via iterative solvers involving E .*
 - **Solve** $(G - C^T \hat{A}^{-1} B)y = h$

using a Krylov subspace method with a **relaxed** stopping criterion

$$\left\| \mathbf{r}_k \right\|_2 / \left\| \mathbf{r}_0 \right\|_2 < \text{tol.}$$

Convergence

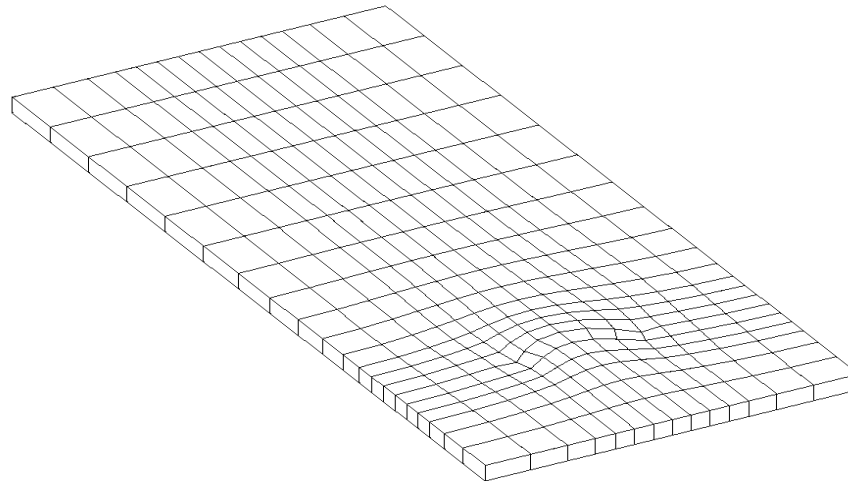
Need to have $\rho(I - \hat{M}^{-1}M) < 1$

- $\lambda(I - \hat{M}^{-1}M) := 0, \mu$
- $\mu = \lambda[(I + K)(I - \hat{A}^{-1}E)]$
- $K = \hat{A}^{-1}B(G - C^T \hat{A}^{-1}B)^{-1}C^T$
 $\lambda(K) = \lambda[(C^T \hat{A}^{-1}B)(G - C^T \hat{A}^{-1}B)^{-1}]$
- $\rho(I - \hat{A}^{-1}E) \leq \alpha < 1$

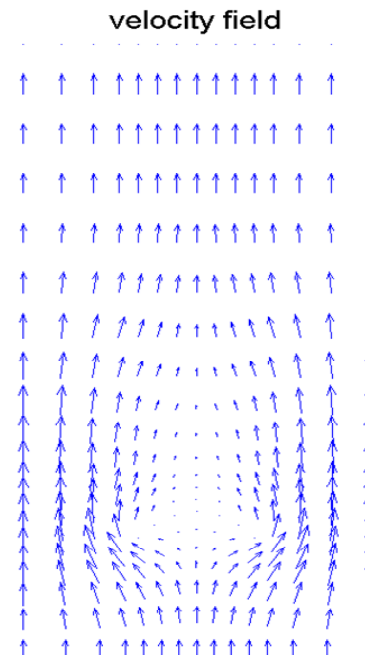
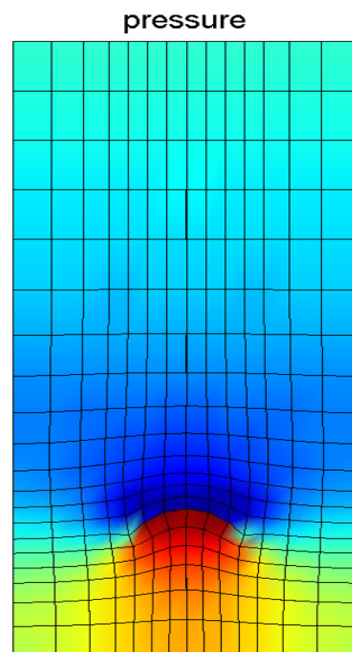
Case 1

Space-Time Computation

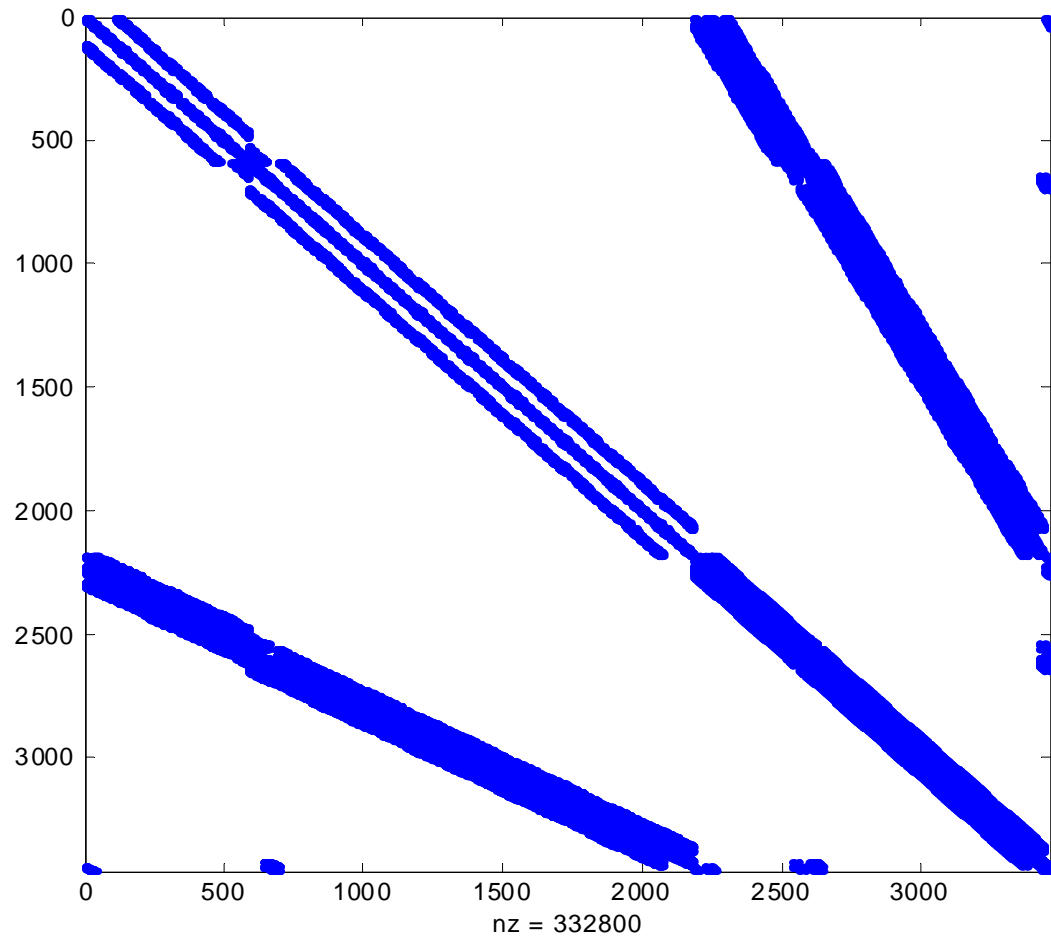
1. Simplified Parachute Model

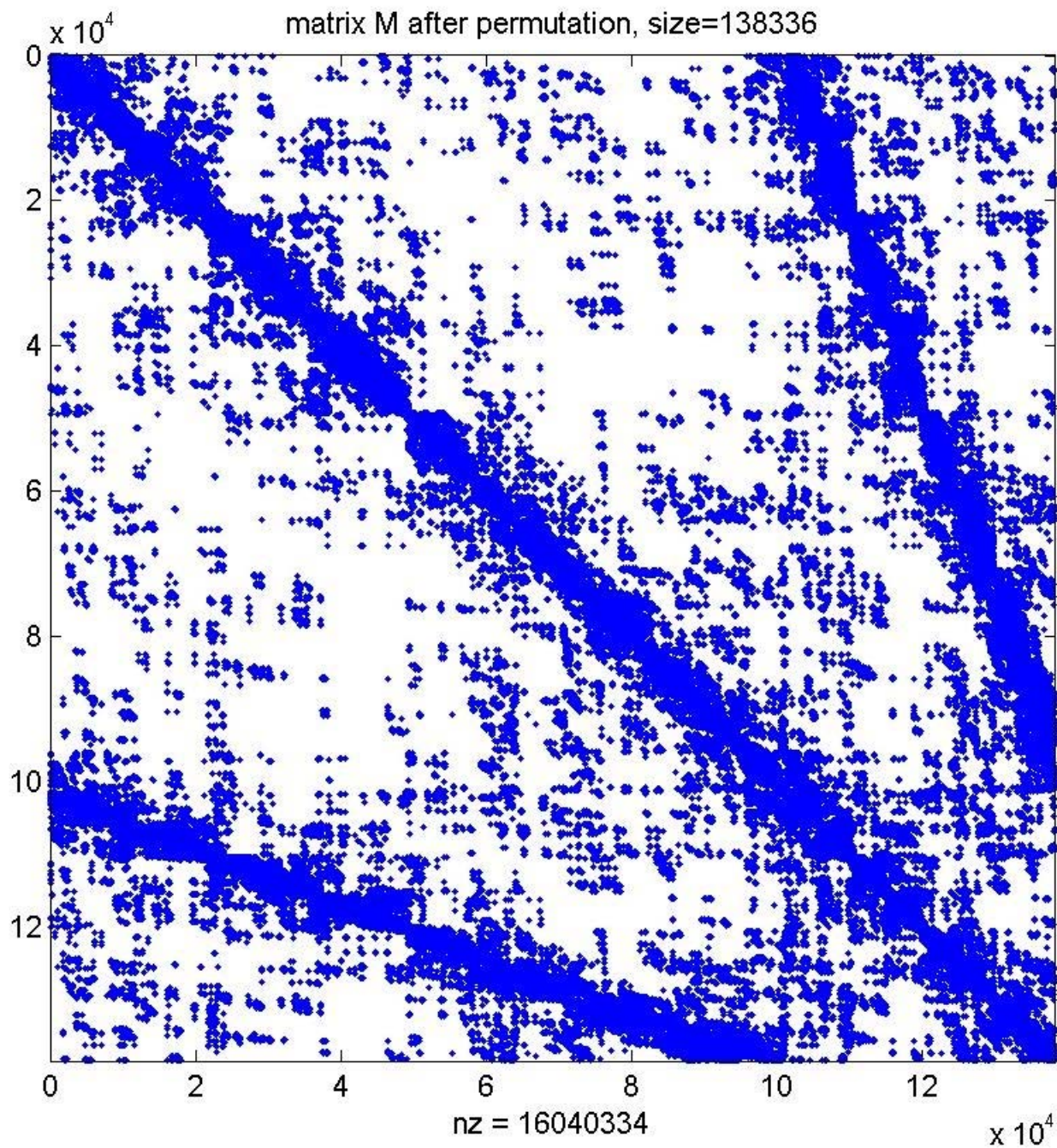


Flow Results:



The Linear System Matrix \tilde{A}





Parachute model...

- $Re = 1000$
- Choose $E = A$ and remove outer layer.
- $\hat{A} = LU$
incomplete LU-factorization of A with
no fill-in, or any of its parallel
counterparts.

Solve
$$\begin{bmatrix} A & B \\ C^T & G \end{bmatrix} \begin{bmatrix} \mathbf{u} \\ \mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

via Richardson scheme: $\|r_{Richardson}\|_2 \leq 10^{-5}$.

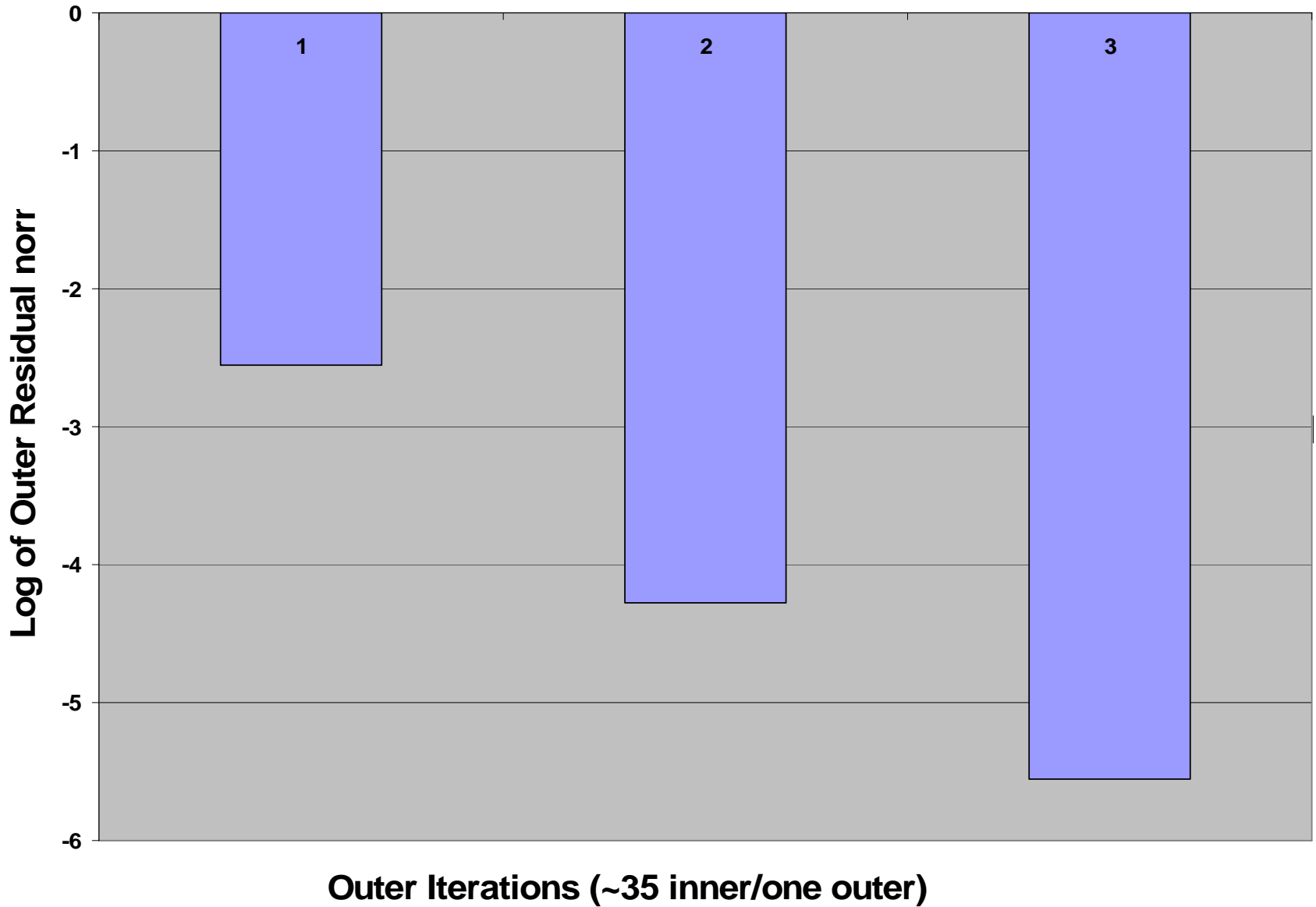
•
$$\hat{M} = \begin{bmatrix} LU & B \\ C^T & G \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{u}_{k+1} \\ \mathbf{p}_{k+1} \end{bmatrix} = \begin{bmatrix} \mathbf{u}_k \\ \mathbf{p}_k \end{bmatrix} + \hat{M}^{-1} \left[\begin{pmatrix} \mathbf{f} \\ \mathbf{g} \end{pmatrix} - \begin{pmatrix} A & B \\ C^T & G \end{pmatrix} \begin{pmatrix} \mathbf{u}_k \\ \mathbf{p}_k \end{pmatrix} \right]$$

Note:

Solve $(G - C^T \hat{A}^{-1} B) \mathbf{w} = \mathbf{s}$ via bicgstab with relative residual $< 10^{-1}$.
G1 := preconditioner

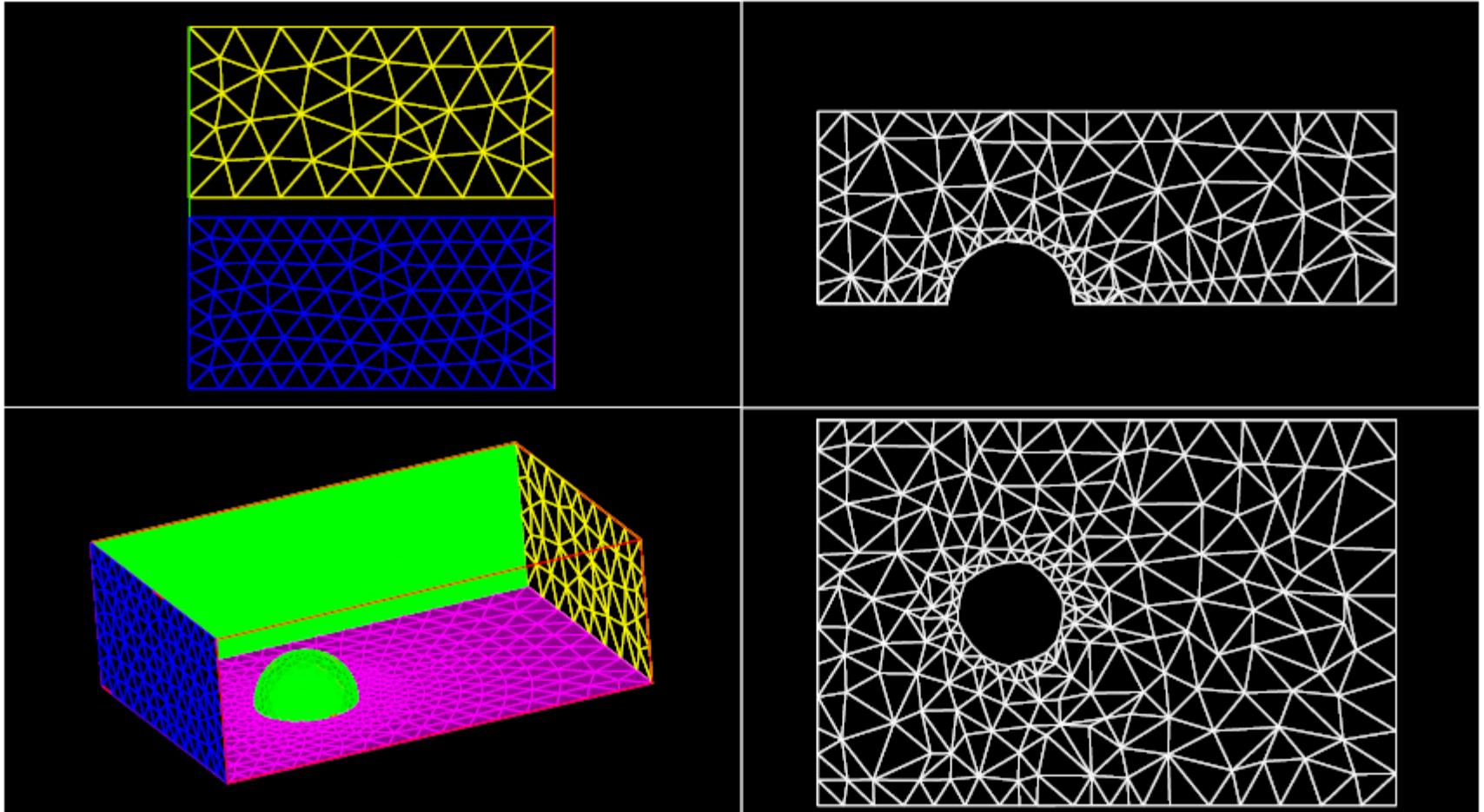
Richardson: Incomplete LU - Bicgstab (10^{-1})



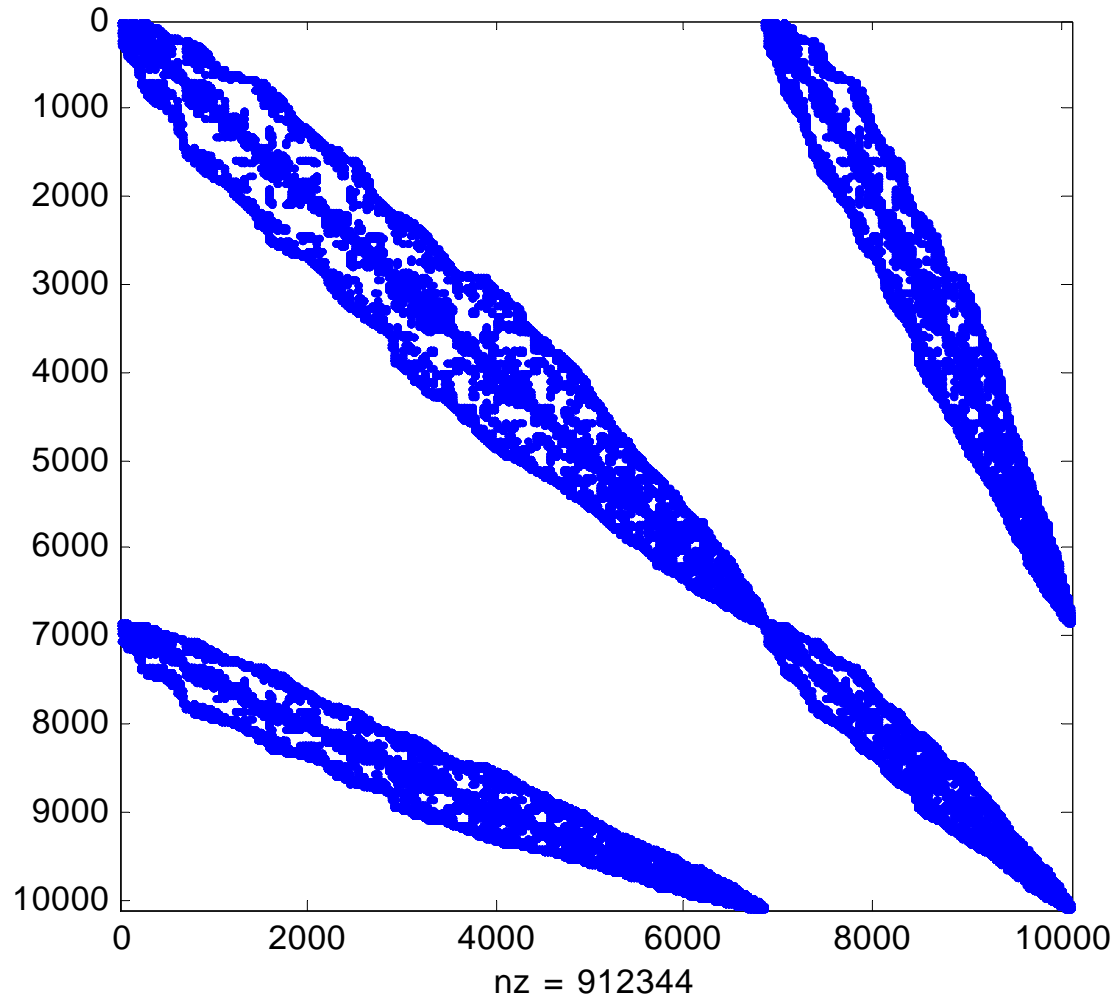
2. *Flow past an inflatable hemisphere*

- *Fluctuating inflow.*
- *Reynold's number = 100*
- *System under consideration is extracted at a given time step.*
- **Cond(A) ~ 1.5 e+3**
- **Cond [A, B; C^T, G] ~ 0.8 e+5**

Flow past a hemisphere



Linear System



Flow past a hemisphere...

- $Re = 100$.
- Choose $E = A$ and remove outer layer.
- *Action of \hat{A}^{-1} is realized via solving systems of the form $A x = b$ using Bicgstab with a diagonal approximate inverse of A as a preconditioner.*

Preconditioned Scheme

Solve
$$\begin{pmatrix} A & B \\ C^T & G \end{pmatrix} \begin{pmatrix} u \\ p \end{pmatrix} = \begin{pmatrix} f \\ g \end{pmatrix}$$

via a Krylov subspace method

Preconditioner

$$M = \begin{pmatrix} E & B \\ C^T & G \end{pmatrix}$$

Solve $Mz = r$ via Richardson scheme

$$z_{k+1} = z_k + \hat{M}^{-1}(r - Mz_k)$$

where,

$$\hat{M} = \begin{pmatrix} \hat{A} & B \\ C^T & G \end{pmatrix}$$

Flow past a hemisphere...

- *Action of \hat{A}^{-1} is realized via solving systems of the form $A x = b$ using Bicgstab with a diagonal approximate inverse of A as a preconditioner.*
 - # of bicgstab iterations < 20 for relative residuals of $O(10^{-9})$.
- Solve $S y = h$, where $S = (G - C^T \hat{A}^{-1} B)$, via bicgstab with a diag. approximate inverse of S to achieve a relative residual $< 10^{-2}$.

Solve
$$\begin{bmatrix} A & B \\ C^T & G \end{bmatrix} \begin{bmatrix} \mathbf{u} \\ \mathbf{p} \end{bmatrix} = \begin{bmatrix} \mathbf{f} \\ \mathbf{g} \end{bmatrix}$$

via Richardson scheme with relative residual $< 10^{-6}$

•
$$\hat{M} = \begin{pmatrix} \hat{A} & B \\ C^T & G \end{pmatrix}$$

$$\begin{bmatrix} \mathbf{u}_{k+1} \\ \mathbf{p}_{k+1} \end{bmatrix} = \begin{bmatrix} \mathbf{u}_k \\ \mathbf{p}_k \end{bmatrix} + \hat{M}^{-1} \left[\begin{pmatrix} \mathbf{f} \\ \mathbf{g} \end{pmatrix} - \begin{pmatrix} A & B \\ C^T & G \end{pmatrix} \begin{pmatrix} \mathbf{u}_k \\ \mathbf{p}_k \end{pmatrix} \right]$$

Note:

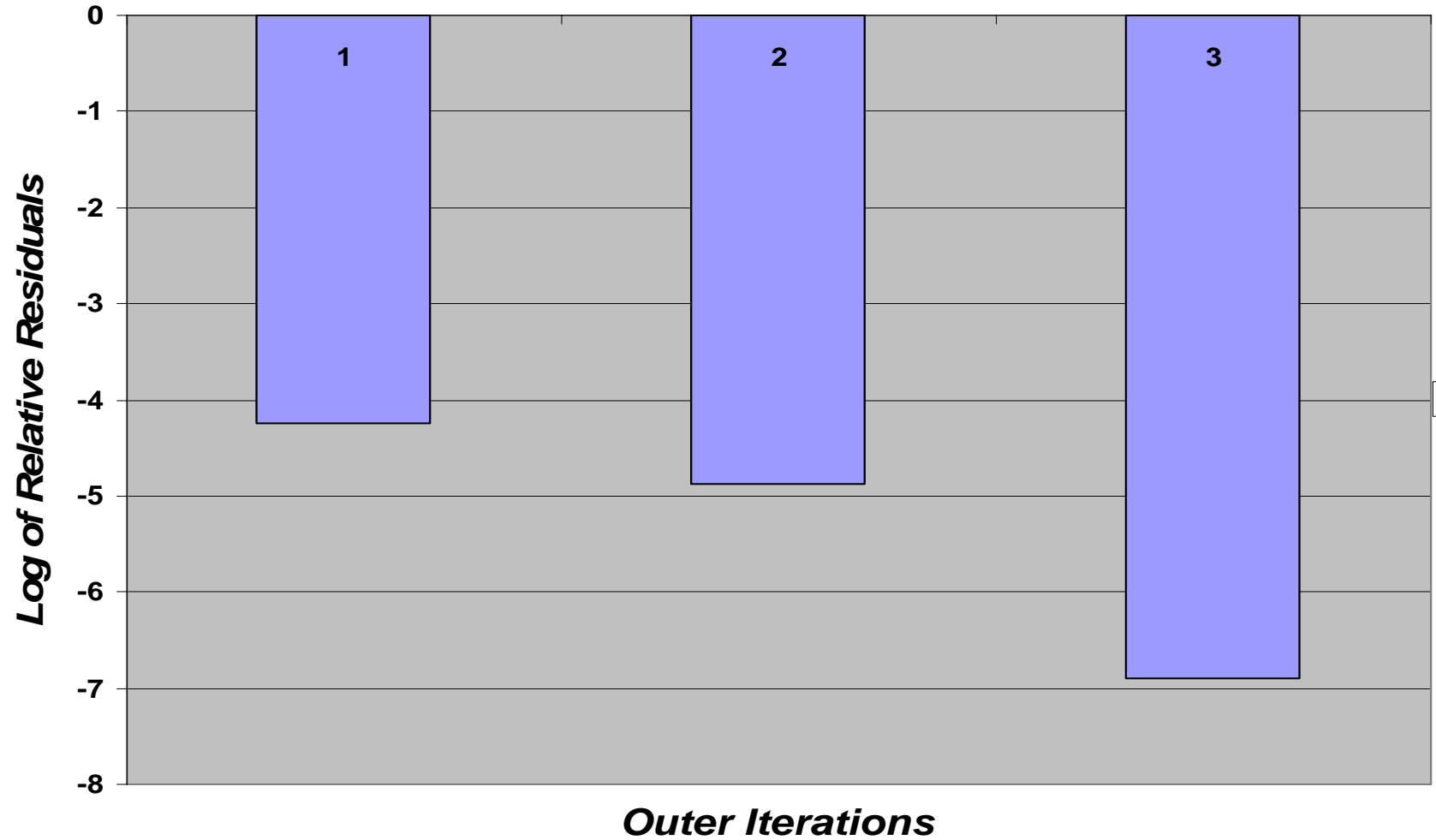
**Solve $(G - C^T \hat{A}^{-1} B)y = h$ via bicgstab with relative residual $< 10^{-2}$.
Preconditioner: diagonal approximate inverse**

Convergence

Need to have $\rho(I - \hat{M}^{-1}M) < 1$

- $\lambda(I - \hat{M}^{-1}M) := 0, \mu$
- $\mu = \lambda[(I + K)(I - \hat{A}^{-1}E)]$
- $K = \hat{A}^{-1}B(G - C^T \hat{A}^{-1}B)^{-1}C^T$
 $\lambda(K) = \lambda[(C^T \hat{A}^{-1}B)(G - C^T \hat{A}^{-1}B)^{-1}]$
- $\rho(I - \hat{A}^{-1}E) \leq \alpha < 1$

Convergence of Richardson Scheme

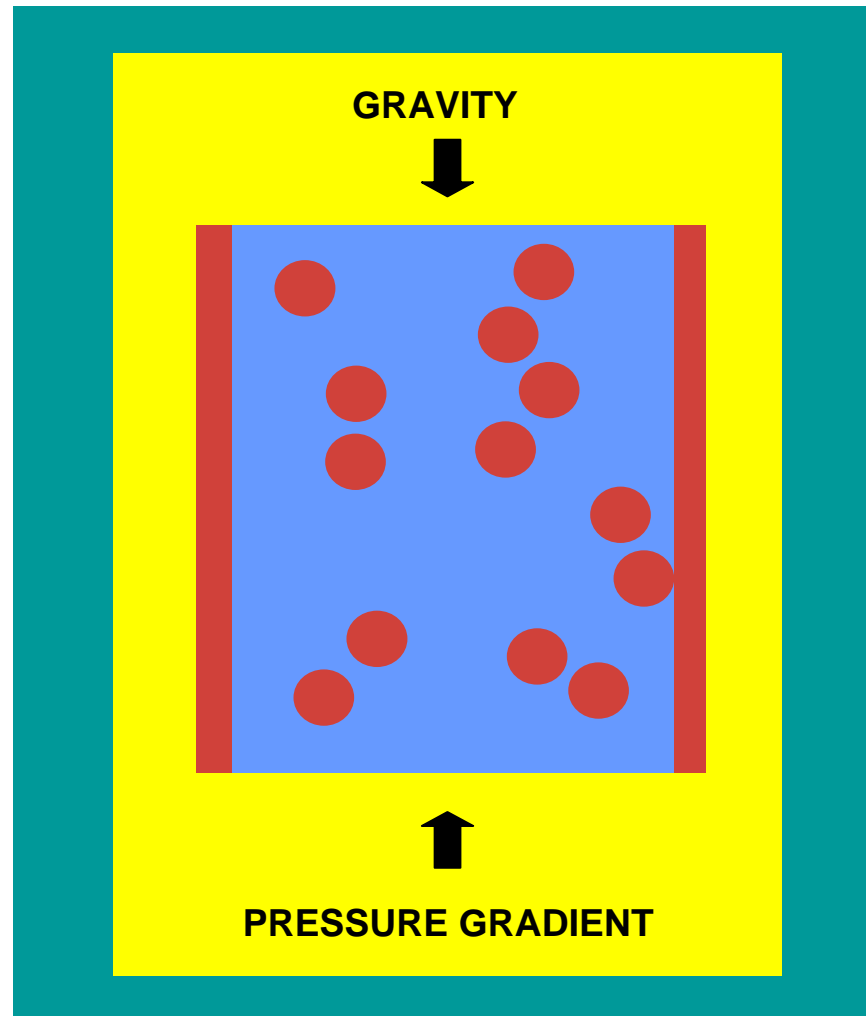


37 inner Bicgstab(10^2) iterations for solving $S y = h$ / outer iteration

Case 2

Particulate Flow

Motivating Application: Particulate Flows



Modeling Particulate Flows

- Navier-Stokes equations coupled with Newton's equations for particles

$$\rho \frac{\partial u}{\partial t} + \rho(u \cdot \nabla)u = \rho g - \nabla p + \nabla \cdot \tau$$

$$\nabla \cdot u = 0$$

$$M \frac{dU}{dt} = F$$

$$\frac{dX}{dt} = U$$

$$\tau = \mu(\nabla u + \nabla u^T)$$

- No-slip on particle surface.
- Unstructured mesh, generated after every few steps.
- Mixed finite elements approximation: $P2/P1$ pair of elements.

Case 2...

Particulate Flow:

- viscosity = 1/100
- G = zero matrix
- $E = A_s = (A + A^T)/2$
- $\hat{A}^{-1} :=$ **diagonal approximate inverse of A_s ,**
i.e.

$$\min \left\| I - \hat{A}^{-1} A_s \right\|_F$$

- $\rho(I - \hat{A}^{-1} A_s) = \alpha < 1$

Preconditioned Scheme: Case 2

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

via Gmres $\left[\frac{\|\mathbf{r}_k\|}{\|\mathbf{r}_0\|} \leq 10^{-6} \right]$

Preconditioner

$$M = \begin{pmatrix} A_s & B \\ B^T & 0 \end{pmatrix}$$

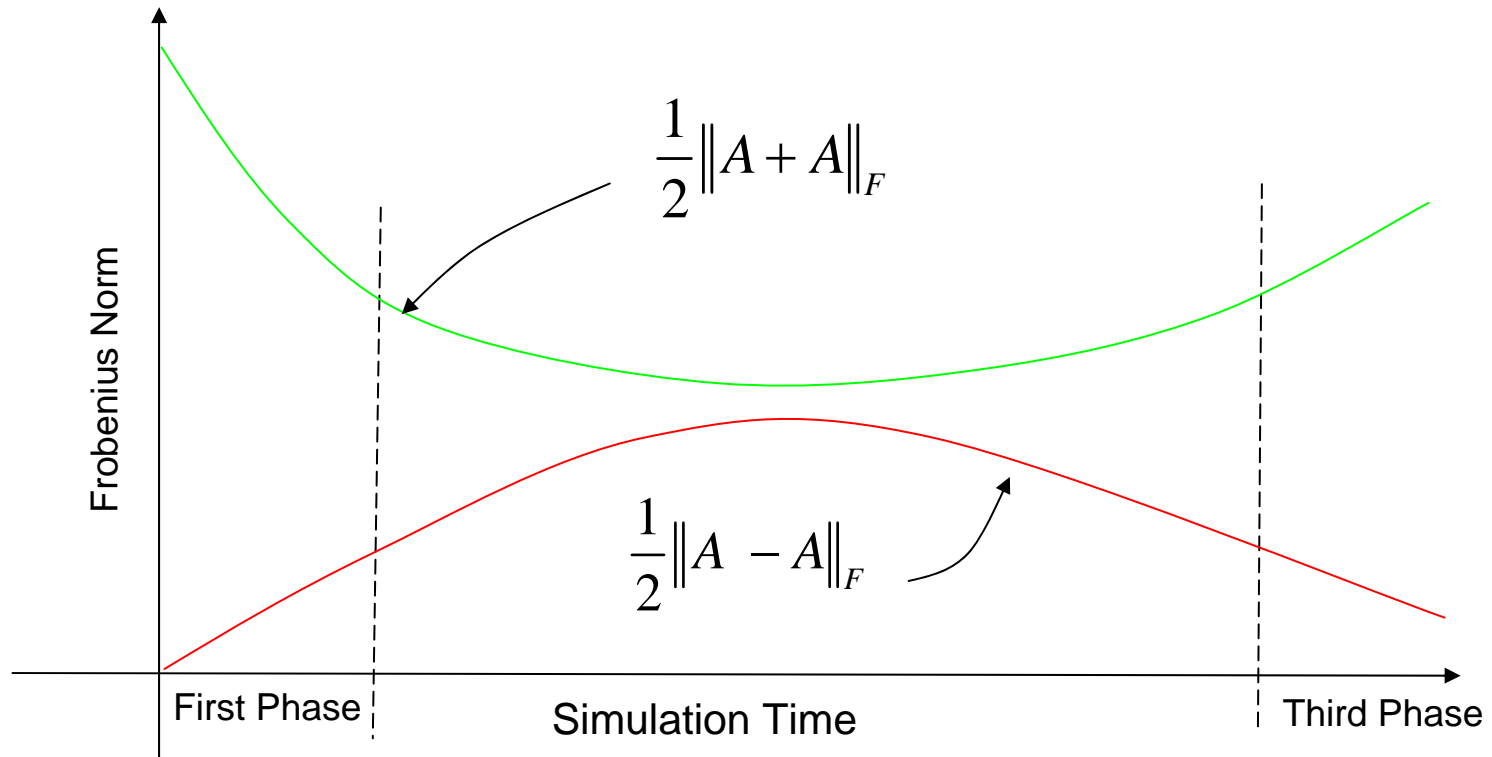
Solve $M\mathbf{z} = \mathbf{r}$ via Richardson scheme $\left[\frac{\|\mathbf{r}_k\|}{\|\mathbf{r}_0\|} \leq 10^{-6} \right]$

$$\mathbf{z}_{k+1} = \mathbf{z}_k + \hat{M}^{-1}(\mathbf{r} - M\mathbf{z}_k)$$

where,

$$\hat{M} = \begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix}$$

Frobenius Norm of the Skew-Symmetric Part of A



- To quantify the skew-symmetry, we introduce the “degree of skewness”

measured by
$$\frac{\|A - A^T\|_F}{\|A + A^T\|_F} = \|A_{ss}\|_F / \|A_s\|_F .$$

Solving $\hat{M}u = c$; Case 2

$$\text{Solve } \begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \mathbf{w} \end{pmatrix} = \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix}$$

Since

$$\begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix} = \begin{pmatrix} \hat{A} & 0 \\ B^T & I \end{pmatrix} \begin{pmatrix} \hat{A}^{-1} & 0 \\ 0 & -\hat{G} \end{pmatrix} \begin{pmatrix} \hat{A} & B \\ 0 & I \end{pmatrix}, \text{ with } \hat{G} = B^T \hat{A}^{-1} B,$$

the solution scheme is given by,

- $\mathbf{v} = \hat{A}^{-1} \mathbf{a}$
- $\hat{G} \mathbf{w} = B^T \mathbf{v} - \mathbf{b}$ via CG with $\|\mathbf{r}_k\|_2 / \|\mathbf{r}_0\| \leq 10^{-2}$.
- $\mathbf{v} = \hat{A}^{-1} (\mathbf{a} - B \mathbf{w})$

Convergence....

- $\lambda(I - \hat{M}^{-1}M) := 0, \mu$
- $\mu = \lambda[(I - K)(I - \hat{A}^{-1}A_s)]$
- $\lambda(K) = \lambda[\hat{A}^{-1/2}B(B^T \hat{A}^{-1}B)^{-1} B^T \hat{A}^{-1/2}]$; [a projector]
- Thus $\rho(I - \hat{M}^{-1}M) < 1$

$$\hat{A}^{-1} := SPAI(A_s) \text{ [diagonal]}$$

$$\alpha = \rho(I - \hat{A}^{-1} A_s) < 1$$

t	# particles	system size	Gmres(k)		
			k	outer	inner
$20 \Delta t$	20	8,777	20	1	13
$100 \Delta t$	240	95,749	50	3	14
$200 \Delta t$	240	111,326	50	4	15

Robustness of solver $(\|r_k\|_2 / \|r_0\|_2 < 10^{-6})$

- **Gmres with approx. LU factorizations of \tilde{A} (as preconditioners) has failed.**

Case 3

Driven Cavity: steady-state case

Case 3...

- Square domain: $-1 \leq x, y \leq 1$
B.C^S.: $u_x = u_y = 0$ on $x, y = -1 ; x = 1$
 $u_x = 1, u_y = 0$ on $y = 1$
- Picard's iterations; **Q2/Q1 elements**: Linearized equations (**Oseen problem**)
- $G \equiv 0$
- $E = A_s = (A + A^T)/2$
- viscosity: 0.1, 0.02, 0.01, 0.002

Preconditioned Scheme

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

via a Krylov subspace method

Preconditioner

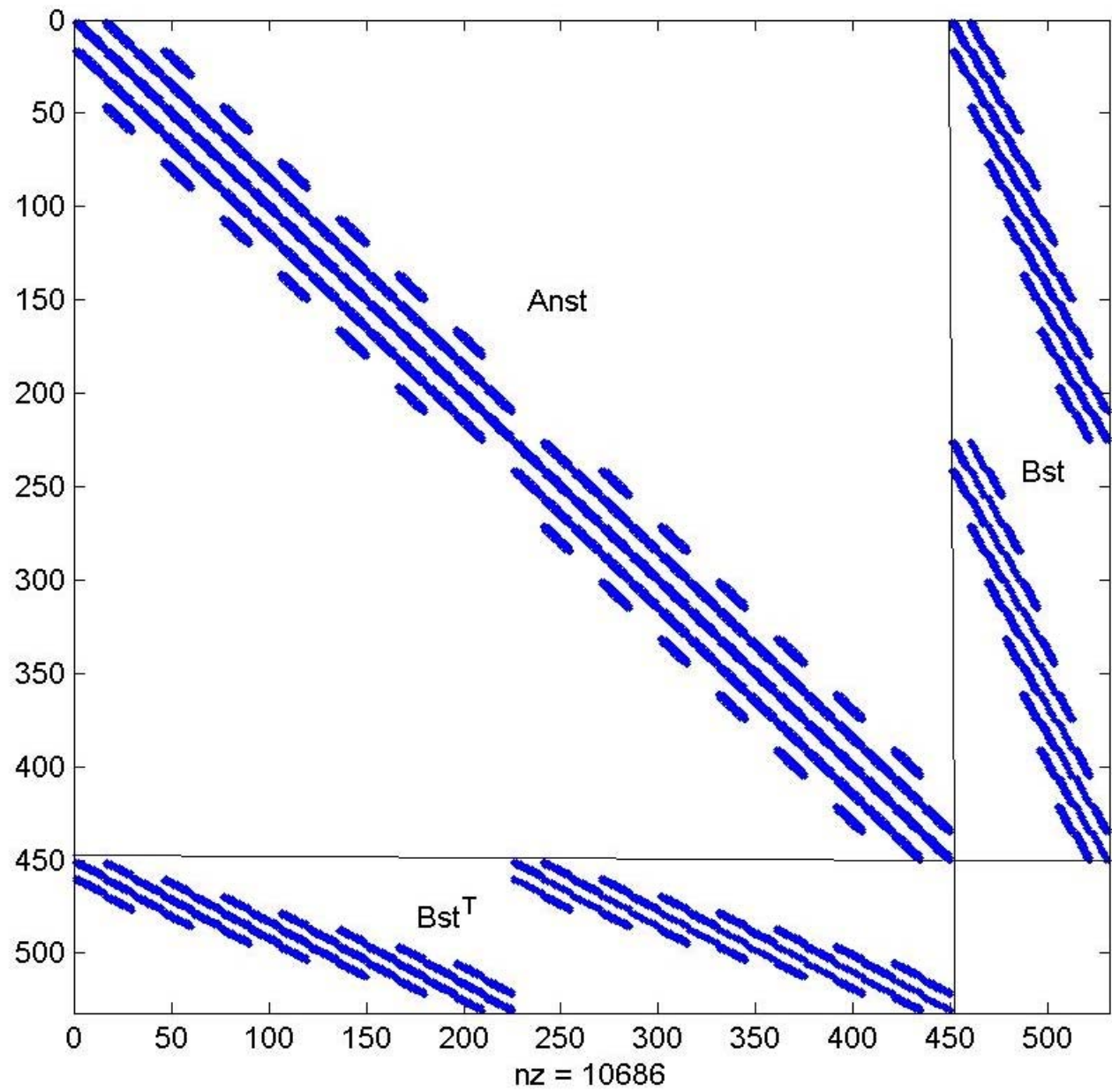
$$M = \begin{pmatrix} A_s & B \\ B^T & 0 \end{pmatrix} ; A_s = (A + A^T) / 2$$

Solve $M\mathbf{z} = \mathbf{r}$ via Richardson scheme

$$\mathbf{z}_{k+1} = \mathbf{z}_k + \hat{M}^{-1}(\mathbf{r} - M\mathbf{z}_k)$$

where,

$$\hat{M} = \begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix}$$



Case 3...

- $A_s = P + N$; [Axelsson & Kolotolina]
 $P(i,i) = 0$; $P(i,j) \geq 0$
 $N(i,i) > 0$; $N(i,j) \leq 0$
- $D e = P e$
 $D := \text{diagonal}$; $e^T = (1,1,\dots,1)$
- $\hat{A} = D + N$
 \hat{A} is a Stieltjes matrix
 $\hat{A}(i,i) > 0$; $\hat{A}(i,j) \leq 0$; $\hat{A}^{-1} > 0$

Case 3: Scheme I

Solving $\hat{M}u = c$; Case 3

$$\text{Solve } \begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \mathbf{w} \end{pmatrix} = \begin{pmatrix} \mathbf{a} \\ \mathbf{b} \end{pmatrix}$$

Since

$$\begin{pmatrix} \hat{A} & B \\ B^T & 0 \end{pmatrix} = \begin{pmatrix} \hat{A} & 0 \\ B^T & I \end{pmatrix} \begin{pmatrix} \hat{A}^{-1} & 0 \\ 0 & -\hat{G} \end{pmatrix} \begin{pmatrix} \hat{A} & B \\ 0 & I \end{pmatrix}, \text{ with } \hat{G} = B^T \hat{A}^{-1} B$$

we solve the following systems using pcg with $\|\mathbf{r}_k\|_2 / \|\mathbf{r}_0\| \leq 10^{-1}$.

- $\hat{A}\mathbf{v} = \mathbf{a}$
- $\hat{G}\mathbf{w} = B^T \mathbf{v} - \mathbf{b}$
- $\hat{A}\mathbf{v} = \mathbf{a} - B\mathbf{w}$

Each requires no more than 2 pcg iterations.

Convergence of Richardson Scheme

$$\rho(I - \hat{A}^{-1}A_s) \approx 0.37$$

- for all mesh sizes: $2^k \times 2^k$

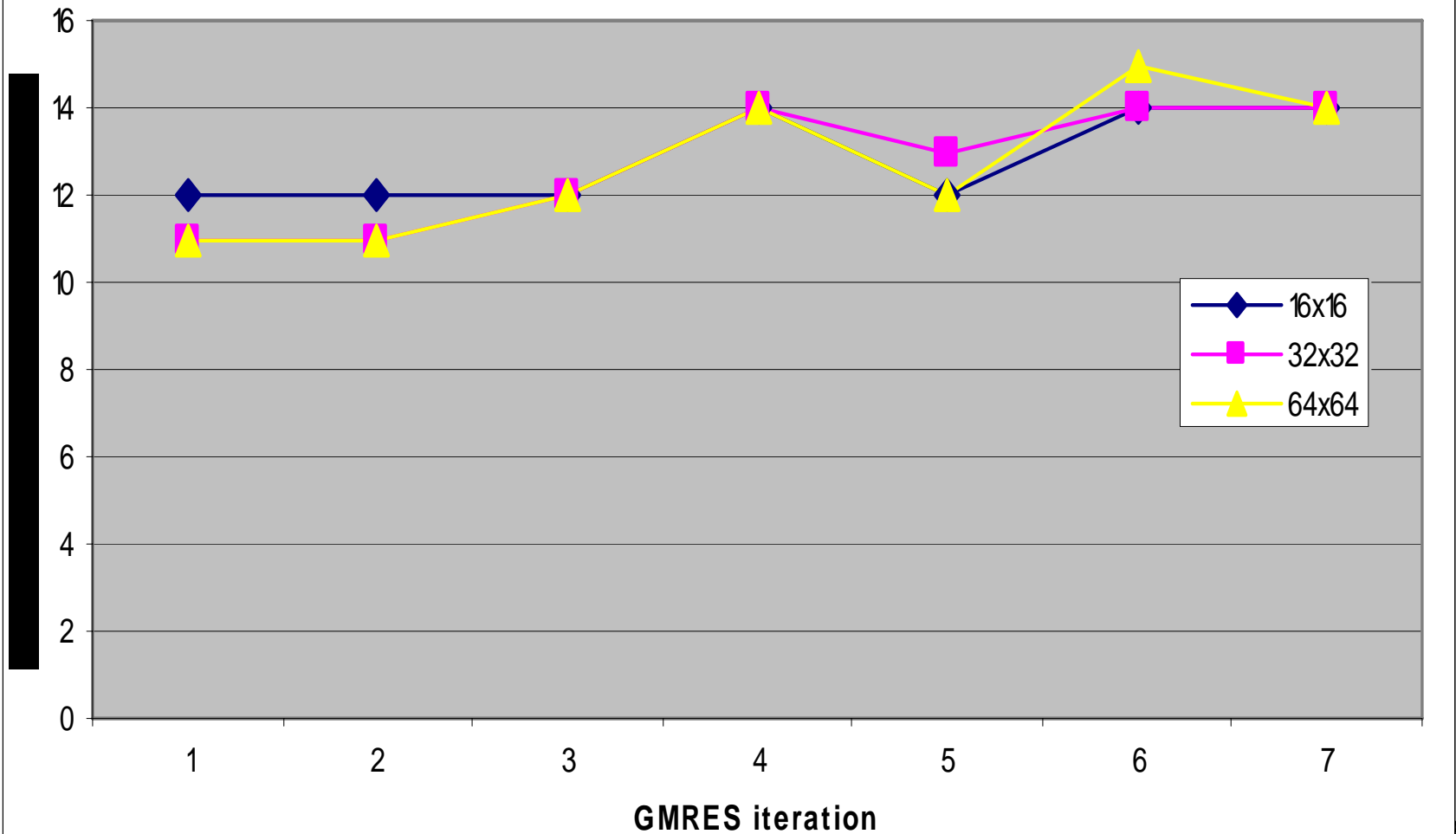
$$k = 4, 5, 6, 7, 8, \dots$$

- for viscosity:

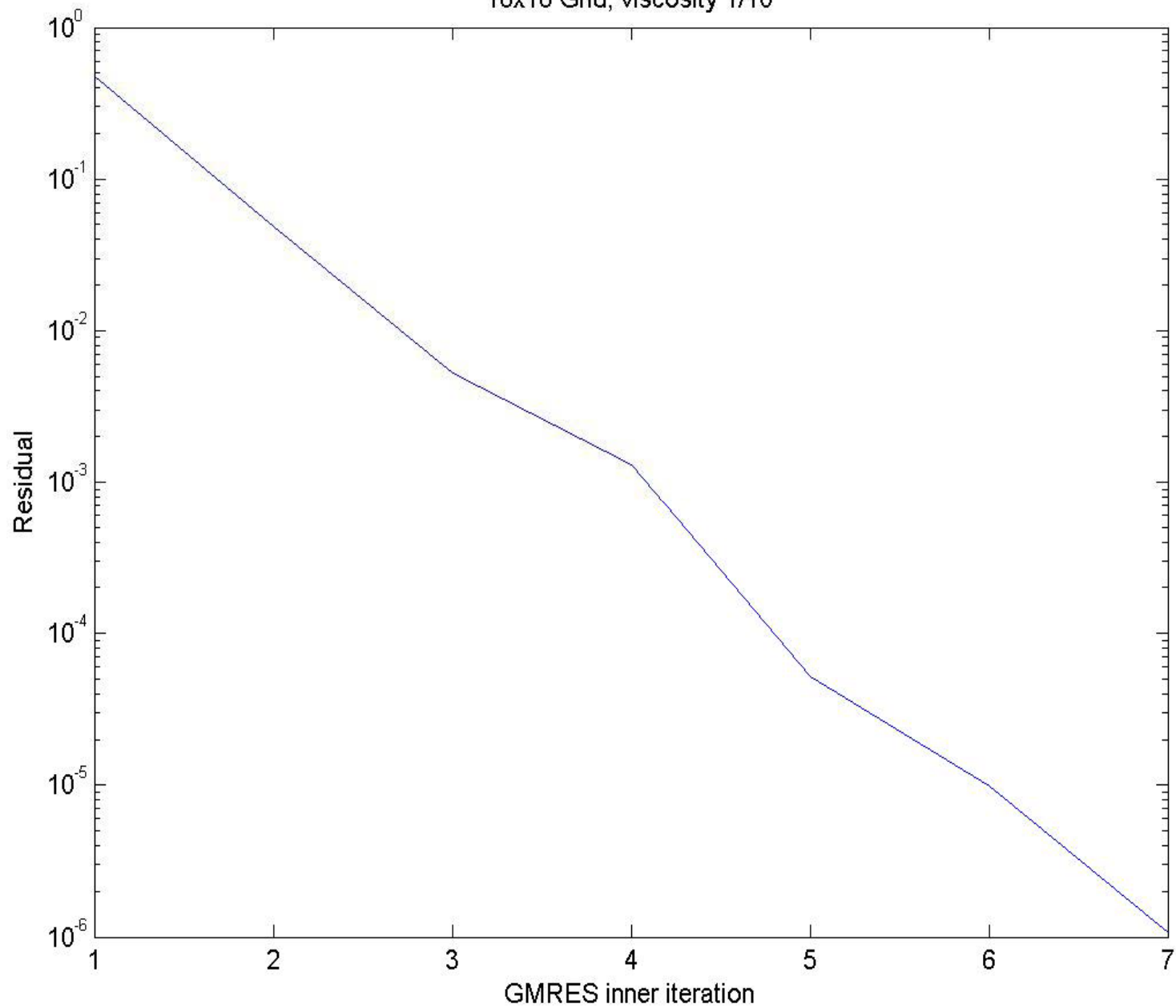
$$0.1; 0.02; 0.01; 0.002$$

Number of Richardson iterations in each GMRES iteration

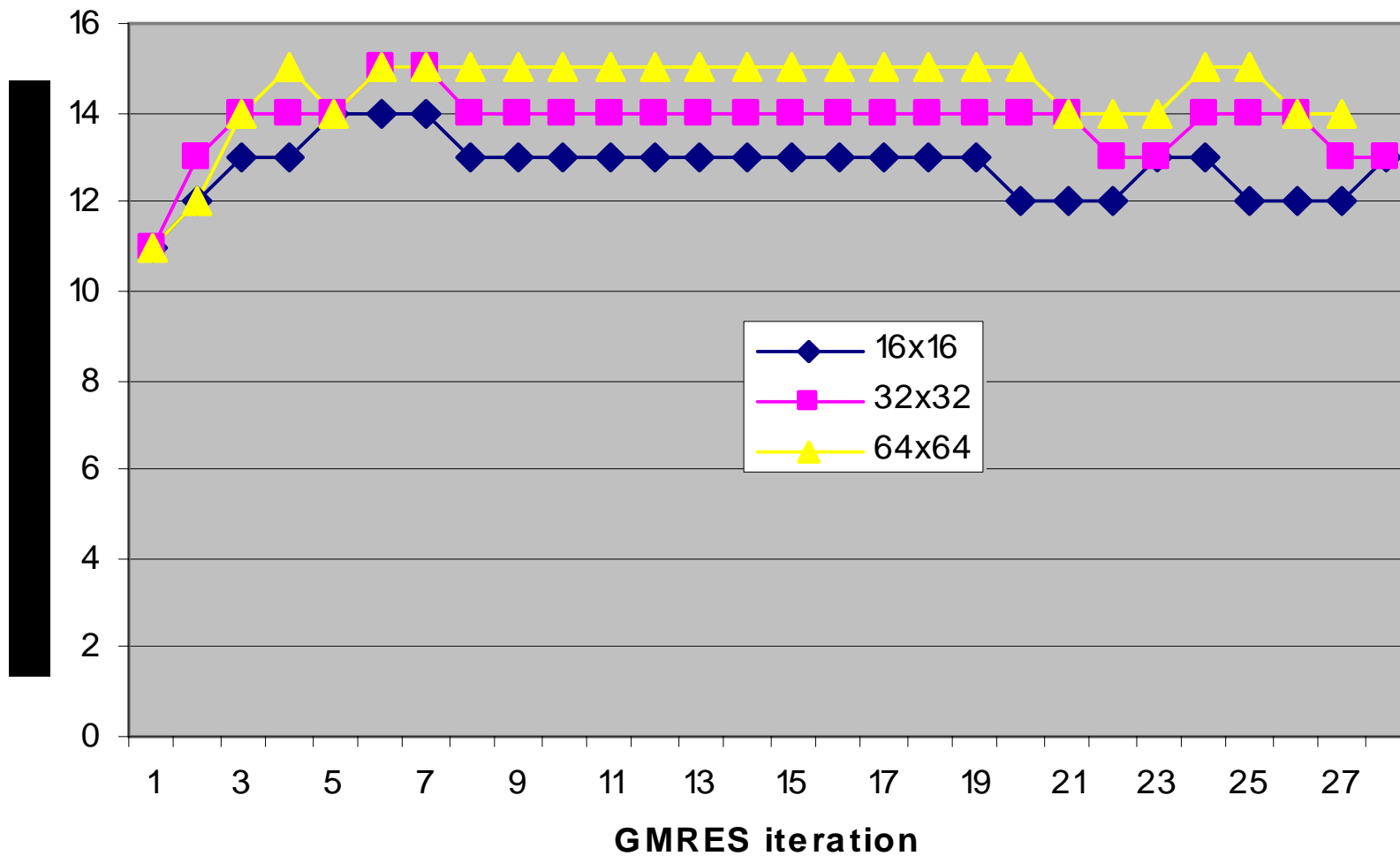
Viscosity 1/10



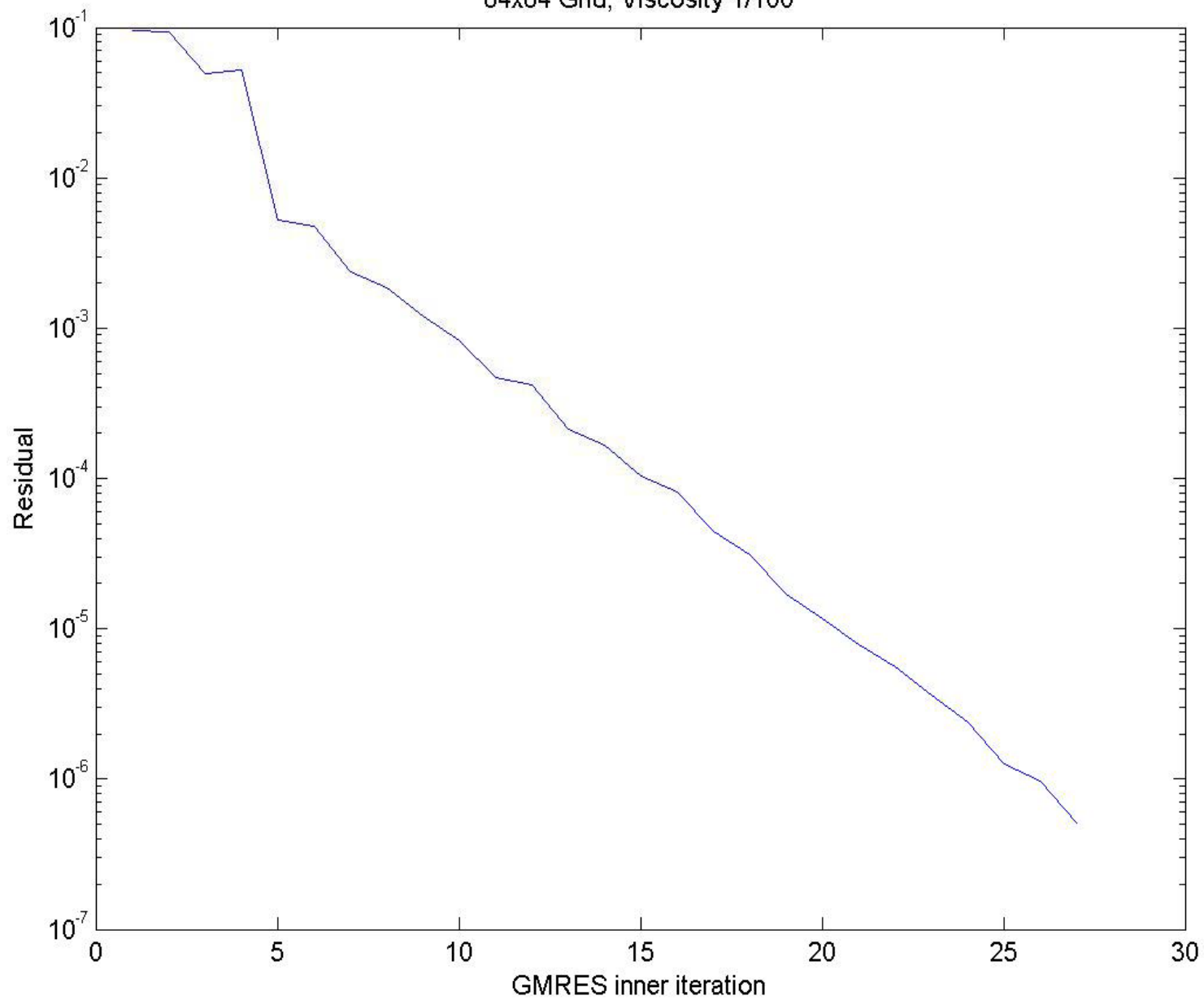
16x16 Grid, viscosity 1/10



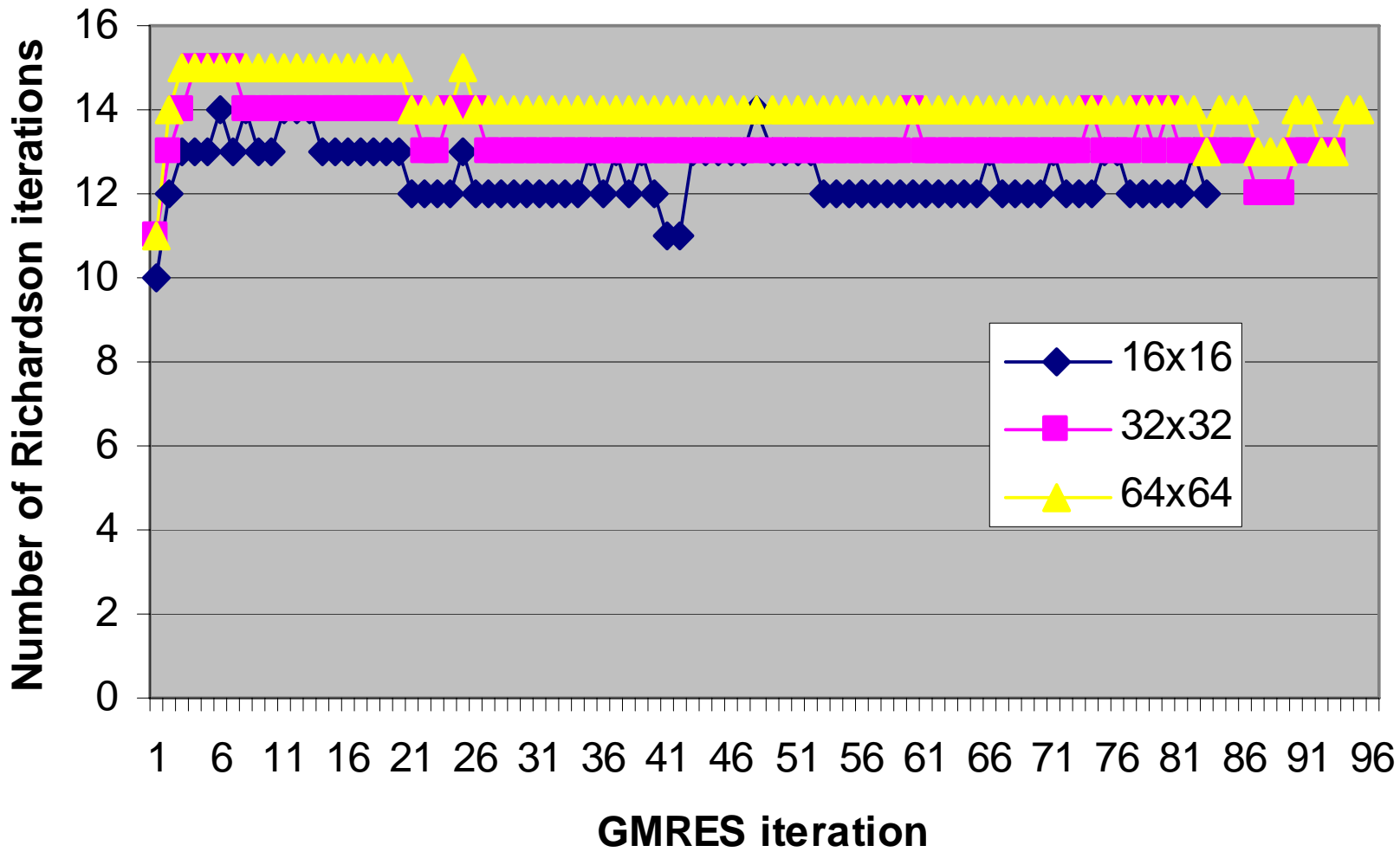
Number of Richardson iterations in each GMRES iteration Viscosity 1/100



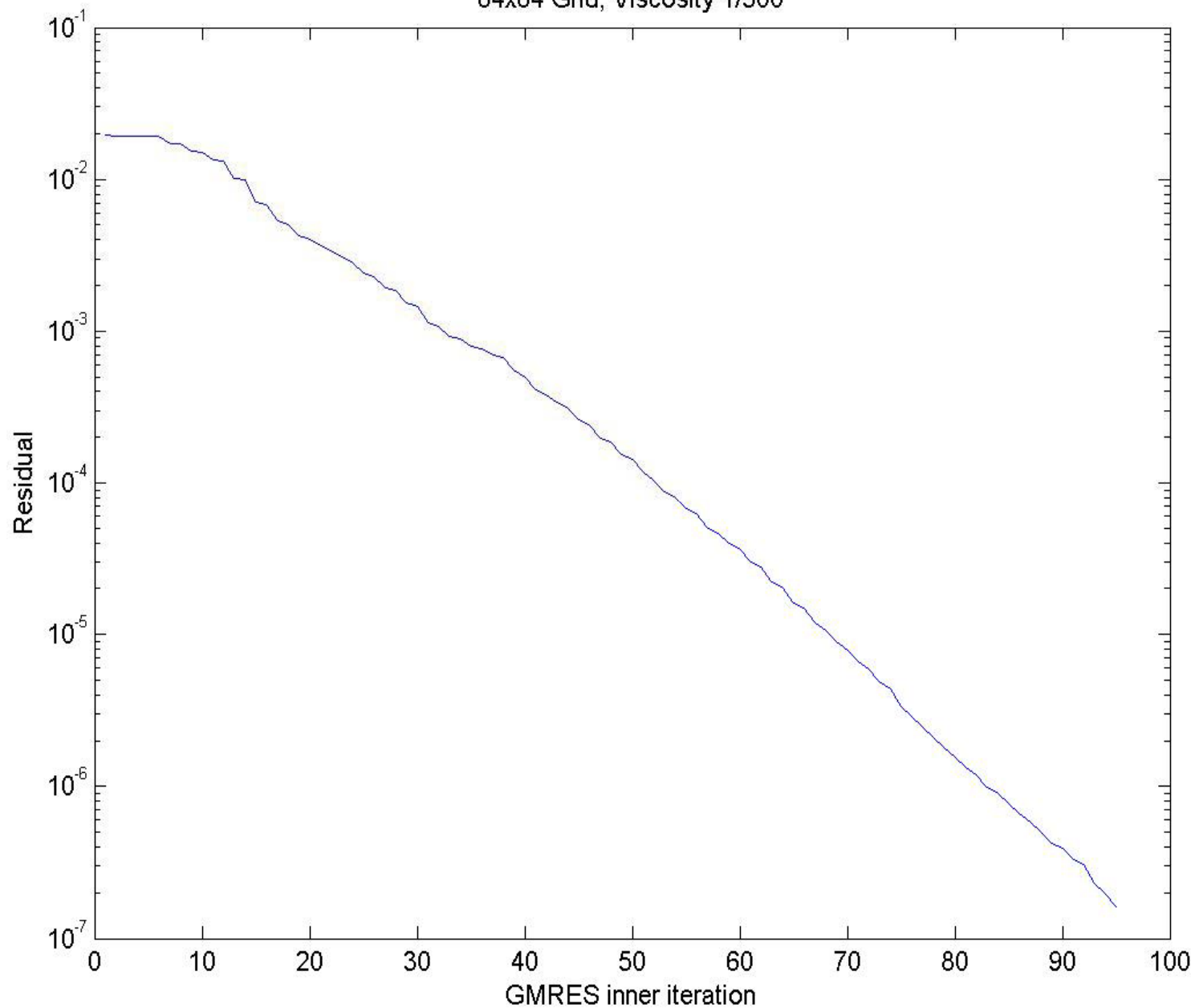
64x64 Grid, Viscosity 1/100



Number of Richardson iterations in each GMRES iteration Viscosity 1/500

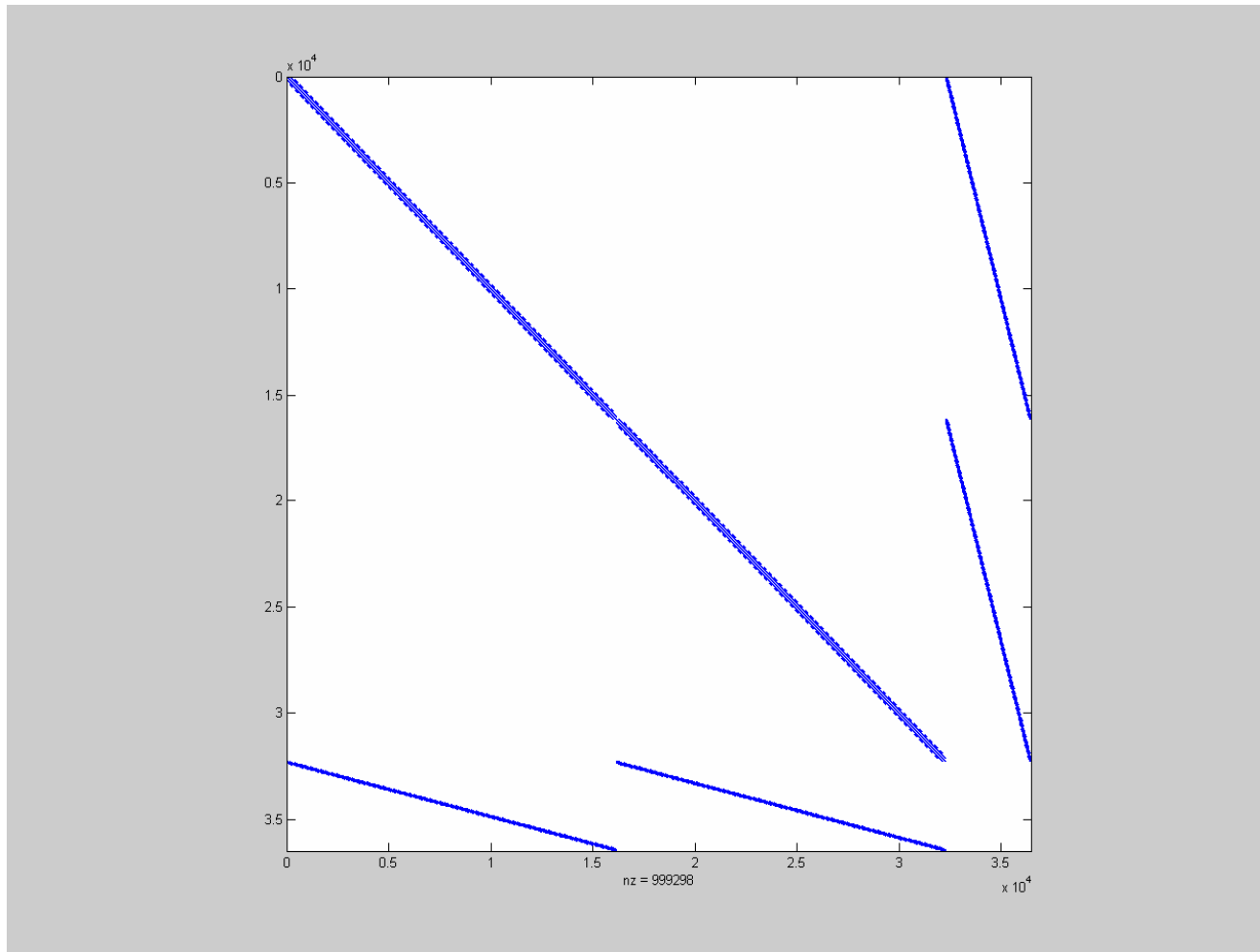


64x64 Grid, Viscosity 1/500

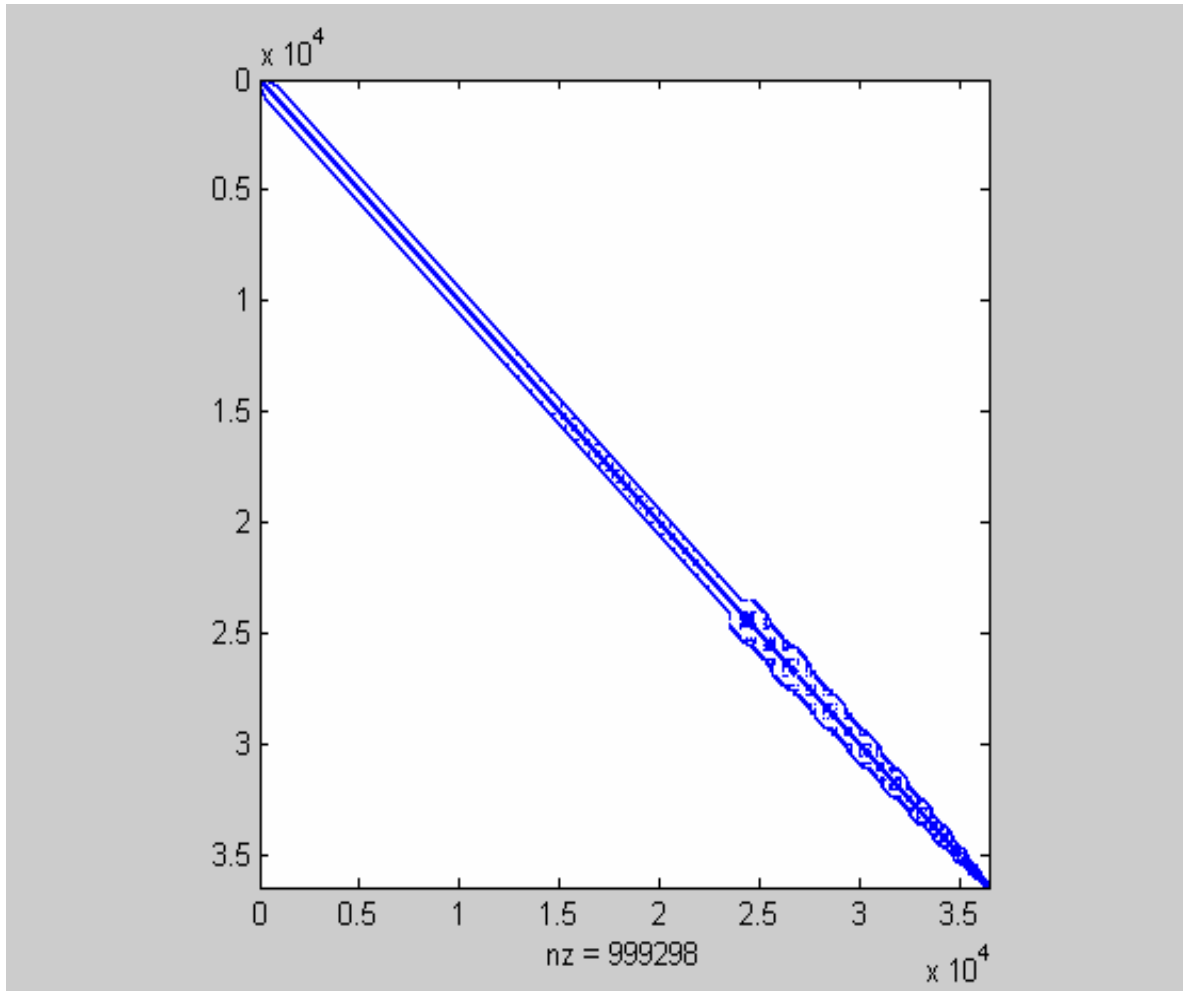


Case 3: Scheme II

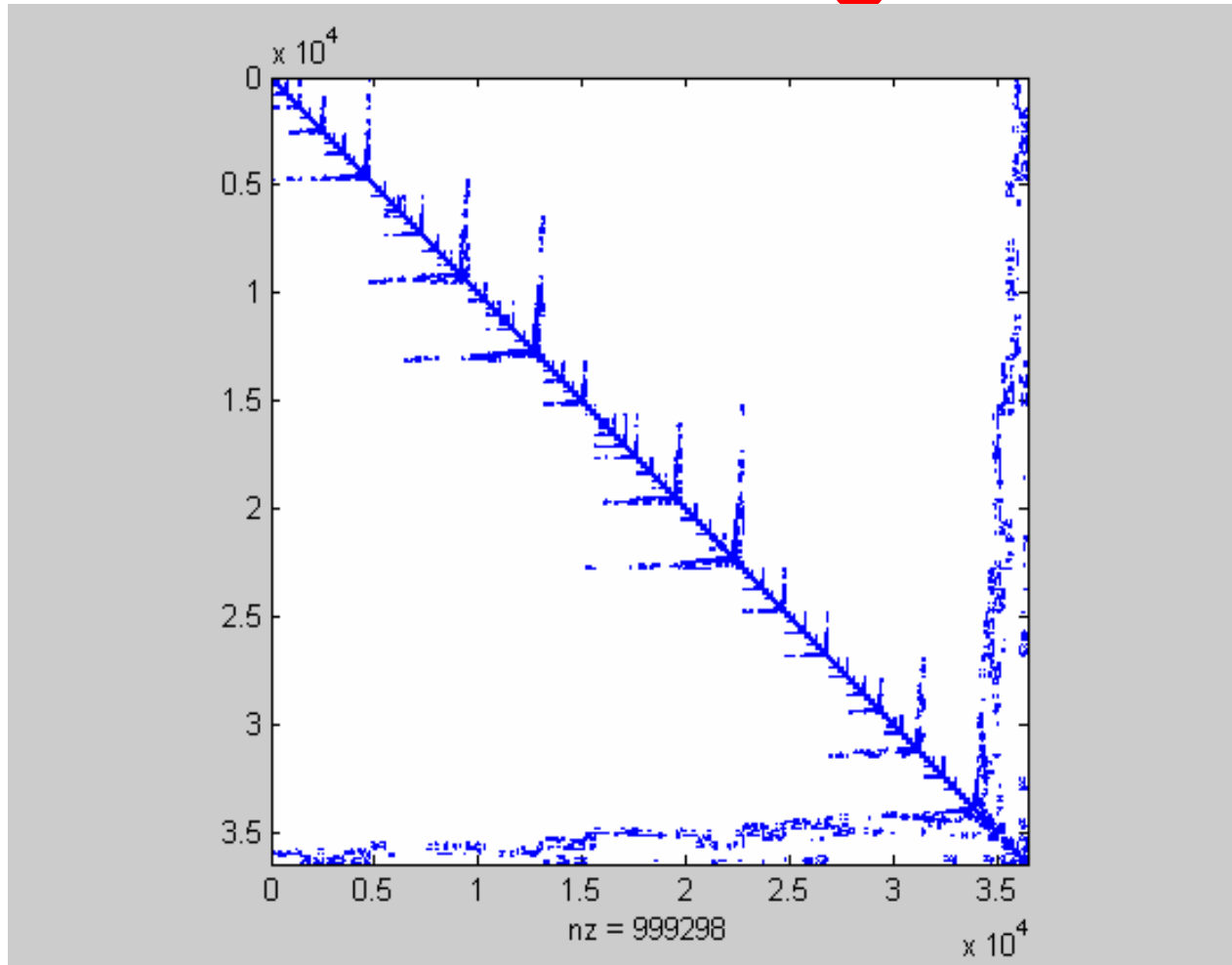
Linear system for 128 x 128 grid



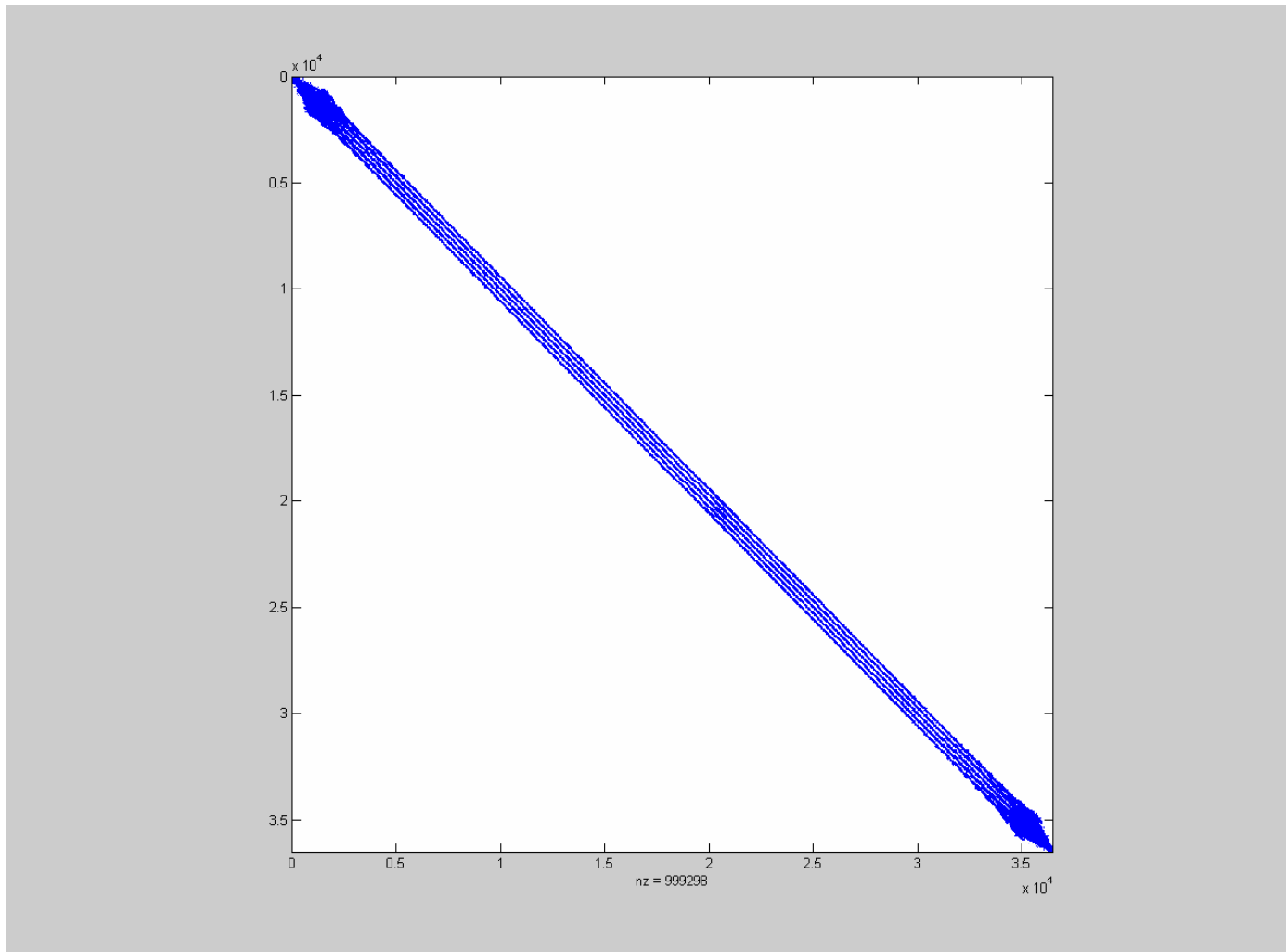
after RCM reordering



after approx. min. degree reordering



after spectral reordering



Spectral Reordering

- Obtain second smallest eigenpairs of the Laplacian.
- Sort the elements of the eigenvector (Fiedler vector).
- Permute the rows and columns of the matrix using the sorted order.
- W : [sum of magnitudes of the elements within the selected band] / [sum of the magnitudes of all the matrix elements]
 - $W = 99.9\%$, for bandwidth = 1401, and viscosity=1/50.

MA48 as a preconditioner for GMRES on a uniprocessor

<i>Droptol</i>	<i># of iter.</i>	<i>T(factor.)</i>	<i>T(GMRES iters.)</i>	<i>nnz(L+U)</i>	<i>final residual</i>	<i>v</i>
0	2	142.7	0.3	7,195,157	2.07E-13	1/50
1.00E-04	24	115.5	1.4	2,177,181	2.83E-09	1/50
1.00E-03	248	109.6	17.3	1,611,030	5.11E-08	1/50

**** for drop tolerance = .0001, total time ~ 117 sec.**

**** No convergence for a drop tolerance of 10^{-2}**

Spike – Pardiso for solving systems involving the banded preconditioner

- ***On 4 nodes (8 CPU's) of a Xeon-Intel cluster:***
 - ***bandwidth of extracted preconditioner = 1401***
 - ***total time = 16 sec.***
 - ***# of Gmres iters. = 131***
 - ***2-norm of residual = 10^{-7}***
- ***Speed improvement over sequential procedure with drop tolerance = .0001:***
 - ***117/16 ~ 7.3***

Conclusion & Future Work

- 1. Proposed nested scheme is:**
 - versatile with appropriate choice of preconditioners at the three nesting levels,
 - suitable for implementation on parallel computing platforms.
- 2. Reordering the resulting “segregated” system yields often a prominent band that can be used as an effective parallel preconditioner.**
- 3. Plan to build an environment for construction of these solvers with fewer input parameters from the application user.**

Thank you!