Scalable implicit algorithms for stiff hyperbolic PDE systems

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Outline

- ► Motivation: the tyranny of scales
- ► Block-factorization preconditioning of hyperbolic PDEs
- ► Compressible resistive MHD
- ► Compressible extended MHD
- ► Incompressible Navier-Stokes and MHD (infinite sound-speed limit)



"The tyranny of scales" (2006 NSF SBES report)

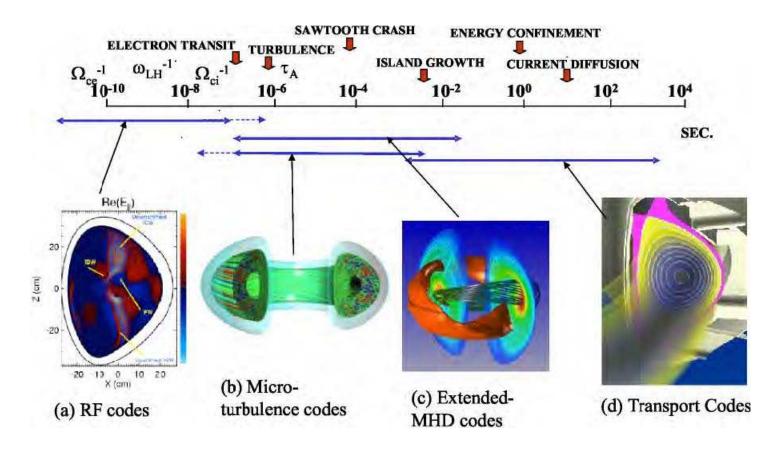


Figure 1: Time scales in fusion plasmas (FSP report)



Algorithmic challenges in temporal scale-bridging

- ► PDE systems of interest typically have mixed character, with hyperbolic and parabolic components. □ Hyperbolic stiffness (linear and dispersive waves): $\kappa(J) \sim \Delta t \, \omega_{fast} \sim \frac{\Delta t}{\Delta t_{CFL}} \gg 1$ □ Parabolic stiffness (diffusion): $\kappa(J) \sim \frac{\Delta t D}{\Delta r^2} \gg 1$
- ➤ In some applications, fast hyperbolic modes carry a lot of energy (e.g., shocks, fast advection of solution structures), and the modeler must follow them.
- In others, however, fast time scales are parasitic, and carry very little energy.
 - □ These are the ones that are usually targeted for scale-bridging.
- Bridging the time-scale disparity requires a combination of approaches:
 - □ Analytical elimination (e.g., reduced models).
 - □ Well-posed numerical discretization (e.g., asymptotic preserving methods)
 - **Some level of implicitness in the temporal formulation (for stability; accuracy requires care)**.
- ► Key algorithmic requirement: SCALABILITY

$$CPU \sim \mathcal{O}\left(\frac{N}{n_p}\right)$$



Algorithmic scalability vs. parallel scalability

"The tyranny of scales will not be simply defeated by building bigger and faster computers" (NSF SBES 2006 report, p. 30)

► Optimal algorithm: $CPU \sim N/n_p$

$$CPU \sim \frac{N^{1+\alpha}}{n_p^{1-\beta}} \quad ; \ N = \left(\frac{L}{\delta}\right)^d \begin{cases} \alpha \ge 0, \text{ algorithmic scalability} \\ \beta \ge 0, \text{ parallel scalability} \end{cases}$$

- > Much emphasis has been placed on parallel scalability (β).
- ► However, parallel (weak) scalability is limited by the lack of algorithmic scalability: □ $N \propto n_p \Rightarrow CPU \sim n_p^{\alpha+\beta} \Rightarrow$ requires $\alpha = \beta = 0!$

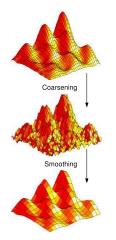
ExplicitImplicit (direct)Implicit (Krylov iterative)Implicit (multilevel) $\alpha = 1/d$ $\alpha = 2 - 2/d$ $\alpha > 1$ (varies) $\alpha \approx 0$

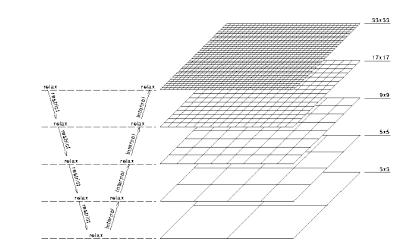


How do multilevel (multigrid) methods work?

MG employs a divide-and-conquer approach to attack error components in the solution.
 Oscillatory components of the error are "EASY" to deal with (if a SMOOTHER exists)
 Smooth components are DIFFICULT.

Idea: coarsen grid to make "smooth" components appear oscillatory, and proceed recursively





- SMOOTHER is make or break of MG!
- Smoothers are hard to find for hyperbolic systems, but fairly easy for parabolic ones:

Can one make hyperbolic PDEs MG-friendly?



Implicit discretization of hyperbolic PDEs: a case study

$$\partial_t u = rac{1}{\epsilon} \partial_x v$$
 , $\partial_t v = rac{1}{\epsilon} \partial_x u$; $\omega = \pm rac{k}{\epsilon}$

 \succ ϵ is a measure of hyperbolic stiffness. Discretize implicitly in time:

$$u^{n+1} = u^n + rac{1}{\epsilon} \partial_x v^{n+1}$$
 , $v^{n+1} = v^n + rac{1}{\epsilon} \partial_x u^{n+1}.$

Very ill conditioned as $\epsilon
ightarrow 0!$ However, if one combines equations:

$$\left[I - \left(\frac{\Delta t}{\epsilon}\right)^2 \partial_x^2\right] u^{n+1} = u^n + \frac{\Delta t}{\epsilon} \partial_x v^n$$

- **Equation is now well-posed when** $\epsilon \to 0$ (i.e., it is **asymptotic-preserving**)!
 - □ Limit system is elliptic/parabolic (MG-friendly!)
 - □ *Temporally unresolved* hyperbolic time scales have been "parabolized."
 - ➡ No further manipulation of PDE than implicit differencing (no terms added to PDE)!
 - This fact can be exploited to devise optimal solution algorithms (block factorization)!



Block-factorization of hyperbolic PDEs

$$u^{n+1} = u^n + rac{\Delta t}{\epsilon} \partial_x v^{n+1}$$
, $v^{n+1} = v^n + rac{\Delta t}{\epsilon} \partial_x u^{n+1}$

► Coupling structure:

$$\begin{bmatrix} \mathrm{I} & -\frac{\Delta t}{\epsilon} \partial_x \\ -\frac{\Delta t}{\epsilon} \partial_x & \mathrm{I} \end{bmatrix} \begin{pmatrix} u^{n+1} \\ v^{n+1} \end{pmatrix} = \begin{pmatrix} u^n \\ v^n \end{pmatrix}$$

 \blacktriangleright 2×2 block can be formally inverted via block factorization:

$$\begin{bmatrix} D_1 & \frac{1}{\epsilon}U\\ \frac{1}{\epsilon}L & D_2 \end{bmatrix} = \begin{bmatrix} I & \frac{1}{\epsilon}UD_2^{-1}\\ 0 & I \end{bmatrix} \begin{bmatrix} D_1 - \frac{1}{\epsilon^2}UD_2^{-1}L & 0\\ 0 & D_2 \end{bmatrix} \begin{bmatrix} I & 0\\ \frac{1}{\epsilon}D_2^{-1}L & I \end{bmatrix}$$

> Only inverse of $D_1 - UD_2^{-1}L$ (Schur complement) is required!

$$D_1 - \frac{1}{\epsilon^2} U D_2^{-1} L = I - \left(\frac{\Delta t}{\epsilon}\right)^2 \partial_x^2$$



Nonlinear hyperbolic PDEs: JFNK and block factorization preconditioning

> Objective: solve nonlinear system $\vec{G}(\vec{x}^{n+1}) = \vec{0}$ efficiently (scalably).

Converge nonlinear couplings using Newton-Raphson method:

► Jacobian-free implementation:

$$\left(\frac{\partial \vec{G}}{\partial \vec{x}}\right)_{k} \vec{y} = J_{k} \vec{y} = \lim_{\epsilon \to 0} \frac{\vec{G}(\vec{x}_{k} + \epsilon \vec{y}) - \vec{G}(\vec{x}_{k})}{\epsilon}$$

Krylov method of choice: GMRES (nonsymmetric systems).

> Right preconditioning: solve equivalent Jacobian system for $\delta y = P_k \delta \vec{x}$:

$$J_k P_k^{-1} \underbrace{\underline{P_k \delta \vec{x}}}_{\delta \vec{y}} = -\vec{G}_k$$

- > Approximations in preconditioner do not affect accuracy of converged solution; only efficiency!
- Block-factorization+MG will be our preconditioning strategy.



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 $\frac{\partial \vec{G}}{\partial \vec{x}} \bigg|_{\vec{k}} \delta \vec{x}_k = -\vec{G}(\vec{x}_k)$

Implicit resistive MHD solver

L. Chacon, Phys. Plasmas (2008)



Resistive MHD model equations

$$\begin{split} \frac{\partial \rho}{\partial t} &+ \nabla \cdot (\rho \vec{v}) = 0, \\ \frac{\partial \vec{B}}{\partial t} &+ \nabla \times \vec{E} = 0, \\ \frac{\partial (\rho \vec{v})}{\partial t} + \nabla \cdot \left[\rho \vec{v} \vec{v} - \vec{B} \vec{B} &- \rho \nu \nabla \vec{v} + \overleftarrow{I} (p + \frac{B^2}{2}) \right] = 0, \\ \frac{\partial T}{\partial t} + \vec{v} \cdot \nabla T &+ (\gamma - 1) T \nabla \cdot \vec{v} = 0, \end{split}$$

> Plasma is assumed polytropic $p \propto n^{\gamma}$.

► Resistive Ohm's law:

$$\vec{E} = -\vec{v} \times \vec{B} + \eta \nabla \times \vec{B}$$



Resistive MHD Jacobian block structure

► The linearized resistive MHD model has the following couplings:

$$\begin{split} \delta \rho &= L_{\rho}(\delta \rho, \delta \vec{v}) \\ \delta T &= L_{T}(\delta T, \delta \vec{v}) \\ \delta \vec{B} &= L_{B}(\delta \vec{B}, \delta \vec{v}) \\ \delta \vec{v} &= L_{v}(\delta \vec{v}, \delta \vec{B}, \delta \rho, \delta T) \end{split}$$

> Therefore, the Jacobian of the resistive MHD model has the following coupling structure:

$$J\delta \vec{x} = \begin{bmatrix} D_{\rho} & 0 & 0 & U_{v\rho} \\ 0 & D_{T} & 0 & U_{vT} \\ 0 & 0 & D_{B} & U_{vB} \\ L_{\rho v} & L_{Tv} & L_{Bv} & D_{v} \end{bmatrix} \begin{pmatrix} \delta \rho \\ \delta T \\ \delta \vec{B} \\ \delta \vec{v} \end{pmatrix}$$

Diagonal blocks contain advection-diffusion contributions, and are "easy" to invert using MG techniques. Off diagonal blocks L and U contain all hyperbolic couplings.



Block factorization of resistive MHD

► We consider the block structure:

$$J\delta\vec{x} = \begin{bmatrix} M & U \\ L & D_v \end{bmatrix} \begin{pmatrix} \delta\vec{y} \\ \delta\vec{v} \end{pmatrix} ; \ \delta\vec{y} = \begin{pmatrix} \delta\rho \\ \deltaT \\ \delta\vec{B} \end{pmatrix} ; \ M = \begin{pmatrix} D_\rho & 0 & 0 \\ 0 & D_T & 0 \\ 0 & 0 & D_B \end{pmatrix}$$

► *M* is "easy" to invert (advection-diffusion, not very stiff, MG-friendly).

Schur complement analysis of 2x2 block J yields:

$$\begin{bmatrix} M & U \\ L & D_v \end{bmatrix}^{-1} = \begin{bmatrix} I & 0 \\ -LM^{-1} & I \end{bmatrix} \begin{bmatrix} M^{-1} & 0 \\ 0 & P_{Schur}^{-1} \end{bmatrix} \begin{bmatrix} I & -M^{-1}U \\ 0 & I \end{bmatrix},$$
$$P_{Schur} = D_v - LM^{-1}U.$$

EXACT Jacobian inverse only requires M^{-1} and P_{Schur}^{-1}



Physics-based preconditioner (PBP)

► 3-step EXACT inversion algorithm:

Predictor :
$$\delta \vec{y}^* = -M^{-1}G_y$$

Velocity update : $\delta \vec{v} = P_{Schur}^{-1} [-G_v - L\delta \vec{y}^*], \quad P_{Schur} = D_v - LM^{-1}U$
Corrector : $\delta \vec{y} = \delta \vec{y}^* - M^{-1}U\delta \vec{v}$

> MG treatment of P_{Schur} is impractical due to M^{-1} .

WE CONSIDER HERE THE SMALL-FLOW LIMIT: $v \ll v_A \Rightarrow M^{-1} \approx \Delta t \mathbb{I}$ ("CHEAP")

> We have extended the formulation to arbitrary-flows, $v \sim v_A$ based on commutation ideas¹ (more expensive, but more robust²).

¹Elman, *SISC* **2**7, 1651 (2006) ²L. Chacón, *J. Physics: Conf. Series*, **125**, 012041 (2008)



PBP: Small-flow limit

> Small flow approximation: $M^{-1} \approx \Delta t \mathbb{I}$ in steps 2 & 3 of Schur algorithm:

$$\begin{split} \delta \vec{y}^* &= -M^{-1} G_y \\ \delta \vec{v} &\approx P_{SI}^{-1} \left[-G_v - L \delta \vec{y}^* \right] ; \ P_{SI} = D_v - \Delta t L U \\ \delta \vec{y} &\approx \delta \vec{y}^* - \Delta t U \delta \vec{v} \end{split}$$

where:

$$P_{SI} = \rho^{n} \left[\mathbb{I} / \Delta t + \theta (\vec{v}_{0} \cdot \nabla \mathbb{I} + \mathbb{I} \cdot \nabla \vec{v}_{0} - \nu^{n} \nabla^{2} \mathbb{I}) \right] + \Delta t \theta^{2} W(\vec{B}_{0}, p_{0})$$
$$W(\vec{B}_{0}, p_{0}) = \vec{B}_{0} \times \nabla \times \nabla \times \left[\mathbb{I} \times \vec{B}_{0} \right] - \vec{j}_{0} \times \nabla \times \left[\mathbb{I} \times \vec{B}_{0} \right] - \nabla \left[\mathbb{I} \cdot \nabla p_{0} + \gamma p_{0} \nabla \cdot \mathbb{I} \right]$$

> Operator $W(\vec{B}_0, p_0)$ is ideal MHD energy operator, which has real eigenvalues!

- \blacktriangleright P_{SI} is parabolic, and hence block diagonally dominant by construction!
- > We employ multigrid methods (MG) to approximately invert P_{SI} and M: 1 V(4,4) cycle



Grid convergence study ($\Delta t = 1.0 au_A$)					
Ν	$GMRES/\Delta t$	CPU_{exp}/CPU	$\Delta t / \Delta t_{CFL}$		
32x32	14	2.43	159		
64×64	11.8	5.8	322		
128x128	11.2	13.3	667		
256x256	11.4	28.5	1429		

PBP: 2D serial performance (tearing mode)

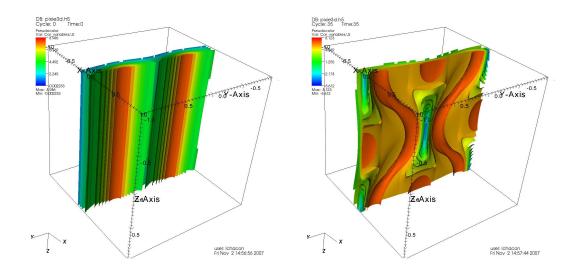
 $CPU \sim \mathcal{O}(N)$ OPTIMAL SCALING!

Δt convergence study (128x128)					
Δt	$GMRES/\Delta t$	CPU_{exp}/CPU	$\Delta t / \Delta t_{CFL}$		
0.5	8.0	8.0	380		
0.75	9.5	10.0	570		
1.0	11.2	12.7	760		
1.5	14.6	14.6	1140		

 $CPU \sim \mathcal{O}(\Delta t^{-0.6})$ FAVORABLE SCALING!



PBP: 3D serial performance (island coalescence)

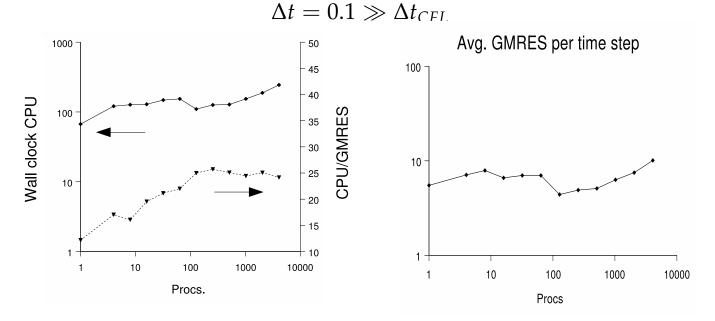


10 time steps, $\Delta t = 0.1$, V(3,3) cycles, mg_tol=1e-2

Grid	$GMRES/\Delta t$	CPU
16 ³	5.5	81
32 ³	7.9	1176
64 ³	7.0	11135



PBP: 3D *parallel* performance (island coalescence) (Weak scaling, 16³ points per processor, Cray XT4)



Key to parallel performance:

□ Matrix-light multigrid, where only diagonals are stored; residuals are calculated matrix-free.

- Operator coarsening via rediscretization: avoids forming/communicating a matrix.
- > Current limitations: we do not feature a coarse-solve beyond the processor skeleton grid.
 - \Box This eventually degrades algorithmic scalability (only shows at > 1000-processor level).



Implicit extended MHD solver



Extended (two-fluid, Hall) MHD model equations

$$\begin{aligned} \frac{\partial \rho}{\partial t} &+ \nabla \cdot (\rho \vec{v}) = 0, \\ \frac{\partial \vec{B}}{\partial t} &+ \nabla \times \vec{E} = 0, \\ \frac{\partial (\rho \vec{v})}{\partial t} + \nabla \cdot \left[\rho \vec{v} \vec{v} - \vec{B} \vec{B} &+ \overleftarrow{\Pi} + \overleftarrow{I} (p + \frac{B^2}{2}) \right] = 0, \\ \frac{\partial T_e}{\partial t} + \vec{v}^* \cdot \nabla T_e &+ (\gamma - 1) T_e \nabla \cdot \vec{v}^* = (\gamma - 1) \frac{Q - \nabla \cdot \vec{q}}{(1 + \alpha) \rho}, \end{aligned}$$

$$\overleftrightarrow{\Pi} = \overleftrightarrow{\Pi_i} + \overleftrightarrow{\Pi_e} ; \ \overleftrightarrow{\Pi_e} = -\nu_e \nabla \vec{v}_e ; \ \vec{v}_e = \vec{v} - d_i \frac{j}{\rho} ; \ \vec{v}^* = \vec{v} - \frac{d_i}{1 + \alpha} \frac{j}{\rho} ; \ \alpha = \frac{T_i}{T_e}$$

Ohm's Law : $\begin{cases} \vec{E} = -\vec{v} \times \vec{B} + \eta \vec{j} + \frac{d_i}{\rho} (\vec{j} \times \vec{B} - \nabla p_e - \nabla \cdot \overleftrightarrow{\Pi_e}) & \text{electron EOM} \\ \vec{E} = -\vec{v} \times \vec{B} + \eta \vec{j} + d_i [\partial_t \vec{v} + \vec{v} \cdot \nabla \vec{v} + \frac{1}{\rho} (\nabla p_i + \nabla \cdot \overleftrightarrow{\Pi_i})] & \text{ion EOM} \end{cases}$

Note that $EOM_i - EOM_e = EOM$. Admits an energy principle.

This model supports fast dispersive waves $\omega \sim k^2$.



Extended MHD Jacobian block structure: electron EOM (standard choice)

$$\vec{E} = -\vec{v} \times \vec{B} + \eta \vec{j} + \left[\frac{d_i}{\rho} (\vec{j} \times \vec{B} - \nabla p_e - \nabla \cdot \overleftarrow{\Pi_e}) \right]$$

> Linearized induction equation $\delta \vec{B} = -\nabla \times \delta \vec{E}$ has the following couplings:

$$\delta \vec{B} = L_B(\delta \vec{B}, \delta \vec{v}, \delta \rho, \delta T)$$

Jacobian coupling structure:

$$J\delta \vec{x} = \begin{bmatrix} D_{\rho} & 0 & 0 & U_{v\rho} \\ L_{TB} & D_{T} & U_{BT} & U_{vT} \\ L_{\rho B} & L_{TB} & D_{B} & U_{vB} \\ L_{\rho v} & L_{Tv} & L_{Bv} & D_{v} \end{bmatrix} \begin{pmatrix} \delta \rho \\ \delta T \\ \delta \vec{B} \\ \delta \vec{v} \end{pmatrix}$$

- > We have added off-diagonal couplings to block M.
- **Stiffest block** is $D_B \Rightarrow$ breaks approximations in block-factorization approach. UNSUITABLE!



Extended MHD Jacobian block structure: ion EOM

$$\vec{E} \approx -\vec{v} \times \vec{B} + \eta \vec{j} + \left[d_i [\partial_t \vec{v} + \vec{v} \cdot \nabla \vec{v} + \frac{1}{\rho} (\nabla p_i + \nabla \cdot \overleftarrow{\Pi_i} [\vec{v}])] \right]$$

- > Hall coupling is mainly via $\partial_t \vec{v}$.
- ► Jacobian coupling structure becomes:

$$J\delta \vec{x} \approx \begin{bmatrix} D_{\rho} & 0 & 0 & U_{v\rho} \\ 0 & D_{T} & 0 & U_{vT} \\ 0 & 0 & D_{B} & U_{vB}^{R} + U_{vB}^{H} \\ L_{\rho v} & L_{Tv} & L_{Bv} & D_{v} \end{bmatrix} \begin{pmatrix} \delta \rho \\ \delta T \\ \delta \vec{B} \\ \delta \vec{v} \end{pmatrix}$$

We can therefore reuse ALL resistive MHD PC framework!



Extended MHD preconditioner

- ► Use same block factorization approach.
- > M block contains ion time scales only $\Rightarrow M^{-1} \approx \Delta t \mathbf{I}$ is a very good approximation
- ► Additional block U_{vB}^H :

$$P_{SI}\delta\vec{v} = \rho^n \left[\delta\vec{v}/\Delta t + \theta(\vec{v}_0 \cdot \nabla\delta\vec{v} + \delta\vec{v} \cdot \nabla\vec{v}_0 + \nabla\cdot\delta\overleftrightarrow{\Pi})\right] + \Delta t\theta^2 W(\vec{B}_0, p_0)\delta\vec{v}$$

$$W(\vec{B}_0, p_0) = \vec{B}_0 \times \nabla \times \nabla \times [\mathbb{I} \times \vec{B}_0 - \frac{d_i}{\theta \Delta t} \mathbb{I}] - \vec{j}_0 \times \nabla \times [\mathbb{I} \times \vec{B}_0] - \nabla [\mathbb{I} \cdot \nabla p_0 + \gamma p_0 \nabla \cdot \mathbb{I}]$$

- > Additional term brings in dispersive waves $\omega \sim k^2!$
- We can show analytically that additional term (yellow) is amenable to simple damped Jacobi smoothing!

We can use classical MG!



On the issue of dissipation in extended MHD

- > Dispersive waves $\omega \sim k^2$ require higher order dissipation.
- Resistivity is unable to provide a dissipation scale.
- > Dissipation scale defined by electron viscosity, $\nabla \cdot \delta \overleftrightarrow{\Pi_e}$:

$$abla \cdot \delta \overleftrightarrow{\Pi_e} \sim -
u_e
abla^2 (
abla imes
abla imes \delta ec v) \sim
u_e
abla^4 \delta ec v$$

Viscosity coefficient can be determined to provide adequate dissipation of dispersive waves

$$\omega \sim v_A d_i k^2 : \nu_e k^4 \Rightarrow v_e > C \frac{d_i v_A k_{\parallel,max}}{k_{max}^3}$$

► In the preconditioner, we deal with $\nabla \cdot \delta \overleftarrow{\Pi_e}$ by considering 2 second-order systems, and solving them coupled within MG.



Extended MHD performance results (2D tearing mode)

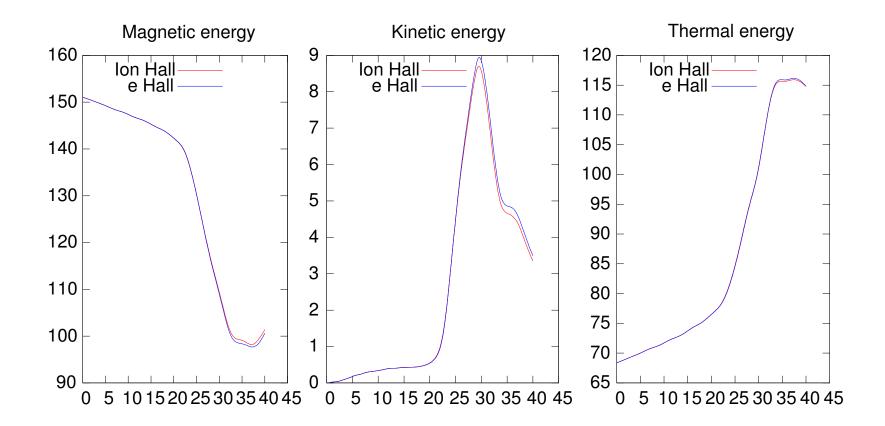
$$d_i = 0.05, \, \nu_e = 2.5 \times 10^{-6}$$

100 time steps, $\Delta t = 1.0$, 1 V(4,4) MG cycle

Grid	$GMRES/\Delta t$	CPU_{exp}/CPU	$\Delta t / \Delta t_{exp}$	$\Delta t / \Delta t_{CFL}$
32x32	22.3	0.74	135	110
64x64	15.4	10.9	1582	384
128x128	10.6	214	23809	1436
256x256	13.1	3097	370370	5660



2D nonlinear verification: GEM challenge Ion Hall vs. electron Hall



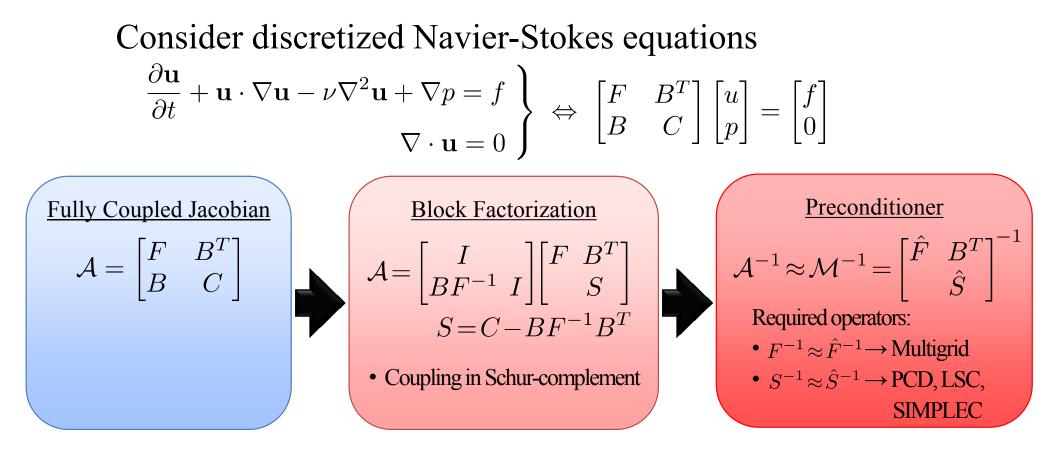


Incompressible Navier-Stokes solver

Cyr, Shadid, Tuminaro, JCP 2011 Elman, Howle, Shadid, Shuttleworth, Tuminaro, JCP 2008



Block preconditioning: CFD example



Properties of block factorization

- 1. Important coupling in Schur-complement
- 2. Better targets for AMG \rightarrow leveraging scalability

Properties of approximate Schur-complement

- 1. "Nearly" replicates physical coupling
- 2. Invertible operators \rightarrow good for AMG

Brief Overview of Block Preconditioning Methods for Navier-Stokes: (A Taxonomy based on Approximate Block Factorizations, JCP – 2008)

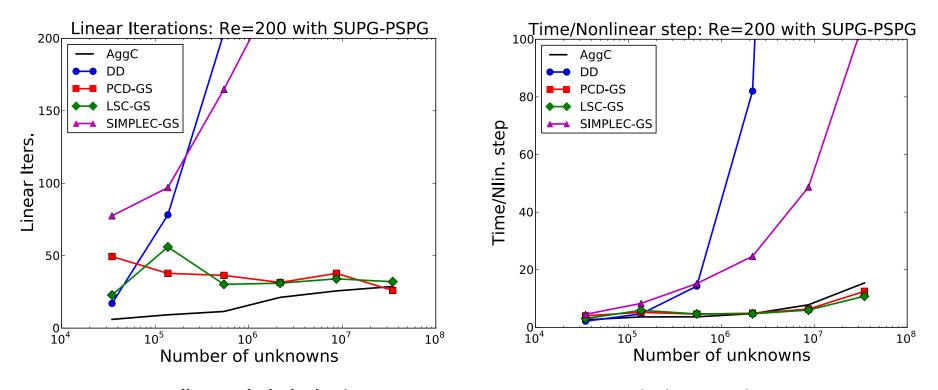
Discrete N-S			Exact LDU Factorization Approx		. LDU		
$ \begin{pmatrix} F & B^T \\ \hat{B} & -C \end{pmatrix} \begin{pmatrix} \mathbf{\Delta} \mathbf{u_k} \\ \Delta p_k \end{pmatrix} = \begin{pmatrix} \mathbf{g_u^k} \\ g_p^k \end{pmatrix} $		$ \begin{pmatrix} I & 0 \\ \hat{B}F^{-1} & I \end{pmatrix} \begin{pmatrix} F & 0 \\ 0 & -S \end{pmatrix} \begin{pmatrix} I & F^{-1}B^T \\ 0 & I \end{pmatrix} $		$\begin{bmatrix} I & 0 \\ \hat{B}H_1 & J \end{bmatrix}$	$\begin{bmatrix} F & 0 \\ 0 & -\hat{S} \end{bmatrix} \begin{bmatrix} I & H \\ 0 \end{bmatrix}$	$\begin{bmatrix} I_2 B^T \\ I \end{bmatrix}$	
		$S = C + \hat{B}F^{-1}B^T$					
Precond. Type	H_1		H_2	\hat{S}		References	
Pres. Proj; 1 st Term Neumann Series	$\mathbf{F^{-1}}$	$(\Delta t I)^{-1}$		$\mathbf{C} + \mathbf{\Delta} \mathbf{t} \hat{\mathbf{B}} \mathbf{B}^{\mathbf{T}}$		Chorin(1967);Temam (1969); Perot (1993): Quateroni et. al. (2000) as solvers	
SIMPLEC	$\mathbf{F^{-1}}$	(diag($\sum \mathbf{F}))^{-1}$	$\mathbf{C} + \mathbf{\hat{B}}(\mathbf{diag}(\sum \mathbf{F}$	$))^{-1}\mathbf{B^T}$	Patankar et. al. (19 solvers; Pernice ar (2001) as smoothe	nd Tocci
Pressure Convection / Diffusion	0	\mathbf{F}	-1	$ m A_p F_p^{-1}$		Kay, Loghin, Watha Silvester, Elman (1 2006); Elman, How Shuttleworth, Tum (2003,2008)	999 - Ie, S.,

Now use AMG type methods on sub-problems. Momentum transient convection-diffusion: $F\Delta u = r_u$

Pressure – Poisson type:

 $-\hat{S}\Delta p = \mathbf{r}_p$

CFD Weak Scaling: Steady Backward Facing Step

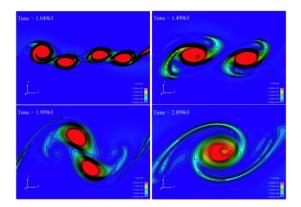


Fully coupled Algebraic AggC: Aggressive Coarsening Multigrid DD: Additive Schwarz Domain Decomposition Block Preconditioners PCD & LSC: Commuting Schur complement SIMPLEC: "Physics-based" Schur complement

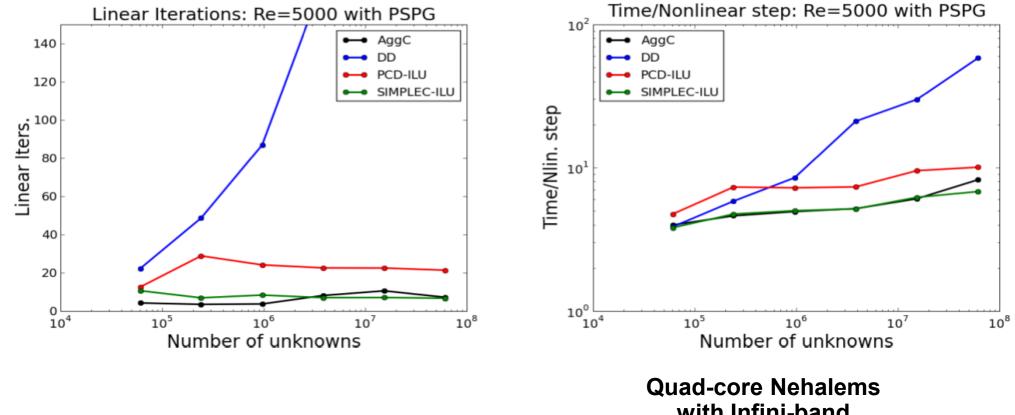
Take home: Block preconditioners competitive with fully coupled multigrid for CFD

* Paper accepted: E. C. Cyr, J. N. Shadid, R. S. Tuminaro, Stabilization and Scalable Block Preconditioning for the Navier-Stokes Equations, Accepted by J. Comp. Phys., 2011.

Weak Scaling of NK Solver with Fullycoupled AMG and Approx. Block Factorization Preconditioners



Transient Kelvin-Helmholtz instability (Re = 5 x 10³ shear layer, constant CFL = 2.5)



with Infini-band SNL Red Sky

Incompressible MHD solver





Magnetohydrodynamics (MHD) equations couple fluid flow to Maxwell's equations

$$\begin{aligned} \frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} - \nu \nabla^2 \mathbf{u} + \nabla p + \nabla \cdot \left(-\frac{1}{\mu_0} \mathbf{B} \otimes \mathbf{B} + \frac{1}{2\mu_0} \|\mathbf{B}\|^2 \mathbf{I} \right) &= f \\ \nabla \cdot \mathbf{u} &= 0 \\ \frac{\partial A_z}{\partial t} + \mathbf{u} \cdot \nabla A_z - \frac{\eta}{\mu_0} \nabla^2 A_z &= -E_z^0 \end{aligned}$$

where $\mathbf{B} = \nabla \times \mathbf{A}, \ \mathbf{A} = (0, 0, A_z)$

Discretized using a stabilized finite element formulation

Block LU Factorization

$$\begin{bmatrix} F & B^T & Z \\ B & C & 0 \\ Y & 0 & D \end{bmatrix} = \begin{bmatrix} I & & & \\ BF^{-1} & I & & \\ YF^{-1} & -YF^{-1}B^TS^{-1} & I \end{bmatrix} \begin{bmatrix} F & B^T & Z \\ S & -BF^{-1}Z \\ & P \end{bmatrix}$$

where

$$S = C - BF^{-1}B^{T}$$
$$P = D - YF^{-1}(I + B^{T}S^{-1}BF^{-1})Z$$

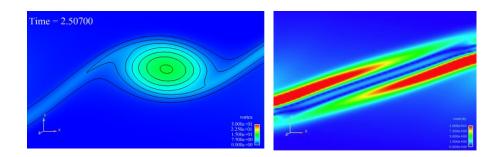
- C is zero for mixed interpolation FE and staggered FV methods, nonzero for stabilized FE
- Indefinite system hard to solve with incomplete factorizations without pivoting
- Block factorization of 3x3 system leads to nested Schur complements
- Use an operator splitting approximation to factor

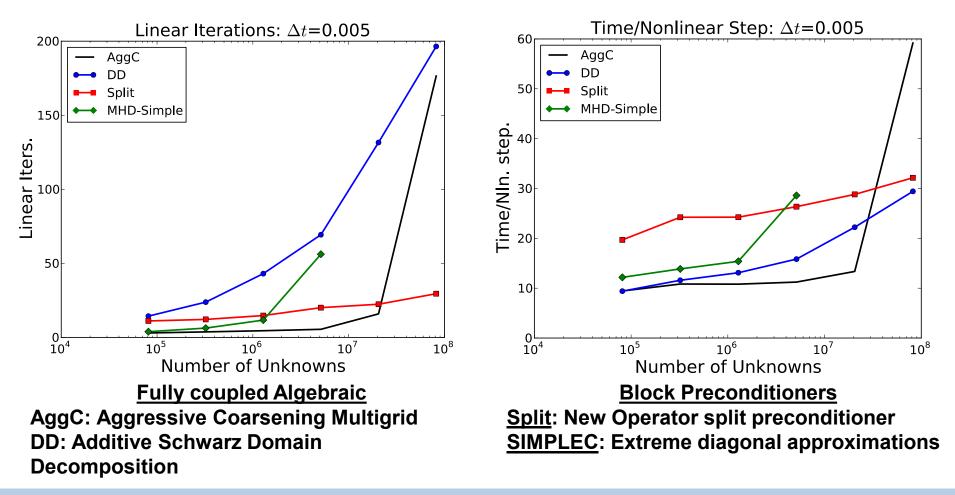
$$\begin{bmatrix} F & B^T & Z \\ B & C & 0 \\ Y & 0 & D \end{bmatrix} \approx \begin{bmatrix} F & Z \\ I \\ Y & D \end{bmatrix} \begin{bmatrix} F^{-1} & & \\ I & \\ & I \end{bmatrix} \begin{bmatrix} F & B^T & \\ B & C \\ & & I \end{bmatrix} = \begin{bmatrix} F & B^T & Z \\ B & C \\ Y & YF^{-1}B^T & D \end{bmatrix}$$

- Reduces to 2 2x2 systems for Navier-Stokes and magnetics-velocity blocks;
- C need not be non-zero or invertible (C⁻¹ doesn't need to exist!)



Transient Hydro-Magnetic Kelvin-Helmholtz Problem (Re = 700, S = 700)

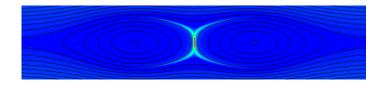


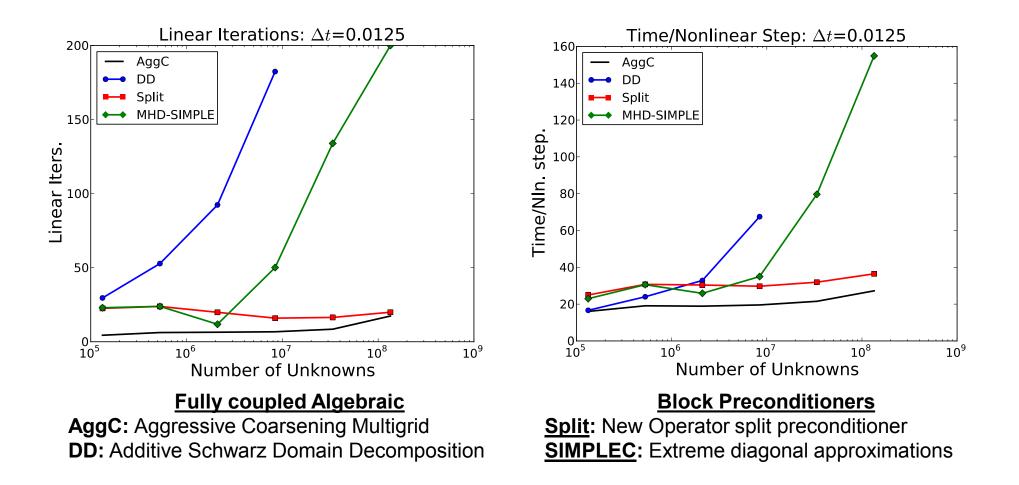


Take home: AggC and Split preconditioner scale algorithmically

- 1. SIMPLE preconditioner performance suffers with increased CFL
- 2. Run times are for unoptimized code
- 3. AggC not applicable to mixed discretizations, block factorization is

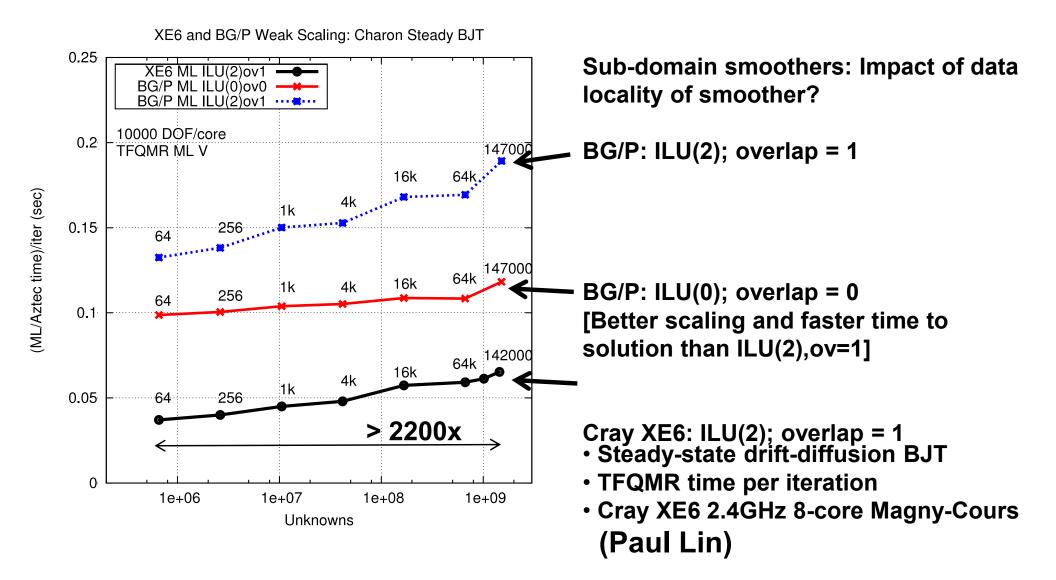
Driven Magnetic Reconnection: Magnetic Island Coalescence Half domain symmetry on [0,1]x[-1,1]with S = 10e+4





Take home: Split preconditioner scales algorithmically

Initial Weak Scaling Performance of AMG V-cycle on Leadership Class Machines Cray XE6 and BG/P Weak Scaling (Transport-reaction: Drift-diffusion simulations)



Summary and Conclusions

- **Stiff hyperbolic PDEs** describe many applications of interest to DOE.
- In applications where fast time scales are parasitic, an implicit treatment is possible to bridge time-scale disparity.
- A fully implicit solution may only realize its efficiency potential if a suitable scalable algorithmic route is available.
- Here, we have identified stiff-wave block-preconditioning (aka physics-based preconditioning) in the context of JFNK methods as a suitable algorithmic pathway.
 - An important property is that it renders the numerical system suitable for multilevel preconditioning.
- ➤ We have demonstrated the effectiveness of the approach in incompressible Navier-Stokes, incompressible MHD, and compressible resistive MHD and extended MHD.
 - In all these applications, the approach is robust and scalable, both algorithmically and in parallel.

