Scalable parallel implicit solvers for 3D magnetohydrodynamics

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Outline

- Motivation: XMHD and the tyranny of scales
- Parabolization of XMHD: key for SCALABILITY
- 3D compressible resistive MHD
- 3D extended MHD
- Massively parallel performance



"The tyranny of scales"



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





Algorithmic challenges in XMHD

- XMHD has mixed character, with strongly hyperbolic and parabolic components.
- Numerically, XMHD is a nonlinear algebraic system of very stiff equations:
 - Elliptic stiffness (diffusion): $\kappa(J) \sim \frac{\Delta t D}{\Delta x^2} \gg 1$
 - Hyperbolic stiffness (linear and dispersive waves): $\kappa(J) \sim \Delta t \, \omega_{fast} \sim \frac{\Delta t}{\Delta t_{CFL}} \gg 1$
- Brute-force algorithms will not be able to cover the span between disparate time/length scales, regardless of computer power (SBES report).
- Key algorithmic requirement: SCALABILITY $[CPU \sim \mathcal{O}(N/n_p)]!$
 - Minimize number of degrees of freedom N: spatial adaptivity.
 - Follow slowest time scales (application dependent): implicit time stepping.
- Scalable implicit methods require MULTILEVEL approaches:

$$CPU \sim \mathcal{O}\left(rac{N\log(N)}{n_p^{\beta}}
ight)$$
 , $\beta \lesssim 1$





Why multilevel (multigrid) methods?

- 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!
- In general, smoothers are easy to find for parabolic systems: PARABOLIZATION!





Parabolization and Schur complement: an example

• PARABOLIZATION EXAMPLE:

$$\partial_t u = \partial_x v$$
, $\partial_t v = \partial_x u$.
 $u^{n+1} = u^n + \Delta t \partial_x v^{n+1}$, $v^{n+1} = v^n + \Delta t \partial_x u^{n+1}$

$$(I - \Delta t^2 \partial_{xx})u^{n+1} = u^n + \Delta t \partial_x v^n$$

• PARABOLIZATION via SCHUR COMPLEMENT:

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

Stiff off-diagonal blocks L, U now sit in diagonal via Schur complement $D_1 - UD_2^{-1}L$. The system has been "PARABOLIZED."

$$D_1 - UD_2^{-1}L = (I - \Delta t^2 \partial_{xx})$$





Our approach to a successful fully implicit algorithm for XMHD

- Even if a smoother exists, MG is remarkably temperamental.
- Combination of Krylov methods and MG preconditioning is optimal:
 - MG preconditioning provides scalability
 - Krylov solver provides robustness

We seek to develop a successful algorithm for XMHD based on Newton-Krylov-MG

- Rest of the talk focuses on preconditioning development!
- Prove the concept in resistive MHD, and then move to XMHD.





Massively parallel performance of 3D resistive MHD with PETSc toolkit





Jacobian-Free Newton-Krylov Methods

- Objective: solve nonlinear system $\vec{G}(\vec{x}^{n+1}) = \vec{0}$ efficiently (scalably).
- Converge nonlinear couplings using Newton-Raphson method:
- Jacobian-free implementation:

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

 $\frac{\partial \vec{G}}{\partial \vec{x}}$

 $\delta \vec{x}_k = -\vec{G}(\vec{x}_k)$

- 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; they only affect efficiency!

• The rest of the talk will discuss the development of suitable preconditioners $P_k!$





3D resistive MHD implicit solver

L. Chacón, Phys. Plasmas 15, 056103 (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:

$$T\deltaec{x} = \left[egin{array}{cccc} D_{
ho} & 0 & 0 & U_{v
ho} \ 0 & D_T & 0 & U_{vT} \ 0 & 0 & D_B & U_{vB} \ L_{
hov} & L_{Tv} & L_{Bv} & D_v \end{array}
ight] \left(egin{array}{c} \delta
ho \ \delta T \ \delta ec{B} \ \delta ec{v} \end{array}
ight)$$

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





PARABOLIZATION: Schur complement formulation

• 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, 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}
- Schur complement formulation is fundamentally unchanged in Hall MHD!





Physics-based preconditioner (I): small-flow approximation

• The Schur complement analysis translates into the following 3-step EXACT inversion algorithm:

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

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

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

Need suitable simplifications (SEMI-IMPLICIT)!

- We consider the small-flow-limit case: $M^{-1} \approx \Delta t$
- This approximation is equivalent to splitting flow in original equations.



Physics-based preconditioner (II)

• Small flow approximation: $M^{-1} \approx \Delta t$ 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:

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

- *P*_{SI} is block diagonally dominant by construction!
- We employ multigrid methods (MG) to approximately invert P_{SI} and M: 1 V(4,4) cycle



Serial performance (2D tearing mode)

| _ | 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 | | |

Grid convergence study ($\Delta t = 1200\Delta t_{CFL}$)

| 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 |

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





Serial performance (3D 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 |





Application: m=n=1 kink mode in 3D CDXU Tokamak

64x32x32 mesh with 16 processors





3D extended MHD implicit solver



Extended MHD model equations

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

- Plasma is assumed polytropic $p \propto n^{\gamma}$.
- We assume cold ion limit: $T_i \ll T_e \Rightarrow \boxed{p \approx p_e}$
- Generalized Ohm's law:

$$ec{E} = -ec{v} imes ec{B} + \eta
abla imes ec{B} - rac{d_i}{
ho} (ec{j} imes ec{B} -
abla p_e)$$





Extended MHD Jacobian block structure

• The linearized extended MHD model has the following couplings:

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

• Jacobian coupling structure:

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

• We have added off-diagonal couplings.





Extended MHD Jacobian block structure (cont.)

• The coupling structure can be substantially simplified if we note $(p \approx p_e)$:

$$\frac{1}{\rho}(\vec{j}\times\vec{B}-\nabla p_e)\approx\frac{D\vec{v}}{Dt}$$

and therefore:

$$\vec{E} \approx -\vec{v} \times \vec{B} + \eta \nabla \times \vec{B} - d_i \frac{D\vec{v}}{Dt}$$

• This transforms jacobian coupling structure to:

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

We can therefore reuse ALL resistive MHD PC framework!





Extended MHD preconditioner

- Use same Schur complement approach.
- *M* block contains ion scales only! Approximation $M^{-1} \approx \Delta t$ is very good in extended MHD (ion scales do NOT contribute to numerical stiffness).
- Additional block U_{vB}^{H} results, after the Schur complement treatment, in new term in Schur complement:

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

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

- This system supports dispersive waves $\omega \sim k^2!$
- We have shown analytically that additional term is amenable to damped JB smoothing!

We can use classical MG!





Preliminary serial performance results (2D tearing mode)

 $d_i = 0.05$

1 time step, $\Delta t = 1.0$, V(3,3) cycles, mg_tol=1e-2

| Grid | $GMRES/\Delta t$ | CPU_{exp}/CPU | $\Delta t / \Delta t_{exp}$ |
|---------|------------------|-----------------|-----------------------------|
| 32x32 | 32 | 0.9 | 110 |
| 64×64 | 9 | 9.3 | 384 |
| 128x128 | 8 | 37.9 | 1436 |
| 256x256 | 10 | 117.0 | 5660 |



Massively parallel performance

L. Chacón, Phys. Plasmas 15, 056103 (2008)



Approach to parallelization

- We employ the PETSc toolkit (Parallel Extensible Toolkit for Scientific computing)
- In our preconditioner, we have parallelized our own multigrid solver (using PETSc's distributed arrays constructs)
 - We implement a matrix-light multigrid, where only diagonals are stored; residuals are calculated matrix-free.
 - Operator coarsening is done via rediscretization (instead of Galerkin procedure); this avoids forming a matrix at any grid level: avoids communication of matrix elements.
 - Prolongation and restriction are performed with second-order splines locally at each processor: no communication required.
- As a result, our MG implementation features excellent parallel scalability.
- We do not feature a coarse-solve beyond the processor skeleton grid.
 - This eventually degrades algorithmic scalability (only shows at > 1000-processor level).





Massively parallel performance with PETSc toolkit (3D island coalescence, 16³ grid points per processor, on Cray XT (Franklin) at NERSC)





Conclusions

• Developed a scalable, multilevel-based, fully implicit NK-MG solver for XMHD.

Key algorithmic breakthrough: PARABOLIZATION + MG.

- Equivalence between parabolization and the Schur decomposition:
 - Provides a rigorous foundation for the parabolization step.
 - Provides a path to generalize approach when more complete XMHD models are considered.
- Demonstrated excellent algorithmic performance under grid refinement and with time step.
- Demonstrated excellent parallel performance in massively parallel environments.
- Future work
 - Bring Hall MHD to production stage (high-order dissipation required).
 - Add MG coarse-grid solve to avoid algorithmic performance loss with thousands of processors.
 - Extend MG functionality to singular-point coordinate systems.



