Discussions on the MHD adaptive solvers in the AMROC framework for space plasmas applications

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Outline

Motivation MHD Equations Adaptive schemes MR refinement indicator AMROC framework Experiments





Space-physics context

Instabilities: in the magnetosphere and/or ionosphere regions











Image credits: MMS and CINDI NASA-Mission

Credits: Hasegawa et al. (2004 Nature)

http://aer.nict.go.jp/en/people/spe_yokoyama.html

Compressible Magnetohydrodynamic Equations I

$$\boxed{\frac{\partial \mathbf{q}}{\partial t} + \nabla \cdot \mathbf{f} = 0}$$
 with the solenoidal condition $\nabla \cdot \mathbf{B} = 0$



 $\rho = \rho(\mathbf{x}, t)$ is the density, $p = p(\mathbf{x}, t)$ is the pressure, **I** is a 3 × 3 identity matrix, $\mathbf{u} = \mathbf{u}(\mathbf{x}, t)$ is the velocity, $\mathbf{B} = \mathbf{B}(\mathbf{x}, t)$ is the magnetic field, $\mathbf{u}\mathbf{u}$ or **BB** stands for a 3 × 3 tensor, $\mathbf{x} = (x, y, z)$, and $e = e(\mathbf{x}, t)$ is the total energy density,

$$\boldsymbol{e} = \frac{\boldsymbol{p}}{\gamma - 1} + \rho \frac{\boldsymbol{u}^2}{2} + \frac{\boldsymbol{B}^2}{2}$$



Numerical aspects

- 1. Approximations of MHD Equations present a numerically and computationally challenging task
- 2. Uniform meshes are not practical when high resolution is only needed locally.
- 3. Our space adaptive strategy combines finite volumes with:
 - Wavelet based analysis: multiresolution (MR)

1989 Mallat after his book, Harten (1995, Comm. Pure Appl. Math.), Kaibara and Gomes (2000,

Klumer/Plenum), Cohen et al (2002, Math. Comp.), Roussel et al. (2003, J. Comput. Phys.)

Structured adaptive mesh refinement (SAMR)

[Berger and Olinger (1984, J. Comput. Phys.), Berger & Colella (1988, J. Comput. Phys.)]

4. Explicit time-space adaptation schemes are used.

Goal: Improvement of **computational efficiency** while controlling the precision using dynamically adapted meshes in space-time in an MPI parallel distributed memory patch-based AMROC framework.



This work context I

- We compared (serial based): MR: Carmen code [Roussel et al. (2003, J. Comput. Phys.)] with SAMR: AMROC framework [Deiterding (2011, ESAIM Proceedings), PhD Thesis, 2003] for Euler Eqs. [Deiterding (2009, ESAIM Proceedings); 2016 (Deiterding et al. SIAM JSciC)]
 - * MR representation is more compact, while the AMROC framework is much more efficient in terms of CPU time.
- $\hookrightarrow\,$ Then, MR alg. in the AMROC framework in its patch-based MPI parallel, exact same numerics.
 - We confirmed the superiority of MR compact representation in the patch-based structure. [Deiterding et al., submitted Comput.& Fluids, 2018]
- In meanwhile, we developed a resistive GLM-MHD MR serial solver. [Gomes et al. 2015, APNUM; PhD Thesis, 2017], then, we introduced this solver in AMROC patch based MPI parallel [Moreira Lopes et al. 2018, Comput.& Fluids, PhD Thesis, 2019].



This work context II

- \hookrightarrow Now, we present our results of
 - * Our MHD AMROC patch-based MPI parallel solver
 - * Wavelet based adaptive algorithm in this solver.

[Domingues et al. C&F 2019, Moreira Lopes, PhD Thesis 2019].



Adaptive schemes - main ideas I

$$\frac{\partial \mathbf{q}(\mathbf{x},t)}{\partial t} + \nabla \cdot \mathbf{f} \Big(\mathbf{q}(\mathbf{x},t) \Big) = \mathbf{0}$$

Numerical scheme:

(2D example)

$$\mathbf{Q}_{j,k}^{n+1} = \mathbf{Q}_{j,k}^{n} - \frac{\Delta t}{\Delta x} \left[\mathbf{F}_{j+\frac{1}{2},k}^{x} - \mathbf{F}_{j-\frac{1}{2},k}^{x} \right] - \frac{\Delta t}{\Delta y} \left[\mathbf{F}_{j,k+\frac{1}{2}}^{y} - \mathbf{F}_{j,k-\frac{1}{2}}^{y} \right]$$

▶ Reference scheme: $Q^{n+1} = D Q^n$

 $\mathbf{Q}^{\mathbf{n}} = \mathbf{Q}^{\mathbf{n},\mathbf{L}}$ cell average in a regular mesh \mathcal{M}^{L} em $t^{n} = n\Delta t$

 $\mathcal{D} = \mathcal{D}^{L,\Delta t}$ stable and consistent operator in the time evolution



Adaptive schemes - main ideas II

► Adaptive scheme: $(\mathcal{M}^n, \mathbf{Q}^n_a) \to (\mathcal{M}^{n+1}, \mathbf{Q}^{n+1}_a)$

 $\mathcal{M}^n \subset \mathcal{M}^L$ adaptive mesh Q^n_a cell average in \mathcal{M}^n

Main steps:

- 1. Refinement: $(\mathcal{M}^n, Q^n_a) \xrightarrow{\mathcal{R}} (\mathcal{M}^{n+}, Q^{n+}_a)$
- 2. Evolution: $(\mathcal{M}^{n+}, \mathbf{Q}^{n+}_{a}) \xrightarrow{\mathcal{D}_{a}} (\mathcal{M}^{n+}, \breve{\mathbf{Q}}^{n+1})$
- 3. Adaptation: $(\mathcal{M}^{n+}, \breve{\mathbf{Q}}_{a}^{n+1}) \xrightarrow{\mathcal{T}_{\epsilon}, \eta_{\text{tol}}} (\mathcal{M}^{n+1}, \mathbf{Q}_{a}^{n+1}).$
 - Cluster organization: if it is the case



Multiresolution (MR) MR principles of interest: I

Two-level wavelet transform:

$$\boldsymbol{\mathsf{Q}}^{\ell+1} \rightleftharpoons \; \{\boldsymbol{\mathsf{Q}}^\ell\} \;\cup\; \{\boldsymbol{\mathsf{d}}^\ell\},$$

- Information in a certain level can be obtained by the combination of the coarser levels with the wavelet coefficient contributions and vice-versa
- ► The algorithm are efficient: fast and stable.
- Wavelet coefficients \mathbf{d}^{ℓ} are :
 - local approximation polynomial errors.
 - regularity indicators in adaptive strategies.
 - Iow amplitudes of the coefficients are associated to regions where the solution is smooth
 - high amplitudes appear only in regions where the solution is less regular.



Notation:

one-dimensional example

Nested meshes:

$$(\Omega_k^\ell)_{0\leq k< 2^\ell; 0\leq \ell\leq L}$$

Data: cell average value

on Ω_k^ℓ :



 \mathcal{V} – volume



Notation: $\mathbf{Q}^\ell = (\mathbf{Q}^\ell_{\mathbf{k}})_{\mathbf{0} \leq \mathbf{k} < \mathbf{2}^\ell}$ Harten, SIAM J. Num. Anal. 1993



MR operations for FV methods

1 Projection (restriction):

$$\left[P_{\ell+1}^{\ell}: \mathbf{Q}^{\ell+1}
ightarrow \mathbf{Q}^{\ell}
ight]$$

2 Prediction (prolongation):

$$P_\ell^{\ell+1}: \mathbf{Q}^\ell o \mathbf{ ilde Q}^{\ell+1}$$





$$P_{\ell+1}^{\ell} : \mathbf{Q}_{i}^{\ell} = \frac{1}{2} \left(\mathbf{Q}_{2i}^{\ell+1} + \mathbf{Q}_{2i+1}^{\ell+1} \right) \qquad P_{\ell,0}^{\ell+1} : \tilde{\mathbf{Q}}_{2i}^{\ell+1} = \mathbf{Q}_{i}^{\ell} - \frac{1}{8} (\mathbf{Q}_{i+1}^{\ell} - \mathbf{Q}_{i-1}^{\ell}),$$
$$P_{\ell,1}^{\ell+1} : \tilde{\mathbf{Q}}_{2i+1}^{\ell+1} = \mathbf{Q}_{i}^{\ell} + \frac{1}{8} (\mathbf{Q}_{i+1}^{\ell} - \mathbf{Q}_{i-1}^{\ell})$$

Linear polynomial interpolation as proposed by Harten (1995, Comm. Pure Appl. Math.).

Two-dimensional operators

Projection $P_{\ell+1}^{\ell}$:

$$\mathbf{Q}_{i,j}^{\ell} = \frac{1}{4} \left(\mathbf{Q}_{2i,2j}^{\ell+1} + \mathbf{Q}_{2i,2j+1}^{\ell+1} + \mathbf{Q}_{2i+1,2j}^{\ell+1} + \mathbf{Q}_{2i+1,2j+1}^{\ell+1} \right)$$

Prediction $P_{\ell}^{\ell+1}$ by Bihari and Harten (1995):

$$\begin{split} \tilde{\mathbf{Q}}_{2i+m,2j+n}^{\ell+1} &= \mathbf{Q}_{i,j}^{\ell} + \frac{1}{8} \left[(-1)^m \left(\mathbf{Q}_{i+1,j}^{\ell} - \mathbf{Q}_{i-1,j}^{\ell} \right) + (-1)^n \left(\mathbf{Q}_{i,j+1}^{\ell} - \mathbf{Q}_{i,j-1}^{\ell} \right) \right] \\ &+ \frac{1}{64} \left[(-1)^{mn} \left(\mathbf{Q}_{i+1,j+1}^{\ell} - \mathbf{Q}_{i+1,j-1}^{\ell} - \mathbf{Q}_{i-1,j+1}^{\ell} + \mathbf{Q}_{i-1,j-1}^{\ell} \right) \right] \end{split}$$

- Constructed by tensor product
- Note that these prediction operators are not TVD!



Three-dimensional operators

$$P_{\ell+1}^{\ell}: \quad \mathbf{Q}_{i,j}^{\ell} = \frac{1}{8} \left(\mathbf{Q}_{2i,2j,2k}^{\ell+1} + \mathbf{Q}_{2i,2j+1,2k}^{\ell+1} + \mathbf{Q}_{2i+1,2j,2k}^{\ell+1} + \mathbf{Q}_{2i+1,2j+1,2k}^{\ell+1} + \mathbf{Q}_{2i,2j,2k+1}^{\ell+1} + \mathbf{Q}_{2i,2j,2k+1}^{\ell+1} + \mathbf{Q}_{2i+1,2j,2k+1}^{\ell+1} + \mathbf{Q}_{2i+1,2j,2k+1}^{\ell+1} + \mathbf{Q}_{2i+1,2j+1,2k+1}^{\ell+1} \right)$$

Prediction $P_{\ell}^{\ell+1}$ after Roussel et al. (2003, J. Comput. Phys.):

$$\begin{split} \tilde{\mathbf{Q}}_{2i+m,2j+n,2k+p}^{\ell+1} &= \mathbf{Q}_{i,j,k}^{\ell} + \\ &\frac{1}{8} \left[(-1)^m \left(\mathbf{Q}_{i+1,j,k}^{\ell} - \mathbf{Q}_{i-1,j,k}^{\ell} \right) + (-1)^n \left(\mathbf{Q}_{i,j+1,k}^{\ell} - \mathbf{Q}_{i,j-1,k}^{\ell} \right) \right. \\ &+ (-1)^p \left(\mathbf{Q}_{i,j,k+1}^{\ell} - \mathbf{Q}_{i,j,k-1}^{\ell} \right) \right] \\ &+ \frac{1}{64} \left[(-1)^{mn} \left(\mathbf{Q}_{i+1,j+1,k}^{\ell} - \mathbf{Q}_{i+1,j-1,k}^{\ell} - \mathbf{Q}_{i-1,j+1,k}^{\ell} + \mathbf{Q}_{i-1,j-1,k}^{\ell} \right) \right. \\ &\left. (-1)^{np} \left(\mathbf{Q}_{i,j+1,k+1}^{\ell} - \mathbf{Q}_{i,j-1,k+1}^{\ell} - \mathbf{Q}_{i,j+1,k-1}^{\ell} + \mathbf{Q}_{i,j-1,k-1}^{\ell} \right) \right. \\ &\left. (-1)^{mp} \left(\mathbf{Q}_{i+1,j,k+1}^{\ell} - \mathbf{Q}_{i-1,j,k+1}^{\ell} - \mathbf{Q}_{i+1,j,k-1}^{\ell} + \mathbf{Q}_{i-1,j,k-1}^{\ell} \right) \right] \\ &+ \frac{1}{512} \left[\mathbf{Q}_{i+1,j+1,k+1}^{\ell} - \mathbf{Q}_{i+1,j+1,k-1}^{\ell} - \mathbf{Q}_{i+1,j-1,k+1}^{\ell} - \mathbf{Q}_{i-1,j-1,k+1}^{\ell} \right] \right] \end{split}$$

Proprieties:

- Localization: finite stencil and near the cell.
- Exact polynomial approximation.
- Stability
- **Conservation**: $P_{\ell+1}^{\ell} P_{\ell}^{\ell+1}$ is the identity operator.
- Conservation implies $P_{\ell+1}^{\ell} \mathbf{d}^{\ell} = 0$.

Adaptive transform

► Threshold operation: → data compression modulus of the wavelet coefficients bellow certain tolerance are removed.

MR threshold strategies: hard threshold, level dependent threshold, vector-value threshold.

Buffer zones



Use of wavelet transform for adaptation

Wavelet coefficients:

$$\mathbf{d}^{\ell} = \mathbf{Q}^{\ell+1} - \mathbf{P}^{\ell+1}_{\ell} \, \mathbf{Q}^{\ell}$$
 prediction error

Use of predicton error as refinement criterion:

$$|\mathbf{Q}^{\ell} - P_{\ell-1}^{\ell} P_{\ell}^{\ell-1} \mathbf{Q}^{\ell}| > \epsilon$$

Choice of ϵ :



Use of wavelet transform for adaptation

Wavelet coefficients:

$$\mathbf{d}^{\ell} = \mathbf{Q}^{\ell+1} - \mathbf{P}_{\ell}^{\ell+1} \, \mathbf{Q}^{\ell}$$
 prediction error

Use of predicton error as refinement criterion:

$$|\mathbf{Q}^{\ell} - P_{\ell-1}^{\ell} P_{\ell}^{\ell-1} \mathbf{Q}^{\ell}| > \epsilon$$

Choice of ϵ :

- level-independent threshold parameter $\epsilon \equiv \epsilon_{\ell}$
- Harten's thresholding strategy:

$$\epsilon^{\ell} = rac{\epsilon}{|\Omega|} 2^{2(\ell+1-L)}, \ \ 0 \leq \ell < L$$

 vector-valued threshold in Eucledian norm of velocity field component of Q



Patch-SAMR ideas in AMROC

- * Block independence
- * Refined blocks overlay coarser ones [Berger & Colella (1988, J. Comput. Phys.)]
- ★ Refinement in space-time by a factor r_{ℓ}
- Patch based data structures
- * Efficient cache-reuse
- * Vectorization possible
- \Rightarrow Level transfer/ setting ghost cells
 - Restriction: Conservative averaging
 - Prolongation: Bilinear interpolation
- ⇒ Linear time interpolation for boundary conditions
- ⇒ Conservative flux correction





Credits: R. Deiterding, ESAIM 2006

Patch creation by clustering

Image detection technique to identify cluster signatures. Consider: mesh as a cluster.

Recursive cluster identification:

- Compute # of flagged cells ↑ per row/column.
- 2. Identify where $\Upsilon_{\text{row/col}} = 0$.
- 3. Compute 2^{nd} difference:

 $\Delta = \Upsilon_{k+1} - 2\Upsilon_k + \Upsilon_{k-1}.$

 Localize the largest difference in Δ_{row/col}.

5. Stop flagged cells $\eta_{tol} < \eta_{tol}$

[Deiterding (2011, ESAIM Proceedings); Bell et. al (1994, SIAM J. Sci. Comput.); Berger (1986, SIAM J. Sci. Stat.

Comput.), and Berger & Rigoutsos (1991, IEEE Transactions on Systems)]

 $0 < \eta_{tol} < 1$

$$\eta_{\textit{tol}} = \left\{ \begin{array}{ll} \textbf{0.80} & \textit{parallel} \\ \textbf{0.99} & \textit{equiv. serial} \end{array} \right.$$

Numerical challenge I

The momentum equation

$$\frac{\partial \rho \mathbf{u}}{\partial t} + \nabla \cdot \left[\rho \mathbf{u} \mathbf{u} + \left(\boldsymbol{\rho} + \frac{B^2}{2} \right) \mathbf{I} - \mathbf{B} \mathbf{B} \right] = \mathbf{0},$$

can be rewritten as

$$\frac{\partial \rho \,\mathbf{u}}{\partial t} + \nabla \cdot \left[\rho \,\mathbf{u} \,\mathbf{u}\right] + \left[\nabla \left(\rho + \frac{B^2}{2}\right)\right] - \left(\mathbf{B} \cdot \nabla\right) \mathbf{B} = \boxed{\mathbf{B} \left(\nabla \cdot \mathbf{B}\right)}$$

- $(\nabla \cdot \mathbf{B}) = 0$ implies the non-existence of magnetic monopoles.
- $(\nabla \cdot \mathbf{B}) \neq 0$ results in a non-physical force that is parallel to \mathbf{B} .
 - → Numerically, in the **standard FV uniform scheme** $(\nabla \cdot \mathbf{B}) \neq 0$ and this exerts a destabilizing effect on the numerical scheme or possible non-physical results. [Brackbill & Barnes, 1980 J. Comput. Phys.]



Implemented solution to reduce the effects $(\nabla \cdot \mathbf{B}) \neq 0$

- We use the parabolic-hyperbolic correction to avoid this problem in conjunction with a numerical formulation of the MHD equations called Generalized Lagrangian multipliers for divergence cleaning (GLM-MHD) proposed by [Dedner et al., 2002 J. Comput. Phys.] with the dimensional adjustment proposed by [Mignone et al., 2010 J. Comput. Phys.].
- GLM introduces changes in the induction equation, a scalar function ψ and a new equation, with the parameters c_h , c_p .
- Extended-GLM (EGLM) add also Powell's source terms. Non-conservative MHD system. [Dedner et al., 2002 J. Comput. Phys.]
- GLM (EGLM) does not zero the numerical effect but instead it transports and diffuses the numerical effect.
- Other possibility implemented is the traditional elliptic correction, it is very expensive computationally. It try to reduce this numerical effect. [Brackbill & Barnes, 1980 J. Comput. Phys.]
- \mathcal{D}_{B} ratio between the divergence and magnitude of the magnetic field per unit of volume



Challenging test cases

- Rotor (ROT)- propagation of strong torsional Alfvén waves. A high density fluid (the rotor) rotating in high velocity inside a lighter background magnetised fluid, angular momentum of the rotor to decrease and the magnetic field wraps around the rotor, increasing the magnetic pressure and compressing the fluid.
 - Initial conditions:
 - $\rightarrow \text{ Inside rotor } \xi \leq 0.1:$ (ρ, u_X, u_Y) = (10, -20 γ , -20x)
 - \rightarrow Background fluid $\varepsilon < 0.1$:

 $(\rho, u_x, u_y) = (1, 0, 0)$

→ Transition zone, elsewhere:

$$\left(9\phi+1,-\frac{2y\phi}{\xi},-\frac{2x\phi}{\xi}\right)$$

where $\phi = (0.115 - \xi)/0.015$,

- $\rightarrow \mathbf{B} = \left[\frac{5}{\sqrt{4\pi}}, 0, 0\right],$
- $\rightarrow~$ The rotor is a cylinder with centre in the origin and radius 0.1
- $\rightarrow \xi$ is distance to the origin.
- Comput. domain: $\left[-\frac{1}{2}, \frac{1}{2}\right] \times \left[-\frac{1}{2}, \frac{1}{2}\right]$,
- $T_{end} = 0.15, \alpha_p = 0.4, \text{CFL} = 0.4, \gamma = 1.4$
- Problem from Balsara and Spicer 1999.

- Orszag-Tang (OTV) transition to supersonic two-dimensional MHD
 - Initial cond.: ρ = γ², p = γ,
 - $\mathbf{u} = [-\sin(y), \sin(x), 0]$

$$\mathbf{B} = [-\sin(y), \sin(2x), 0].$$

- Comput. domain: [0, 2π] × [0, 2π],
- T_{end} = π, α_p = 0.5, CFL= 0.5
- Problem from Orszag-Tang (1979) and others.
- Spherical blast wave (BWV)- propagation of strong MHD discontinuities. Explosion of p sphere contained into a uniformly magnetised medium.

• Initial cond.:
$$\rho = 1$$
,
 $p = \begin{cases} 10, & \text{if } x^2 + y^2 < 0.1 \\ 0.1, & \text{elsewhere.} \end{cases}$,
 $\mathbf{u} = \mathbf{0}, \ \mathbf{B} = \left[\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0\right]$,
• Comput. domain: $\left[-\frac{1}{2}, \frac{1}{2}\right] \times \left[\frac{3}{4}, \frac{3}{4}\right]$

- Problem from Zachary et al. 1994 and others.
- For all cases: HLLD num. flux, MC limiter and periodic boundary conditions.









Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$

$$\rho(x, y, 0) = \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0$$
$$\rho(x, y, 0) = \gamma, \quad B_x(x, y, 0) = -\sin(y), \quad B_y(x, y, 0) = 2\sin(x), \quad B_z(x, y, 0) = 0$$



Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$

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Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$

Initial condition

$$\begin{aligned} \rho(x, y, 0) &= \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0 \\ p(x, y, 0) &= \gamma, \quad B_x(x, y, 0) = -\sin(y), \quad B_y(x, y, 0) = 2\sin(x), \quad B_z(x, y, 0) = 0 \end{aligned}$$

time=1.88496

time=1.88496



Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$

► Initial condition

$$\begin{aligned} \rho(x, y, 0) &= \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0 \\ \rho(x, y, 0) &= \gamma, \quad B_x(x, y, 0) = -\sin(y), \quad B_y(x, y, 0) = 2\sin(x), \quad B_z(x, y, 0) = 0 \end{aligned}$$

time=2.19911



time=2.19911

4.0

5.0

Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$

Initial condition

$$\begin{aligned} \rho(x, y, 0) &= \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0 \\ \rho(x, y, 0) &= \gamma, \quad B_x(x, y, 0) = -\sin(y), \quad B_y(x, y, 0) = 2\sin(x), \quad B_z(x, y, 0) = 0 \end{aligned}$$

time=2.51327



- Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$
- Initial condition

$$\rho(x, y, 0) = \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0$$

 $p(x, y, 0) = \gamma$, $B_x(x, y, 0) = -\sin(y)$, $B_y(x, y, 0) = 2\sin(x)$, $B_z(x, y, 0) = 0$

time=2.82743

time=2.82743



- Adaptive solution on 50 \times 50 grid with 4 additional levels refined by $r_l = 2$
- Initial condition

$$\begin{aligned} \rho(x, y, 0) &= \gamma^2, \quad u_x(x, y, 0) = -\sin(y), \quad u_y(x, y, 0) = \sin(x), \quad u_z(x, y, 0) = 0 \\ \rho(x, y, 0) &= \gamma, \quad B_x(x, y, 0) = -\sin(y), \quad B_y(x, y, 0) = 2\sin(x), \quad B_z(x, y, 0) = 0 \end{aligned}$$

time=3.14159

time=3.14159



Orszag-Tang vortex - cells on finest level vs. error



- This is work in progress, and for now, the error is evaluated in p only.
- Compared are SG and MR with hierarchical threshold also applied to ρ only.



Orzag-Tang vortex GLM parabolic-hyperbolic test case



- Refinement of additional features at the maximal level with the gradient criterion and the stronger coarsening.
- These effects are the reason why the MR criterion with hierarchical thresholding achieves still a smaller error
- MR avoids unnecessary over-refinement while avoiding improper coarsening.



Orzag-Tang vortex GLM parabolic-hiperbolic 3D test case

- Proposed by Helzel et al. (2011)
- a perturbation is added into the z axis of the velocity:
 - $\mathbf{u} = [-[1 + 0.2\sin(z)]\sin(y), [1 + 0.2\sin(z)]\sin(x), 0.2\sin(z)]$
- Computational domain $[0, 2\pi]^3$.
- Periodic boundaries.
- Adiabatic constant $\gamma = 5/3$, $\alpha_p = 0.3$ CFL= 0.3 until the final time $t_{end} = \pi$.

Proc. Distribution

3D pressure







[Moreira Lopes et al. 2018, Comput. & Fluids, PhD Thesis 2019]

Kelvin-Helmholtz instability (KHI) I

The KHI is a phenomena which occur in single continuous fluids with a velocity shear.

Initial conditions (cat eye):

ρ	<i>U</i> _x	u_y	B_x	р
1.0	u_x^0	u_y^0	1.0	50

- Periodic boundary conditions.
- Final time = 0.4
- Computational domain of [0, 1] × [−1, 1],

•
$$\gamma = 1.4, \alpha(c_h, c_p) = 0.4.$$

$$u_{x}^{0} = u_{x}^{0}(x, y) := 5 \tanh\left(20\left(y + \frac{1}{2}\right)\right) - \left[\tanh\left(20\left(y - \frac{1}{2}\right)\right) + 1\right]^{-1}$$
$$u_{y}^{0} = u_{y}^{0}(x, y) := \frac{1}{4}\sin(2\pi x)\left[e^{\left(-100\left(y + \frac{1}{2}\right)^{2}\right)} - e^{\left(-100\left(y - \frac{1}{2}\right)^{2}\right)^{2}}\right]$$
$$u_{z} = By = Bz = 0.$$



Moreira Lopes et al., 2018 Comput. & Fluids



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Dedner et al., 2002 J. Comput. Phys.

Instabilities: Rayleigh-Taylor (RTI) multi-modes

- * Initial cond.(single-mode):
 - $\rho = 2$ in the upper domain and 1 in the lower domain.
 - $B_X = 0.0125$
 - Gravitational acceleration g = 0.1 must be added to the equations of motion.
 - Pressure is given by the condition of hydrostatic equilibrium, that is $p = p_0 0.1\rho y$, where $p_0 = 2.5$. This gives a sound speed of 3.5 in the low density medium at the interface.
 - $u_y = 0.01 [1 + \cos(4\pi x)] [1 + \cos(3\pi y)] / 4.$
- Final time = 13.42
- ★ Computational domain of [-0.25, 0.25] × [-0.75, 0.75],
- * $\gamma = 1.4, c_h/c_p = 0.4, \rho, 200 \times 600, MR_{\epsilon} = 0.0001.$



★ Multi-modes

- ★ u_y = A [1 + cos(8πy/3)] /2, where A is a random number with a peak-to-peak amplitude of 0.01
- ★ Pertubation zone:[-0.1, 0.1]
- * Computational domain of

 $[-0.375, 0.375] \times [-0.75, 0.75].$







400 \times 1200, MR $\epsilon = 0.00005$

Test concept Jun et al., 2002 J. Comput. Phys.

* Periodic boundary cond. at |x| and reflecting at |y|.



Final remarks

- ► The AMROC MHD MR-version is implemented and verified.
- Patch-based MPI-parallel distributed memory version
- Considering precision, number of cells and CPU time this new MR implementation presents better results than the scaled gradient criterion.
- We also tested more complex vector-valued wavelet criteria
- Further optimization of further SAMR core parameters stil possible
- Performance could be further increased using methods already implemented in AMROC with some more work, for instance:
 - dimensional splitting;
 - wave propagation method (fully unsplit 2nd order scheme with Roe solver);
 - higher-order WENO methods



Earth's magnetosphere (validation, in testing)





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GLM and EGLM

GLM introduces changes in the induction equation, a scalar function ψ and a new equation, with the parameters c_h , c_p . For 2*D*, we have

$$\frac{\partial B_x}{\partial t} + \frac{\partial \psi}{\partial x} + \frac{\partial}{\partial y} (u_y B_x - B_y u_x) = 0,$$

$$\frac{\partial B_y}{\partial t} + \frac{\partial}{\partial x} (u_x B_y - B_x u_y) + \frac{\partial \psi}{\partial y} = 0,$$

$$\frac{\partial B_z}{\partial t} + \frac{\partial}{\partial x} (u_x B_z - B_z u_x) + \frac{\partial}{\partial y} (u_y B_z - B_y u_z) = 0,$$

$$\frac{\partial \psi}{\partial t} + c_h^2 \left(\frac{\partial B_x}{\partial x} + \frac{\partial B_y}{\partial y} \right) = -\frac{c_h^2}{c_0^2} \psi,$$

EGLM include the source terms introduced by Powell in as a tentative to improve the correction. These terms are not conservative, then this method fail in most cases.



Orzag-Tang vortex test case details I

The usual model problem for testing the transition to supersonic two-dimensional MHD turbulence is Orszag-Tang vortex.

Orszag and Tang, 1998 J. Fluid Mech.

It is used to tests how robust the code is at handling the formation of MHD shocks, and shock-shock interactions and it can also provide how significant the numerical magnetic monopoles affect the numerical solutions.

Frequently, this test is also used for code comparisons. This test can present different fields values depending on the units of the MHD model used.

Since then it has been extensively compared in tests of numerical MHD simulations. A few such examples include Zachary et al. (JSC, 15, 263, 1994), Ryu et al. (ApJ, 452, 785, 1995 and ApJ, 509, 244, 1998), Dai & Woodward (ApJ, 494, 317, 1998), Jiang & Wu (JCP, 150, 561, 1999), and Londrillo & Del Zanna (ApJ, 530, 508, 2000). The problem was also studied as a model for 2-D turbulence by Dahlburg & Picone, Phys. Fluid B, 1, 2153 (1989) and Picone & Dahlburg, Phys. Fluid B, 3, 29 (1991), using Fourier spectral methods.

