# Numerical Modelling of MHD Space Plasmas

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# Collaborators

#### **University of Waterloo**

Lucian Ivan (postdoc)

#### **University of Toronto**

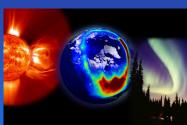
- Prof. Clinton Groth
- Scott Northrup
- CFD & Propulsion Group, Institute for Aerospace Studies

## **CSA Canadian Geospace Monitoring (CGSM) Program**

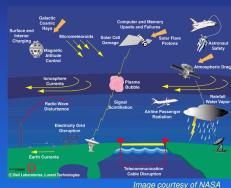
Project: "Solar Drivers of Space Weather: Contributions to Forecasting"



**Goal**: Develop advanced simulation methods for MHD space plasmas and apply to space-weather forecasting **Housed At**: Applied Math., U. Waterloo



Images courtesy of SOHO/EIT consortium



# MHD Equations

#### **Ideal MHD Plasma**

- magnetized, inviscid, fully ionized, compressible gas
- quasi-neutral, isotropic pressure, perfect gas (i.e.  $p = \rho RT$ )

## Flow Governed by 3D Compressible MHD Equations

$$\frac{\partial \mathbf{U}}{\partial t} + \vec{\nabla} \cdot \vec{\mathbf{F}} = \mathbf{S}$$

Conserved solution state:  $\mathbf{U} = \begin{bmatrix} \rho, & \rho \vec{u}, & \vec{B}, & \rho e \end{bmatrix}^{\mathrm{T}}$ 

Flux dyad: 
$$\vec{\mathbf{F}} = \begin{bmatrix} \rho \vec{u}, & \rho \vec{u} \vec{u} + (p + \vec{B} \cdot \vec{B}/2) \vec{\hat{I}} - \vec{B} \vec{B}, \\ \vec{u} \vec{B} - \vec{B} \vec{u}, & (\rho e + p + \vec{B} \cdot \vec{B}/2) \vec{u} - (\vec{u} \cdot \vec{B}) \vec{B} \end{bmatrix}^{\mathrm{T}}$$

Spherically symmetric gravitational field:

$$\mathbf{S} = -\frac{\rho G M_*}{r^3} \begin{bmatrix} 0, \vec{r}, 0, \vec{r} \cdot \vec{u} \end{bmatrix}^{\mathrm{T}}$$

# **CSA Canadian Geospace Monitoring (CGSM) Program**

Project: "Solar Drivers of Space Weather: Contributions to Forecasting"

#### In Collaboration with:

- Prof. Groth's Group (CFFC Computational Framework)
- NRCan Geomagnetic Laboratory (Modeling & Validation)
- Others (e.g. Space Weather Forecasting 'in the cloud')

# **CSA Canadian Geospace Monitoring (CGSM) Program**

Project: "Solar Drivers of Space Weather: Contributions to Forecasting"

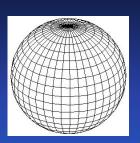
#### **Computational Framework Features/Design Goals**

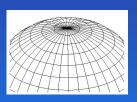
- CFFC: Computational Framework for Fluids and Combustion
- Hybrid unstructured-structured multi-block mesh our approach: "cubed sphere"
- Dynamic adaptive mesh refinement (AMR)
- Parallel and highly scalable implementation
- Implicit timestepping-multigrid/multilevel acceleration
- 2nd-, 3rd- and 4th-order accuracy
- Hardware accelerators: GPU computing

#### **Several Options in the Literature**

- Latitude-longitude grid constructs
- Cubed sphere
- Cartesian cut-cell approach
- Geodesic grid (e.g. icosahedron)

#### **Latitude-Longitude Grid Constructs**





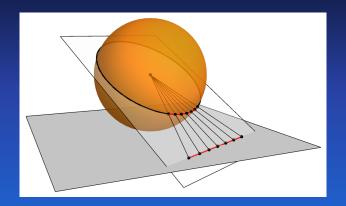
#### **Advantages**

- Natural basis for spherical flows
- Logically rectangular grid  $(r, \theta, \phi)$
- Suitable for application of spectral methods
- Fairly uniform away from poles

#### **Issues** ("pole problems")

- Singularities at the poles, non-uniform
- Severe time-step restrictions
- Parallelization difficulties

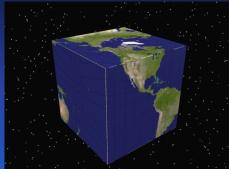
#### **Gnomonic Projection Based Grids**



Great circles are mapped into straight lines and vice-versa

#### "Cubed Sphere"





- The inverse projection maps the 6 straight faces of the cube into 6 adjoining spherical faces free of any strong singularities
- Natural application of a multi-block mesh data structure

# Types of Cubed-Sphere Grids

Sadourny, 1972; Ronchi et. al., 1996

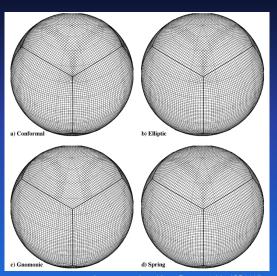
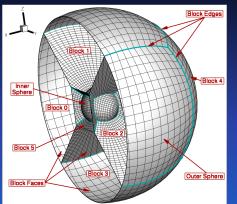


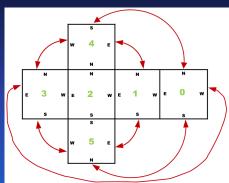
Image reproduced from Putman & Lin JCP2007

#### **Gnomonic Grid**

- The most uniform
- Equiangular or equidistant projection
- Min. length scales with increasing resolution
- Highly non-orthogonal and non-conformal
- Does not require extra meshing algorithms
- Accurate results with adequate schemes (Putman & Lin, 2007)
- Suitable for AMR

# 3D Cubed-Sphere Multi-Block Mesh in CFFC





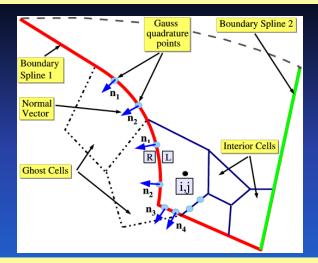
Cross-section of the cubed-sphere grid (**left**) and illustration of connectivity among blocks (**right**)

## Semi-Discrete Integral Form for Hexahedral Cell (i,j,k)

$$\frac{d\mathbf{U}_{i,j,k}}{dt} = -\frac{1}{V_{i,j,k}} \sum_{m=1}^{N_f} \left( \int \vec{\mathbf{F}} \cdot \vec{n} \, dA \right)_{i,j,k,m} + (\mathbf{S})_{i,j,k}$$

#### **Key Elements of Numerical Scheme**

- High-order spatial discretization (2nd, 3rd, 4th)
- Multi-dimensional k-exact least-squares reconstruction (Barth, 1993) and limiters (Barth-Jespersen, 1993; Venkatakrishnan, 1993)
- Upwinding flux evaluation (Roe's 1981; HLLE 1983)
- Multi-stage explicit time marching schemes
- Parallel implicit NKS algorithms (Northrup & Groth 2009)



$$\frac{d\mathbf{U}_{i,j,k}}{dt} = -\frac{1}{V_{i,j,k}} \sum_{m=1}^{N_f} \left( \int \vec{\mathbf{F}} \cdot \vec{n} \ dA \right)_{i,j,k,m} + (\mathbf{S})_{i,j,k}$$

Overview of High-Order Benefits

### What is Meant by High-Order?

- Schemes with order of truncation error greater than 2
- Spatial discretization error:  $\mathcal{O}(\Delta x^n)$ , n > 2

#### **High-Order Schemes vs. Low-Order Methods**

- Less numerical dissipation & dispersion
- Require fewer mesh points to accurately resolve solution
- Potentially less expensive
- Trade-offs: Order of accuracy vs. computational cost
- Our target: 4th-order accurate solutions (n = 4)

# Work in Concert with Other Strategies for Coping with High Computational Cost

- Adaptive Mesh Refinement (AMR)
- Efficient parallel solution schemes

#### k-exact Reconstruction (Barth, 1993)

Piecewise polynomial approximation

$$u_{i,j}^k(\vec{r}) = \sum_{p_1=0}^{N_1} \sum_{p_2=0}^{N_2} (x - \bar{x}_{i,j})^{p_1} (y - \bar{y}_{i,j})^{p_2} D_{p_1 p_2}^k , \quad N1 + N2 \le k$$

- Satisfies the following conditions:
  - reconstruct exactly polynomials of degree  $\leq k$

$$u_{i,j}^k(\vec{r} - \vec{r}_{i,j}) - u(\vec{r}) = \mathcal{O}(\Delta x^{k+1})$$

conserve the mean solution

$$\iint_{\mathcal{A}_{i,j}} u_{i,j}^k(\vec{r} - \vec{r}_{i,j}) \, dxdy = \iint_{\mathcal{A}_{i,j}} u(\vec{r}) \, dxdy$$

have compact support

# Computation of Reconstruction Coefficients, $D_{p_1p_2}^k$

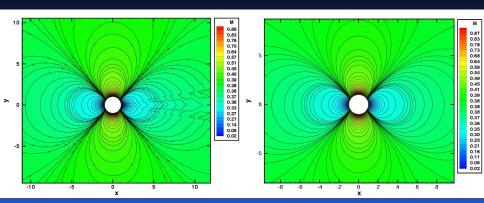
• Error,  $Err_{\gamma,\delta}$ , in each control volume,  $CV_{\gamma,\delta}$ , is minimized using least-squares formulation:

$$Err_{\gamma,\delta} = \frac{1}{A_{\gamma,\delta}} \iint_{\mathcal{A}_{\gamma,\delta}} u_{i,j}^k (\vec{r} - \vec{\bar{r}}_{i,j}) \, dx dy - \bar{u}_{\gamma,\delta}$$

 Results in linear equality-constrained least squares problem:

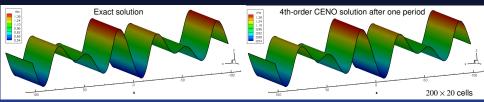
$$Ax - b = r$$

- Solution of overdetermined linear system of equations:
  - Gaussian elimination + Householder QR factorization (or: compute the pseudo-inverse A<sup>-1</sup>)
- remove oscillations at discontinuities: detect discontinuities and reduce order (limited piecewise linear) (CENO scheme)



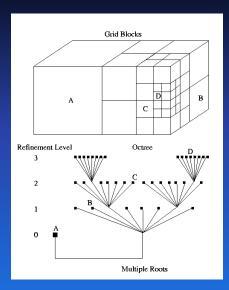
The Mach number prediction for the inviscid flow past a cylinder at M=0.38 obtained with the 2nd- and 4th-order CENO on a mesh with 80x40 cells

2nd vs. 4th Order Accuracy: 2D Example to Illustrate Potential Benefits



# Cells	$\mathcal{O}(\Delta x^2)$		$\mathcal{O}(\Delta x^4)$	
4,000 (200x20)	L1: L2: LMax: Time: Mem:	2.68E-02 3.26E-02 7.36E-02 0:01:20 20,336	L1: L2: LMax: Time: Mem:	1.85E-04 2.18E-04 9.46E-04 <b>0:10:38</b> <b>31,232</b>
8,000 (400x20)	L1: L2: LMax: Time: Mem:	9.38E-03 1.16E-02 2.98E-02 0:04:18 30,000	L1: L2: LMax: Time: Mem:	1.32E-05 2.02E-05 2.11E-04 0:41:03 50,816
80,000 (4000x20)	L1: L2: LMax: Time: Mem:	1.10E-04 2.20E-04 1.33E-03 8:23:38 203,680	L1: L2: LMax: Time: Mem:	- - - -

Block-Based AMR Using Hierarchical Data Structure (Berger 1984; Gao & Groth 2010)



- Mesh refinement by division and coarsening of self-similar structured blocks (hexahedral cells)
- Hierarchical octree data structure provides block connectivity
- Solution transfer among blocks via overlapping ghost cells
- Permits local refinement of mesh
- Physics-based refinement criteria (e.g.  $\epsilon_1 \propto |\vec{\nabla}\rho|$ ,  $\epsilon_2 \propto |\vec{\nabla}\cdot\vec{u}|$ ,  $\epsilon_3 \propto |\vec{\nabla}\otimes\vec{u}|$ )
- Permits parallel implementation via domain partitioning
- Highly efficient load balancing is obtained by equally distributing the solution blocks among CPUs

AMR of Cubed-Sphere Grid

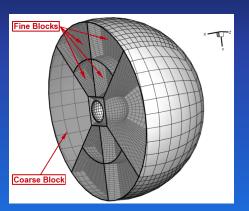
#### **Previous Work**

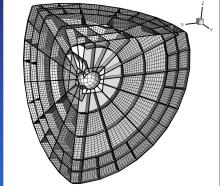
- Multiple radial cuts and stretching (not AMR, many research groups)
- AMR on Cartesian grids

AMR of Cubed-Sphere Grid

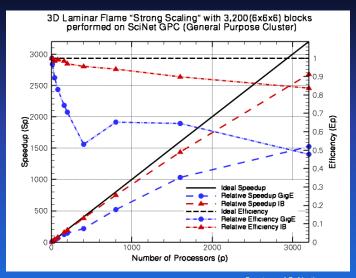
## **CFFC Implementation**

- Truly 3D AMR, using unstructured root-block connectivity
- Body-fitted mesh by constraining the points on the boundary spheres





Assessment of CFFC Parallel Performance on SciNet GPC (Nehalem processors)

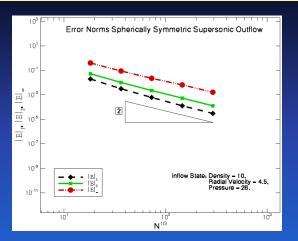


Courtesy of S. Northrup

#### **Summary of Studied Problems**

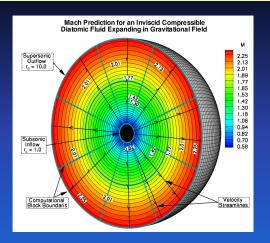
- Supersonic Outflow
- Transonic Wind on Fixed and AMR Meshes (Radially Symmetric)
- Supersonic Flow Past a Sphere
- Supersonic Rotating Outflow

Supersonic Inflow Outflow (Hydro, 2nd Order)  $R_i = 1$  (Inflow),  $R_o = 4$  (Outflow),  $GM_* = 0$ 



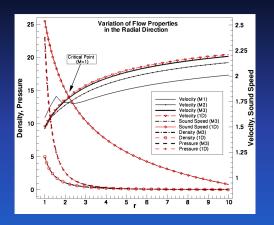
Convergence study based on analytical solution for meshes in the range 6,144 to 25,165,824 total cells

Transonic Wind (Hydro, 2nd Order)  $R_i = 1$ ,  $R_o = 10$ ,  $GM_* = 14$ , Inflow:  $\rho = 5$ , p = 23



Predicted Mach number distribution obtained on a uniform mesh with 1,228,800 total cells and 128 cells in the radial direction

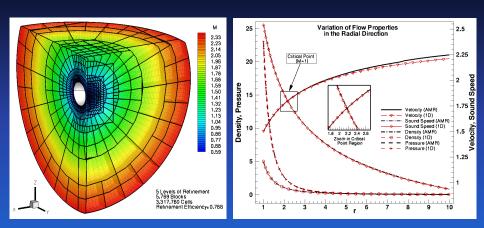
Transonic Wind (Hydro, 2nd Order)  $R_i = 1$ ,  $R_o = 10$ ,  $GM_* = 14$ , Inflow:  $\rho = 5$ , p = 23



Comparison of flow properties along X-axis for M1 (19,200), M2 (153,600) and M3 (1,228,800) meshes relative to a 1D "exact solution" obtained with Newton Critical Point (NCP) method (De Sterck *et. al.* 2009).

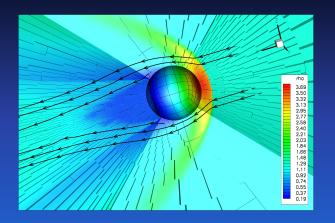
Transonic Wind on AMR Mesh

 $R_i = 1$ ,  $R_o = 10$ ,  $GM_* = 14$ , Inflow:  $\rho = 5$ , p = 23



Predicted Mach number distribution obtained on the adapted cubed-sphere mesh. Comparison of flow properties in the X-axis direction.

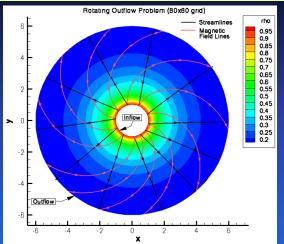
Supersonic Flow Past a Sphere (Hydro, 2nd Order)  $M_{\infty} = 2.0, R_i = 1, R_o = 32, GM_* = 0$ 



Predicted density distribution on the final refined AMR mesh with 10,835 blocks and 8,321,280 computational cells (7 levels of refinement,  $\eta = 0.993$ )

2D Rotating Outflow (MHD, 2nd Order)

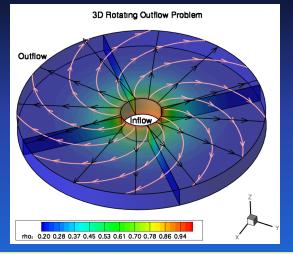
$$R_i = 1, R_o = 6, GM_* = 0, Inflow: \rho = 1, p = 1, V_r = 3, V_\theta = 1, B_r = 1$$



Predicted density field distribution obtained on a mesh with  $80 \times 80$  cells

3D Rotating Outflow (MHD, 2nd Order)

$$R_i = 1, R_o = 6, GM_* = 0, Inflow: \rho = 1, p = 1, V_r = 3, V_\theta = 1, B_r = 1$$



Density distribution obtained on a mesh with 20,160 total cells ( $70 \times 72$  radial)

# Concluding Remarks & Ongoing Research

#### **Parallel Block-Based Adaptive Simulation Framework**

- Developed for 3D cubed-sphere grids and space-physics flows
- Uses multi-dimensional FVM and gnomonic cubed-sphere grids
- Accuracy assessment based on exact and accurate 1D solutions
- Permits local solution-directed mesh refinement
- Handles and resolves regions of strong discontinuities/shocks

# Concluding Remarks & Ongoing Research

#### **Ongoing Research**

- High-order on cubed-sphere grid
- $\nabla \cdot \vec{B} = 0$  with high-order
- Improved implicit time integration
- Application of the method to space-physics problems (solar wind, CME)

# Acknowledgements

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- Computations were performed on the GPC supercomputer at the SciNet HPC Consortium