

Numerical Methods for Astrophysical Magnetohydrodynamics

Jim Stone

July 19, 2016

1 Lecture 1 — Astrophysical MHD

I will mainly talk about one single class of numerical method for MHD, which is finite volume method. [Link to the slides.](#)

1.1 Overview

Why are we interested in doing numerical MHD? Most of the big questions in astrophysics require the study of the dynamics of visible matter in the form of plasma. For example, how do galaxies form, how do stars form, how do planets form, etc. We are going to focus on collisionless plasma today, and for collisional plasma additional methods are required.

The equations of inviscid ideal MHD are as follows (in units so that $\mu = 1$)

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{v}) = 0 \quad (1.1)$$

$$\frac{\partial(\rho \mathbf{v})}{\partial t} + \nabla \cdot (\rho \mathbf{v} \mathbf{v} - \mathbf{B} \mathbf{B} + P^*) = 0 \quad (1.2)$$

$$\frac{\partial E}{\partial t} + \nabla \cdot [(E + P^*) \mathbf{v} - \mathbf{B} (\mathbf{B} \cdot \mathbf{v})] = 0 \quad (1.3)$$

$$\frac{\partial \mathbf{B}}{\partial t} - \nabla \times (\mathbf{v} \times \mathbf{B}) = 0 \quad (1.4)$$

where $P^* = P + B^2/2$ and $E = \rho v^2/2 + e + B^2/2$ are the total pressure and total energy. The above equation is not closed, since one needs an equation of state $P = P(\rho, T)$.

The hydro equations can be written in a compact form (in 1D)

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial \mathbf{F}}{\partial x} = 0, \quad (1.5)$$

This is a hyperbolic system of equations. They admit wavelike solutions, in the form of

$$a = a_0 + a_1 \exp(i\mathbf{k} \cdot \mathbf{x} - i\omega t) \quad (1.6)$$

When $a_1 \ll a_0$ waves have small amplitude, and they are in the linear regime. When $a_1 \gg a_0$ we are in nonlinear regime.

The dispersion relation for MHD waves can be found and summarized together:

$$[\omega^2 - (\mathbf{k} \cdot \mathbf{v}_A)^2] [\omega^4 - \omega^2 k^2 (v_A^2 + C^2) + k^2 C^2 (\mathbf{k} \cdot \mathbf{v}_A)^2] = 0 \quad (1.7)$$

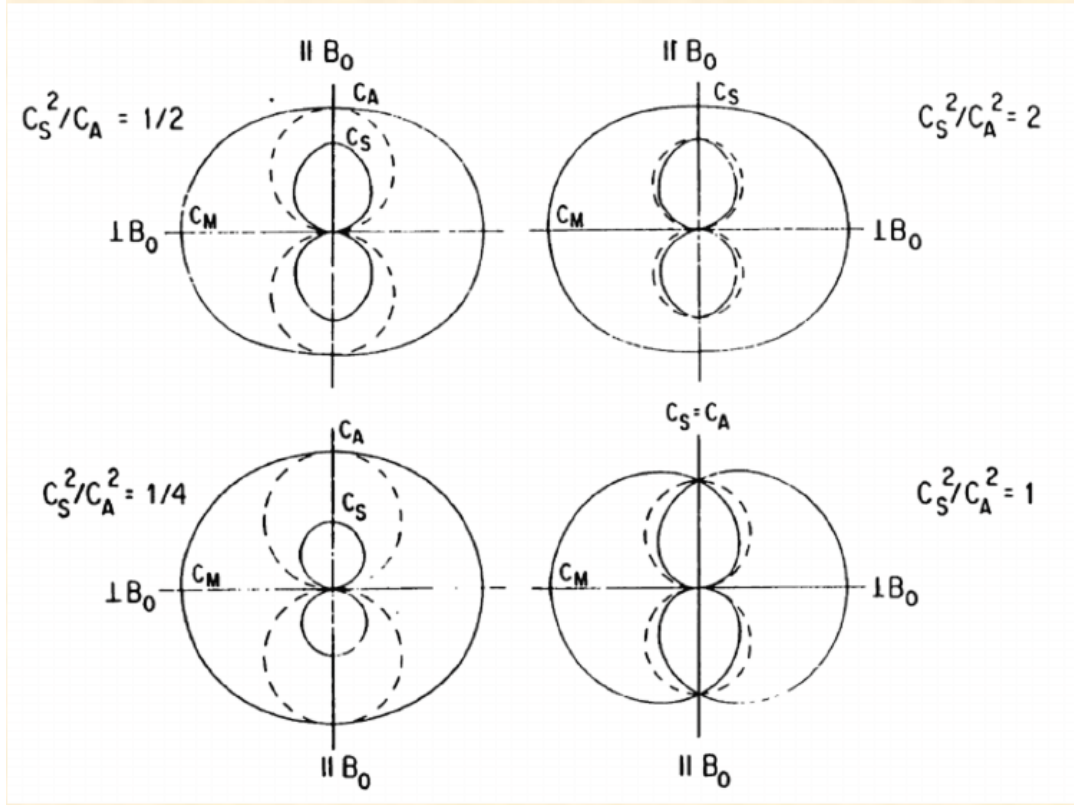


Figure 1.1: Friedrichs diagrams (from the slides)

The phase velocities of MHD waves can be summarized by Friedrichs diagrams.

However, the point of numerical simulations is to solve the full nonlinear problem which can't be solved analytically. The simplest example is a contact discontinuity: discontinuous change in density which constant P advected at constant \mathbf{v} . We can also have shocks where all variables can be changed. In real life for example, when a plane flies across air at speed $v > C$ it creates a shockwave.

These kind of shocks are described by jump conditions, which are changes in conserved variables across the discontinuity. To describe the hydrodynamic shock, in the frame of the shock, there is a steady fluid flow toward the shock from the upstream, and away from the show in the downstream direction. Mass, energy and momentum are conserved in this frame. We can solve the equations for the Rankine-Hugoniot jump conditions

$$\frac{\rho_d}{\rho_u} = \frac{(\gamma + 1)\mathcal{M}^2}{(\gamma - 1)\mathcal{M}^2 + 2} \quad (1.8)$$

where $M = v_u/C_u$ is the shock Mach number.

In MHD the shocks are more complicated since there are more kinds of shocks. There can be Alfvén “shocks” where only the transverse components of v and B change discontinuously. There can be slow or fast shock, switch-on/off shocks. These are shocks with small parameter space but very useful to test codes to see if these can be captured.

Going beyond shocks is a zoo of linear instabilities of MHD fluids. Some of the most important instabilities are gravitational instability, thermal instability, Rayleigh-Taylor instability (light fluid supporting

a heavy fluid under acceleration), Richtmyer-Meshkov instability, Kelvin-Helmholtz instability (two fluids flowing shearing across each other), and magneto-rotational instability (in accretion disks for example). All these instabilities are important in astrophysical conditions so we need to capture these correctly. Instabilities often lead to turbulence, and the study of nonlinear turbulence is an important application of the codes.

What do we mean by correctly? We need to understand what is feasible and what is not. The dispersion relation for most MHD instabilities are unstable for a broad range of wave numbers. However on a grid we can only have a finite resolution depending on the grid size. Correctly means the linear growth rate of all modes between $N\Delta x$ and L are represented accurately. If we are studying unresolved flows that is dominated by truncation error then we will get different answer with different resolutions.

1.2 Numerical Methods

Lets start off by round-off error. Not all floating numbers can be represented due to finite bit length. Rounding is correct if no machine number lies between x and its rounded value x' , and the difference between them is the round-off error.

We can prove that the relative error of a rounded value is always bounded by a small, machine dependent number

$$\frac{|x - x'|}{|x|} < \epsilon \quad (1.9)$$

which is the basis of all rigorous error analysis of numerical methods.

Another error is the truncation error. Numerical algorithms approximate analytic solutions with algebraic operations. The difference between true and approximate (numerical) solutions is the truncation error. This is under programmer control, as oppose to the round-off error.

There are three concepts in numerical analysis that are very important: Convergence, Consistency, and Stability. Convergence means when Δx and Δt decreases, the truncation error should also decrease. Higher order schemes often converge faster, but they cost more which puts a practical limit on the usefulness of high order schemes. All methods are first-order for discontinuities, so high order schemes are not very useful for shock capturing.

Consistency is another important concept which is often forgotten. It means that the solutions of the underlying PDEs should not only converge but also converge to the correct analytic solution. Finally stability is also important and it means that round-off error must remain small and bounded. A solution can be numerically unstable seeded by uncontrolled round-off error and it can grow exponentially, which is undesirable.

The simplest discretization of a simple hyperbolic PDE is the forward-time centered-space (FTCS) finite differencing method.

$$\frac{\partial u}{\partial t} + a \frac{\partial u}{\partial x} = 0, \implies \frac{u_j^{n+1} - u_j^n}{\Delta t} + a \left(\frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x} \right) = 0 \quad (1.10)$$

We can perform a von Neumann stability analysis to show that this method is unconditionally unstable. To show this we can insert

$$u_j^n = \xi^n \exp(ikj\Delta x) \quad (1.11)$$

When we substitute this into the finite difference equation we get $\xi(k) = 1 - i(a\Delta t/\Delta x) \sin k\Delta x$. The amplitude of $\xi(k) > 0$ for all $\Delta t > 0$. Therefore the method will blow up for any evolution in time.

However this is easy to fix by changing the time derivative to use an average of u at t^n (Lax-Friedrichs):

$$\frac{u_j^{n+1} + (u_{j+1}^n + u_{j-1}^n)/2}{\Delta t} + a \left(\frac{u_{j+1}^n - u_{j-1}^n}{2\Delta x} \right) = 0 \quad (1.12)$$

Now if we do the same analysis then we see $\xi(k) < 1$ iff

$$\frac{a\Delta t}{\Delta x} \leq 1 \quad (1.13)$$

which is known as the Courant-Levy-Friedrichs (CFL) stability criterion.

Why does this work? In fact this LF equation introduces explicit numerical viscosity which makes the algorithm stable, since we can rewrite the finite difference equation to be

However the LF method has huge diffusion, so not recommended for practical use nowadays. Another way is to use the Upwind methods

$$\begin{aligned} \frac{u_j^{n+1} - u_j^n}{\Delta t} = & -a \left(\frac{u_j^n - u_{j-1}^n}{\Delta x} \right), \quad \text{if } a > 0 \\ & -a \left(\frac{u_{j+1}^n - u_j^n}{\Delta x} \right), \quad \text{if } a < 0 \end{aligned} \quad (1.14)$$

This method has less diffusion than LF.

Both the above methods are diffusion methods. They add diffusion error to the result, such that a localized perturbation will tend to diffuse away. Some other methods (e.g. LW) will add dispersion error (with different dispersion relation). Diffusion errors are acceptable, but dispersion errors will be disastrous in some situations and should be avoided at all cost.

1.3 Finite Volume Methods

There is a zoo of algorithms for MHD. To list some of them: Finite-differencing with hyper-viscosity, operator split methods, finite-volume methods, spectral methods, discontinuous Galerkin methods, MHD SPH methods.

Finite volume MHD codes are popular and there are a variety of them: Athena, VAC, AstroBEAR, PLUTO, FLASH, HARM, Enzo, Cosmo++, etc. This lecture will focus on the methods in Athena.

1.3.1 Discretization

How does finite volume discretization work? We first discretize the space $\mathbf{x} \rightarrow (x_i, y_i, z_i)$. We discretize time into $t \rightarrow t^n$. We discretize continuous variables to be volume average values for each cell. This is an important facet of the scheme, since it is not a sampling, but an averaging.

We use a staggered mesh where scalars are cell-centered and magnetic fields are defined at cell faces. Cell-centered quantities are volume-averaged and face centered quantities are area-averaged. Since we mostly compute the curl of B , area averaging is the natural discretization of the magnetic field.

We write the conservation laws in the form of hyperbolic equations

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial \mathbf{F}}{\partial x} + \frac{\partial \mathbf{G}}{\partial y} + \frac{\partial \mathbf{H}}{\partial z} = 0 \quad (1.15)$$

We can integrate the equation over the volume of a grid cell, and over a timestep dt , we can get an exact finite difference equation, where the differences are differences over average values.

For the induction equation we use the finite area discretization, which is a similar concept, but averaging over an area instead of a volume, which is natural for the equation. Again we obtain an exact equation.

To summarize, mass, momentum, and energy fields are cell-centered. Fluxes are face centered, and EMFs are edge-centered. The key is now to compute these fluxes and EMFs all at once!

1.3.2 Godunov Method

The answer is to use Godunov’s method. Difference in cell-averaged values at each grid interface define a set of Riemann problems (evolution of initially discontinuous states). The solution of Riemann problems averaged over cell gives time evolution of cell-averaged values, until the wave hits the grid interfaces. Due to conservation we don’t even need to solve the Riemann problem exactly. We only need to compute the flux through the interfaces between the cells, so we only need to solve the Riemann problem at every cell interface.

So we introduce a Riemann solver. There are many possible solvers. In MHD nonlinear Riemann solvers are complex because of the many families of waves and many characteristics. In addition, the equations of MHD are not strictly hyperbolic. Therefore MHD Godunov schemes use approximate or linear Riemann solvers.

There is Roe’s method which keeps all 7 characteristics but treat each as a simple wave. There is HLLE method which keeps only the largest and smallest characteristics, but averages intermediate states in-between. There is also the HLLC method which adds entropy and Alfvén waves back into the HLLE method giving 2(4) intermediate states.

To determine which Riemann solver is the best, one needs to explore the use of each. However the use of a Riemann solver is a benefit, not a weakness, since it makes shock capturing more accurate: we encode in the numerical scheme a inherent way of dealing with discontinuities.

We can use higher order methods to define Riemann problems to reconstruct left-right states within cells using piecewise linear/parabolic methods. If we need to do multidimensions, typically one split the equations directionally. One solve the equations in x directions first, then do it for y with results from the x update, and so on. However in MHD this kind of splitting will not preserve $\nabla \cdot \mathbf{B} = 0$. One needs to use directionally unsplit methods. One first computes first order fluxes at every interface, use these fluxes to advance solution for $\Delta t/2$, compute L/R states using this time-advanced state, and fluxes, and then finally advance solutions over a full timestep using the new fluxes.

1.3.3 Constraints and Tests

We also need to specify boundary conditions. Most codes apply boundary conditions using ghost/guard cells.

How do we keep $\nabla \cdot \mathbf{B} = 0$? There are many ways. One way is to do nothing and assume everything is okay. One way is to evolve B using vector potential \mathbf{A} , but this requires second derivatives which can be numerically dangerous. One way is to remove solenoidal part of \mathbf{B} using flux-cleaning, by setting $\mathbf{B} \rightarrow \mathbf{B} - \nabla \phi$ and solving an elliptic PDE at every time step. Finally the way used in Athena is to evolve the integral form of induction equation with Constrained Transport so that $\nabla \cdot \mathbf{B} = 0$ is automatically conserved.

However the last CT method requires E field at grid cell centers while they are defined on edges. Arithmetic averaging destroys the scheme so we need an algorithm to reconstruct E field at the corner.

For stability we have to observe the CFL condition

$$\Delta t \leq \frac{\Delta t}{v + C} \quad (1.16)$$

We define the Courant number C to be the ratio between lhs and rhs. For stability we need $C < 1$, and for multidimensions we usually need $C < 1/N_{\text{dim}}$.

There are many ways to test correctness. One simple way is to test linear wave convergence in 3D. We measure the RMS error in \mathbf{u} after propagating a pure eigenmode for one wavelength. The error better decrease linearly with increasing resolution. Some other tests include nonlinear circularly polarized Alfvén waves, shocktubes, MHD instabilities, etc.

2 Lecture 2 — Numerical Radiation MHD

Numerical MHD is a relatively mature field, and there are quite a few public codes that run well. However radiation MHD is still a very active field with almost no consensus method at all. We will talk about what is Radiation MHD, and numerical methods that we use.

2.1 Radiation Hydrodynamics

MHD is essential in black hole accretion disks, because angular momentum transport is driven by MHD turbulence! However if we are talking about accretion onto luminous sources, we need to consider radiation as well inside radius (Shakura & Sunyaev 1973)

$$r/r_g < 170(L/L_{\text{Edd}})^{16/21}(M/M_{\odot})^{2/21} \quad (2.1)$$

Radiation pressure needs to be included in dynamic models. Radiation dominated disks are subject to various instabilities, e.g. viscous instability and thermal instability.

Foundations of Radiation Hydrodynamics is a good book. The challenges in this field include:

- Which equations do we solve? Transfer equation vs. its moments.
- Which frame do we solve the equations in? Co-moving frame vs. Eulerian frame vs. mixed
- We need a proper closure of moment equations
- In diffusion limit the equations are hyperbolic-parabolic, mathematically challenging
- Wide range of timescales
- Frequency dependence adds another dimension
- Non-LTE effects requires modeling level populations

Radiation hydro means different things to different people. In some case it just means a source term in the energy equation which is an integral over all frequencies and scattering angles, which is a heating/cooling term by radiation. This could be done in Godunov scheme via operator splitting:

- Update flux divergence terms ignoring source terms

- Update source terms

It is relatively easy to add a cooling term like $\rho^2 \Lambda(T)$. However if we just add this term, then the code is subject to thermal instability. The growth rate is largest at small scales therefore grid scale truncation errors can quickly grow to make too much small scale structure. If one adds heat conduction this can be suppressed. It is crucial that one adds heat conduction to get an accurate solution.

Another application domain is ionizing radiation transport, e.g. studying the growth of HII regions in ISM. Now in the energy equation we have the heating and cooling terms, and in the mass conservation equation we have ionization and recombination terms. To add these ionization terms we need to compute radiative transfer from every grid cell, which is a big challenge.

One way is to use the adaptive ray-tracing method of Abel & Wandelt (2002) and Whalen & Norman (2006). The challenge is parallelization.

Another class of problems in this domain is to consider momentum transfer instead of energy transfer. One now needs to include momentum exchange in the force equation. One application is line-driven winds (assuming gas is isothermal). However, the challenge is that computing this exchange term can be extremely difficult!

Finally one might need to do the most general problem with both energy and momentum transfer. One need to close the equations with radiation field energy and momentum equation. One could even replace photons with neutrinos and do neutrino transport instead of photon transport, like in the core-collapse SNe community. All of the above problems can be called “radiation hydrodynamics”, but the numerical methods required in each regime are very different. In the following part of the lecture we will focus on the problem with both energy and momentum transfer.

Fundamental description of the radiation field is the frequency dependent transfer equation

$$\frac{1}{c} \frac{\partial I_\nu}{\partial t} + \nabla \cdot (\mathbf{n} I_\nu) = j_\nu - \kappa_\nu I_\nu \quad (2.2)$$

One needs to decide to solve this equation or solving the moment of the equation. Even after choosing the equation we can choose our method to be either grid method or particle method. Grid methods are usually more accurate and less noisy, and particle methods (usually Monte Carlo) are usually very flexible and embarassingly parallelizable.

The noise in Monte Carlo methods is a real problem. It gives about 1% noise which could excite sound waves and other instabilities in the fluid. The Monte Carlo is also very expensive compared to grid-based methods. However MC is great for GR-RT which is used in EHT imaging of black holes, because it is not a radiation pressure dominated problem.

2.2 Numerical methods for radiation hydro

Lets look at the radiation moment equations

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{v}) = 0 \quad (2.3)$$

$$\frac{\partial(\rho \mathbf{v})}{\partial t} + \nabla \cdot (\rho \mathbf{v} \mathbf{c} + P + B^2/2 - \mathbf{B}\mathbf{B}) = -PS_M \quad (2.4)$$

$$\frac{\partial E}{\partial t} + \nabla \cdot [(E + P)\mathbf{v} + (B^2/2)\mathbf{v} - \mathbf{B}(\mathbf{B} \cdot \mathbf{v})] = -PCS_E \quad (2.5)$$

$$\frac{\partial \mathbf{B}}{\partial t} - \nabla \times (\mathbf{v} \times \mathbf{B}) = 0 \quad (2.6)$$

$$\frac{\partial E_r}{\partial t} + C \nabla \cdot \mathbf{F}_r = CS_E \quad (2.7)$$

$$\frac{\partial \mathbf{F}_r}{\partial t} + C \nabla \cdot P_r = CS_M \quad (2.8)$$

One problem we have is the closure of these equations, since in general we can't assume the distribution function of photons is Maxwellian. One thing we can do is to assume $\mathbf{P} = \mathbf{f}E$, where \mathbf{f} is called the Eddington tensor. The flux-limited diffusion method assumes that radiation flux is given by Ficks's Law

$$\mathbf{F} = \frac{c\lambda}{\chi} \nabla E \quad (2.9)$$

where $\lambda(E)$ is limited to prevent super-luminal transport in optically thin regions. The other choice is the so-called M1 closure which assume a specific tensor from first principles by going into a frame of the fluid where the flow is isotropic. The third choice is to compute the Eddington tensor numerically which is called the variable Eddington tensor (VET) approach. One solves the time-independent transfer equations using method of short characteristics along N_r rays per cell, and take quadrature to get the tensor.

The flux-limited diffusion method reduces our problem to two-temperature diffusion problem (Turner & Stone 2001). The main pro is that it is easy to solve, but it lost information about the direction of flux. It also does not have radiation inertia, therefore allowing superluminal wave speeds, and it has no radiation shear viscosity.

We need a flux limiter $\lambda(E)$ and the most popular one is

$$\lambda(R) = \frac{2 + R}{6 + 3R + R^2}, \quad R = |\nabla E|/E \quad (2.10)$$

which gives correct flux in both optically thin and optically thick limits.

Material radiation interaction and radiation transport terms have a very restrictive time step limit, so we need to use implicit scheme for the timestepping. The equations give sparse banded matrices which need to be solved every NR iteration. There are iterative methods like GMRES or ICCG. This problem is even worse in 3D. The challenge of doing flux-limited diffusion is solving linear algebra problems.

There are approaches to get around this problem by reducing the speed of light in the medium. Since the flow speed in the ISM is way smaller than c we can make the gap artificially smaller and solve the problem explicitly. However one can derive constraints on slowest speed allowed to give correct dynamics (Skinner & Ostriker 2013).

Lets now consider M1 closure. It is a much improved method over FLD (Gonzalez et. al. 2007). It keeps flux as a separate variable and uses local information to construct direction of flux. This fixes the

shadowing problem with FLD, but replaces with photons colliding and merging problem. Photons behave like a collective fluid rather than particles, because M1 assumes there is a frame where the flow is isotropic while in reality there might not be. M1 is usually more attractive when there is a small number of sources.

Finally there is the VET approach (Trujillo Bueno & Fabiani Bendicho 1995). One could use short characteristics method solving only along ray segments that cross a single zone, or one could use long characteristics where we solve along rays that cross entire grid for each cell. Long characteristics are usually not feasible, but short characteristics can have problems in treating point sources, e.g. having spikes away from point sources which breaks symmetry.

2.3 Radiation in Godunov methods

Lets now consider how to put this into Godunov method. Assuming we have picked a scheme to solve the closure problem, we still face the problem of solving the Godunov method with stiff source terms. One could use semi-implicit (Picard iteration) schemes to ensure stability.

When one consider long evolutions we might want to use implicit methods anyway, but it can be expensive and inaccurate. Instead we could split fully implicit solution of the radiation moment equations from modified Godunov method for MHD equation, and we write our own implicit solvers for the moment equations.

A test for the full code is the classic linear wave convergence test. One could write down the linear waves of this system (quite nontrivial) and we can test the convergence to analytic solution in 3D with respect to grid resolution.

One also need to do nonlinear tests since linear tests don't make use of all terms. One nonlinear test would be radiation shock tests. In radiation hydro one could have a precursor region which is heated up by the photons, and in the post-shock gas might cool (relaxation region) due to photons leaving. This structure depends on the Mach number. The particle mean free path and photon mean free path are two scales which can be very very different. Getting the shock structure is pretty difficult (semi)analytically.

The cost of 3D RMHD equations using explicit differencing scales as $N_x N_y N_z N_m N_n$ where N_m is number of angles, and N_n is the number of frequencies. However this admits very efficient parallelization. One could also use adaptive angles and frequencies could be very powerful.