Hydrodynamics Schemes: Spoilt for Choice

Josh Borrow, Matthieu Schaller, Richard Bower, SWIFT Team



Overview

- Introduction to Cosmological Simulations
- Choosing a hydrodynamics scheme
- Implementing a new hydrodynamics scheme
- Testing hydrodynamics

Cosmological Simulations

- Need to solve gravity and hydrodynamics (along with a sub-grid model)
- Speed prized over accuracy: bigger box-sizes, higher resolution for subgrid physics, etc.



Density from Particles

- Say I am given a distribution of particles of mass *m*. What is the density?
- Can use a kernel-weighted average. $\rho_i = \sum_i m_j W_{ij}$



Smoothed Particle Hydrodynamics

- General method used both in astrophysics and industry
- Represent the fluid as particles (Lagrangian) $L(q, \dot{q}) = \frac{1}{2} \sum_{i=1}^{N} m_i \dot{r}_i^2 - \sum_{i=1}^{N} m_i u_i,$



Smoothed Particle Hydrodynamics and Magnetohydrodynamics, Price, 2012

A Quick Derivation

• Add in a little 1st law...

$$\left. \frac{\partial u_i}{\partial q_i} \right|_A = -\frac{P_i}{m_i} \frac{\partial \Delta V_i}{\partial q_i}$$

• A sprinkle of constraint equation...

$$\phi_i(\mathbf{q}) = \kappa h_i^{n_d} \frac{1}{\Delta \tilde{V}} - N_{ngb} = 0$$

A general class of Lagrangian smoothed particle hydrodynamics methods and implications for fluid mixing problems, Hopkins, 2013

A General Equation of Motion

• Follow the equations through (Lagrange multipliers)

$$\frac{\mathrm{d}\mathbf{v}_i}{\mathrm{d}t} = -\sum_{j=1}^N x_i x_j \left[\frac{f_{ij} P_i}{y_i^2} \nabla_i W_{ij}(h_i) + \frac{f_{ji} P_j}{y_j^2} \nabla_i W_{ji}(h_j) \right]$$

• With a correction for non-constant smoothing lengths

$$f_{ij} \equiv 1 - \frac{\tilde{x}_j}{x_j} \left(\frac{h_i}{n_d \tilde{y}_i} \frac{\partial y_i}{\partial h_i} \right) \left(1 + \frac{h_i}{n_d \tilde{y}_i} \frac{\partial \tilde{y}_i}{\partial h_i} \right)^{-1}$$

Getting an "Actual Scheme"

• Now need to make a *choice* of volume element.

$$P_{\rm eos} = (\gamma - 1)\rho u$$

Density-Energy

$$\rho_i = \sum_j m_j W_{ij}$$
$$\Delta V = \frac{m}{\rho}$$

Pressure-Energy

$$\bar{P}_i = \sum_j m_j u_j (\gamma - 1) W_{ij}$$
$$\Delta V = \frac{(\gamma - 1) m u}{\bar{P}}$$

Getting an "Actual Scheme"

Density-Entropy (Gadget-2)

$$\frac{d\mathbf{v}_i}{dt} = -\sum_j m_j \left[\frac{f_i P_i}{\rho_i^2} \nabla_x W(\mathbf{x}_{ij}, h_i) + \frac{f_j P_j}{\rho_j^2} \nabla_x W(\mathbf{x}_{ij}, h_j) \right]$$

Pressure-Energy

$$\frac{\mathrm{d}\mathbf{v}_i}{\mathrm{d}t} = -\sum_{i} (\gamma - 1)^2 m_j u_j u_i \left[\frac{f_{ij}}{\bar{P}_i} \nabla_i W_{ij}(h_i) + \frac{f_{ji}}{\bar{P}_j} \nabla_i W_{ji}(h_j) \right]$$

What's the Difference? (Physics)



Thermal instabilities in cooling galactic coronae: fuelling star formation in galactic discs, Hobbs+, 2013

The Problem Visualised



Fundamental differences between SPH and grid methods, Agertz+, 2007

- Artificial surface tension, caused by the density (and hence pressure, from the EoS which is linear in density) being discontinuous
- We fix that by smoothing the pressure

Density-Entropy $\rho_i = \sum_j m_j W_{ij}$ $P_{eos} = (\gamma - 1)\rho u$

Pressure-Energy
$$\bar{P}_i = \sum_j m_j u_j (\gamma - 1) W_{ij}$$

One Level of Abstraction Down



Another Level Down (Truly Testable)



Another Level Down (Truly Testable)



Does this Converge?

- Can look at this run with many different numbers of particles
- Note that the L2 norm never converges, stays constant



Can we do Better?



Is Particle Number the Correct Metric?

- Probably not!
- Best to compare at a fixed run-time, especially for more accurate schemes that are significantly more expensive
- At the moment, use SPH!



Reproducible Testing

- Need a framework for testing hydrodynamical schemes on many problems, at many different resolutions
- Enter Swift coNvergence physics Python Execution System (SNIPES)
- Will allow us to run many tests all at once, with each individual scheme
- Answer the central question: given a fixed amount of computing time, what scheme should I use?