



The Abdus Salam
**International Centre
for Theoretical Physics**



Joint ICTP-IAEA College on Plasma Physics

Vlasov simulations of plasma turbulence

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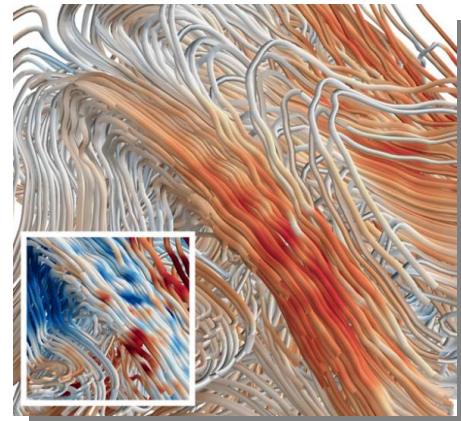
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UNIVERSITÀ DELLA CALABRIA

**DIPARTIMENTO DI
FISICA**



Part I

Numerical methods for kinetic simulations of plasmas

Part II

Direct numerical simulations of plasma turbulence: Eulerian simulations

Part III

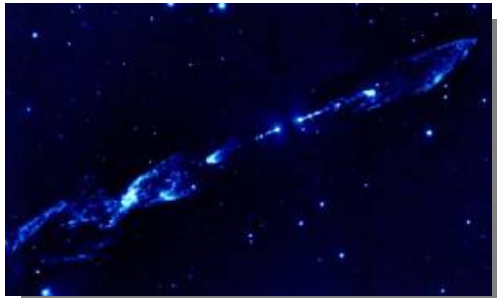
Comparison between simulations and observations;
Particle in cell simulations of turbulence

Part I

Numerical methods for kinetic simulations of plasmas

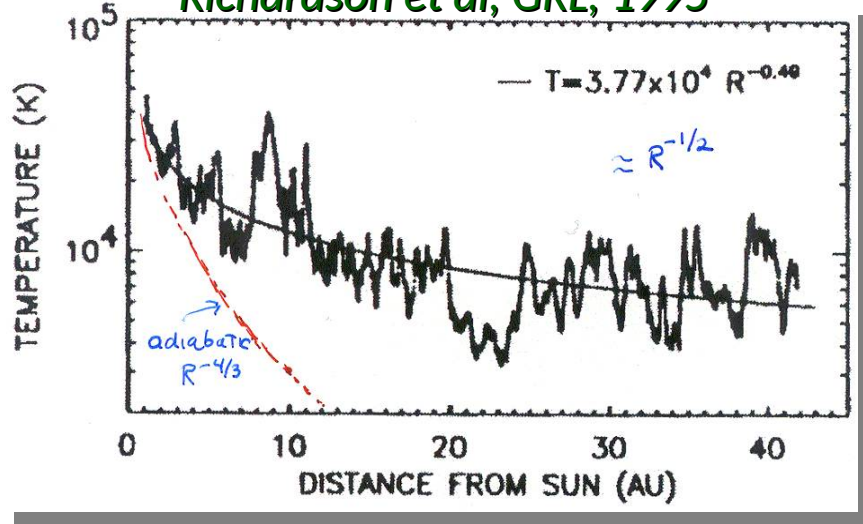
- Introduction to plasma complex dynamics
- Methods to simulate the Vlasov equation. Focus on the Eulerian approach
- Discretization of the equations, basic numerical schemes and numerical stability
- Advanced numerical schemes for simulations of plasma turbulence

- A plasma is a ionized gas where charged particles interact via electromagnetic forces
- More than 99.9 % of matter in the Universe can be considered as a plasma
- Plasma is mostly collisionless
- Observations are somehow limited

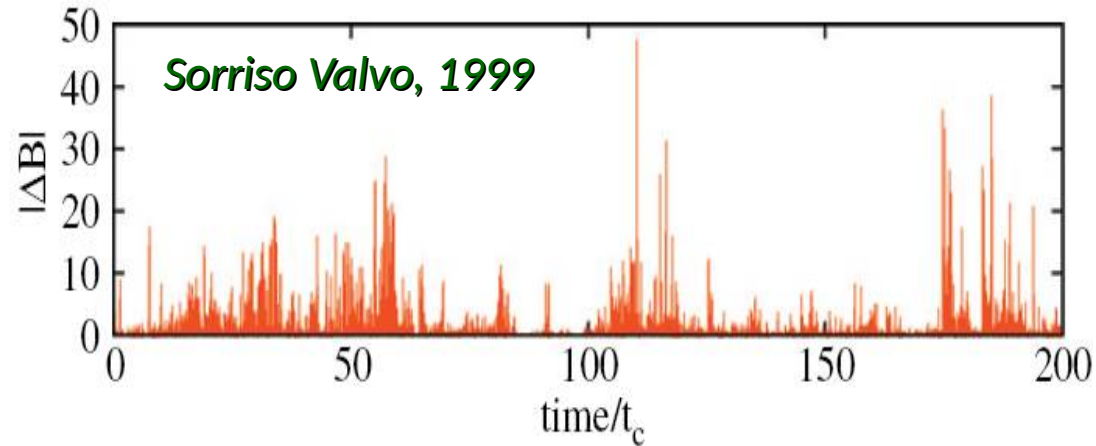


Turbulence in space plasmas

Richardson et al, GRL, 1995



Solar wind is turbulent and intermittent



Sorriso Valvo, 1999

Most of the plasma energization (plasma heating and particle acceleration) occurring in turbulent collisionless plasmas, such as those permeating the solar system is expected to occur at kinetic scales (scales \sim particle gyroradii and below)

How are plasmas heated and particle accelerated?

$$\Pi(\mathbf{x}, t) = m \int \int \int [\mathbf{v} - \mathbf{V}_{bulk}(\mathbf{x}, t)] [\mathbf{v} - \mathbf{V}_{bulk}(\mathbf{x}, t)] f(\mathbf{x}, \mathbf{v}, t) d^3v$$

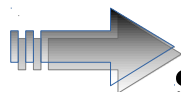
$$\Pi = p_{\perp} I + (p_{\parallel} - p_{\perp}) \mathbf{b} \mathbf{b}$$

$$\mathbf{b} = \frac{\mathbf{B}}{|\mathbf{B}|}$$

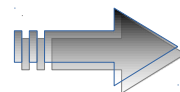
T_{\parallel} and $T_{\perp} \equiv$
parallel and perpendicular
proton temperatures with
respect to the ambient \mathbf{B}

Hellinger et al. GRL (2006);
Kasper et al. JGR (2006);
Kasper et al., (2002)

Kinetic instabilities
influence the solar wind

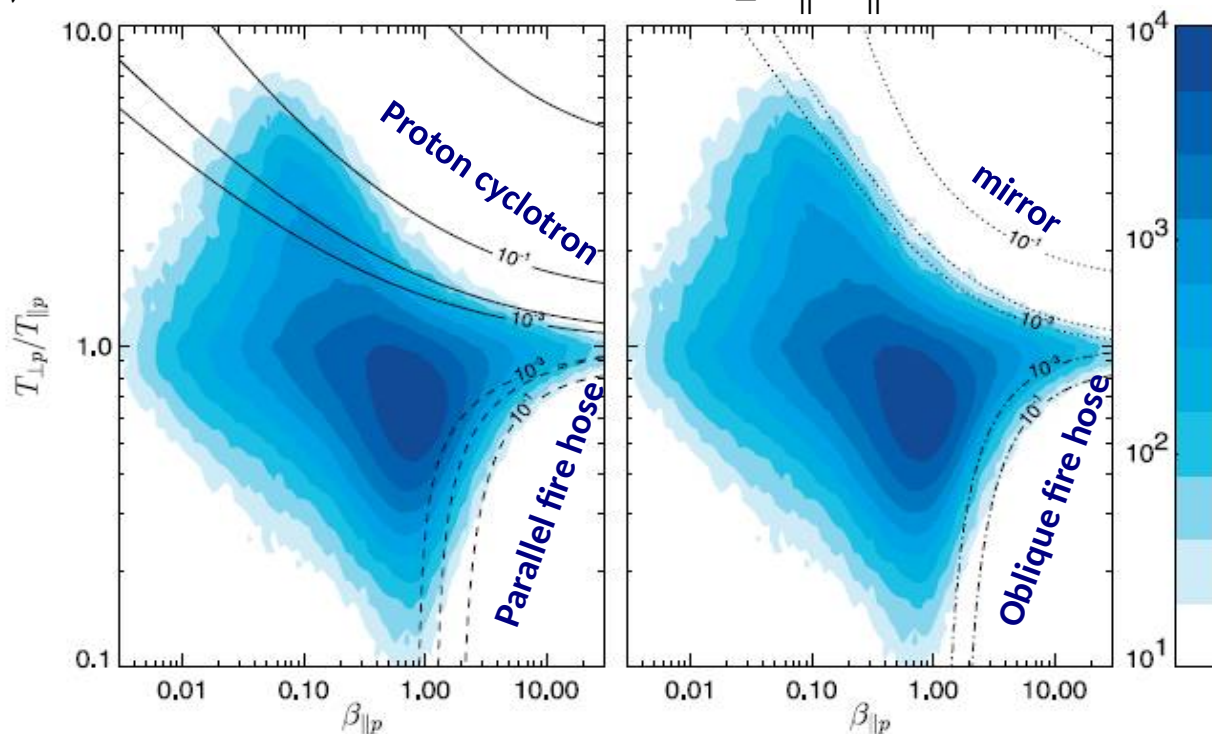


Astrophysical plasmas are
turbulent, intermittent and
show many kinetic (non-fluid)
phenomena



We need
simulations!

Distribution PDF(T_{\perp}/T_{\parallel} , β_{\parallel})



Simulations of Plasma Turbulence



Karimabadi et al, PoP, 2013

Kinetic equations for a collisionless plasma

The Vlasov-Maxwell system

$$\frac{\partial f_\alpha}{\partial t} + \mathbf{v} \cdot \nabla f_\alpha + \frac{q_\alpha}{m_\alpha} \left[\mathbf{E} + \frac{\mathbf{v} \times \mathbf{B}}{c} \right] \cdot \nabla_{\mathbf{v}} f_\alpha = 0 \longrightarrow f_\alpha = f_\alpha(\mathbf{x}, \mathbf{v}, t)$$

$$\nabla \cdot \mathbf{B} = 0 \qquad \int f_\alpha(\mathbf{x}, \mathbf{v}, t) d\mathbf{x} d\mathbf{v} = N_\alpha$$

$$\nabla \cdot \mathbf{E} = 4\pi \sum_{\alpha} q_\alpha \int f_\alpha(\mathbf{x}, \mathbf{v}, t) d\mathbf{v}$$

$$\nabla \times \mathbf{E} = -\frac{1}{c} \frac{\partial \mathbf{B}}{\partial t}$$

$$\nabla \times \mathbf{B} = \frac{4\pi}{c} \sum_{\alpha} q_\alpha \int \mathbf{v} f_\alpha(\mathbf{x}, \mathbf{v}, t) d\mathbf{v} + \frac{1}{c} \frac{\partial \mathbf{E}}{\partial t}$$

6D + time!

Two numerical philosophies

Method 1: the Eulerian approach

The Vlasov equation is solved directly for the particle distribution function, on a phase space grid. Moments (density and current) are evaluated by direct integration of the distribution function

- Zero noise
- ▶ Very computationally demanding because of memory limitations

Method 2: the Lagrangian approach

Vlasov is solved via a *Montecarlo* technique. The equations of motion of a large number of (macro) particles are solved and the distribution function is reconstructed. Maxwell equations are evaluated on a grid, through interpolation

- Very cheap from the computational point of view
- ▶ Numerical noise

We will use both methods, depending on the problem that we want to study. We will start with Method 1

Method 1: Eulerian

Full Vlasov-Maxwell

$$\frac{\partial f_\alpha}{\partial t} + \mathbf{v} \cdot \nabla f_\alpha + \frac{q_\alpha}{m_\alpha} \left[\mathbf{E} + \frac{\mathbf{v} \times \mathbf{B}}{c} \right] \cdot \nabla_{\mathbf{v}} f_\alpha = 0$$

$$\nabla \cdot \mathbf{B} = 0$$

$$\nabla \cdot \mathbf{E} = 4\pi \sum_{\alpha} q_{\alpha} \int f_{\alpha}(\mathbf{x}, \mathbf{v}, t) d\mathbf{v}$$

$$\nabla \times \mathbf{E} = -\frac{1}{c} \frac{\partial \mathbf{B}}{\partial t}$$

$$\nabla \times \mathbf{B} = \frac{4\pi}{c} \sum_{\alpha} q_{\alpha} \int \mathbf{v} f_{\alpha}(\mathbf{x}, \mathbf{v}, t) d\mathbf{v} + \frac{1}{c} \frac{\partial \mathbf{E}}{\partial t}$$

- Nonlinear integro-differential equation in 6D phase space + time
- Very hard and time demanding to solve numerically!
- To date, numerical solutions are available for approximated, reduced systems



Hybrid Vlasov-Maxwell

$$\frac{\partial f}{\partial t} + \mathbf{v} \cdot \nabla f + \frac{e}{m} \left[\mathbf{E} + \frac{\mathbf{v} \times \mathbf{B}}{c} \right] \cdot \nabla_{\mathbf{v}} f = 0$$

$$\mathbf{E} = -\frac{\mathbf{u} \times \mathbf{B}}{c} + \frac{1}{nec} (\mathbf{j} \times \mathbf{B}) - \frac{1}{ne} \nabla P_e$$

$$\nabla \times \mathbf{E} = -\frac{1}{c} \frac{\partial \mathbf{B}}{\partial t}$$

$$\nabla \times \mathbf{B} = \frac{4\pi \mathbf{j}}{c}$$

$$n_e \simeq n_i \simeq n$$

$$P_e = P_e(n)$$

easier



1D-1V Vlasov-Poisson

$$\frac{\partial f_e}{\partial t} + v \frac{\partial f_e}{\partial x} - \frac{eE}{m} \frac{\partial f_e}{\partial v} = 0$$

$$\frac{\partial E}{\partial x} = 4\pi e \left[n_0 - \int f_e(x, v, t) dv \right]$$

Vlasov equation is an advection equation in phase space

- Let us consider the 1D-1V case (we will discuss later the generalization to full phase space)

$$\frac{\partial f}{\partial t} + v \frac{\partial f}{\partial x} + a \frac{\partial f}{\partial v} = \frac{\partial f}{\partial t} + \frac{\partial(vf)}{\partial x} + \frac{\partial(af)}{\partial v} = 0$$

$$f = f(x, v, t); \quad a = a(x)$$

- Let us focus on advection in x first (later we will discuss how to couple it to advection in v)

$$\frac{\partial f}{\partial t} + v \frac{\partial f}{\partial x} = 0$$

$$f = f(x, t); \quad x \in [0, L]; \quad t \in [0, T]; \quad v = \text{const.}$$

$$f(x, 0) = f_0(x); \quad f(0, t) = f(L, t), \quad \forall t \in [0, T]$$

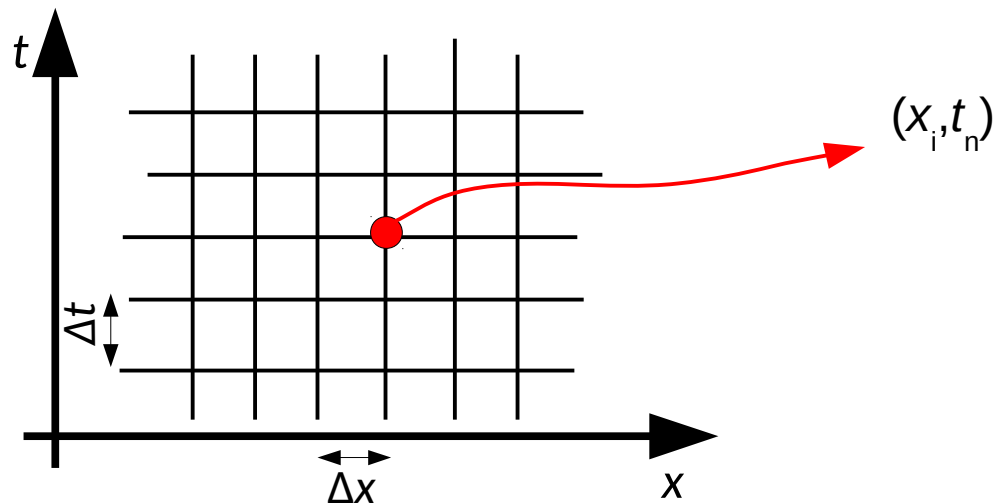


For simplicity, periodic boundary conditions

Three main steps:

- 1) Discretize (x,t) plane
- 2) Approximate derivatives in discretized plane (allowed operations are +, -, x, /)
- 3) Create algorithm to solve the equation

Spatial derivatives



Discretization:

$$x_i = (i - 1)\Delta x; \quad i = 1, \dots, N_x; \quad \Delta x = \frac{L}{N_x}$$

$$x_0 = 0, \quad x_{N_x} = L - \Delta x$$

$$t_n = n\Delta t; \quad n = 0, \dots, N_t; \quad \Delta t = \frac{T}{N_t}$$

$$t_0 = 0, \quad t_{N_t} = T$$

Derivatives approximation (finite differences):

$$f(x_{i+1}) = f(x_i) + \Delta x \left(\frac{df}{dx} \right)_{x_i} + \frac{1}{2} \Delta x^2 \left(\frac{d^2 f}{dx^2} \right)_{x_i} + \frac{1}{3!} \left(\frac{d^3 f}{dx^3} \right)_{x_i} + o(\Delta x^4)$$

$$f(x_{i-1}) = f(x_i) - \Delta x \left(\frac{df}{dx} \right)_{x_i} + \frac{1}{2} \Delta x^2 \left(\frac{d^2 f}{dx^2} \right)_{x_i} - \frac{1}{3!} \left(\frac{d^3 f}{dx^3} \right)_{x_i} + o(\Delta x^4)$$



$$(1) - (2) \Rightarrow \left(\frac{df}{dx} \right)_{x_i} = \frac{f(x_{i+1}) - f(x_{i-1}))}{2\Delta x} + o(\Delta x^2)$$

$$(1) + (2) \Rightarrow \left(\frac{d^2 f}{dx^2} \right)_{x_i} = \frac{f(x_{i+1}) + f(x_{i-1}) - 2f(x_i)}{\Delta x^2} + o(\Delta x^2)$$

(1)



$$\left. \frac{df}{dx} \right|_{x_i} = \frac{\text{Euler forward}}{f(x_{i+1}) - f(x_i)} \Delta x + o(\Delta x)$$

(2)



$$\left. \frac{df}{dx} \right|_{x_i} = \frac{\text{Euler backward}}{f(x_i) - f(x_{i-1}))} \Delta x + o(\Delta x)$$

Centered differences

Phase error for finite differences

Let's consider the centered difference scheme:

$$\left(\frac{df}{dx}\right)_{x_i} = \frac{f(x_{i+1}) - f(x_{i-1}))}{2\Delta x} + o(\Delta x^2)$$

Let's take for example:

$$f(x) = \sin(kx) \Rightarrow f'(x) = k \cos(kx) \Rightarrow f'(x_i) = k \cos(kx_i)$$

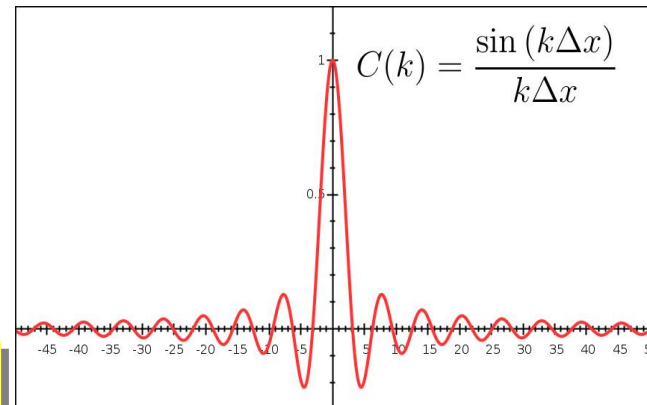
By applying centered-difference formula one gets:

$$f'(x_i)^{num} = \frac{\sin[k(x_i + \Delta x)] - \sin[k(x_i - \Delta x)]}{2\Delta x} = \cos(kx_i) \left[\frac{\sin(k\Delta x)}{k\Delta x} \right] = C(k)f'(x_i)$$

When solving advection equations through finite difference schemes:

$$\frac{\partial f}{\partial t} + v \boxed{C(k)} \frac{\partial f}{\partial x} = 0$$

Keep phase error under control!



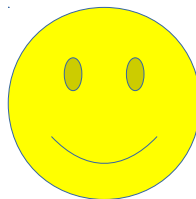
A naive try....

Explicit finite difference scheme

$$\begin{array}{cc} \text{Euler forward} & \text{Centered} \\ \frac{f_i^{n+1} - f_i^n}{\Delta t} = -v \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right) \end{array}$$



$$f_i^{n+1} = f_i^n - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right), \quad \forall i = 1, \dots, N_x$$

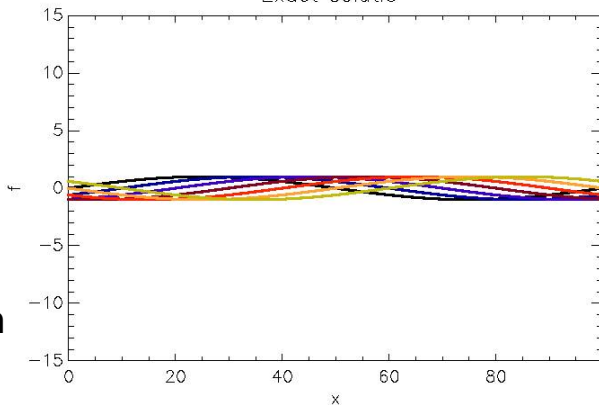


A naive try → numerical instabilities

Explicit finite difference scheme

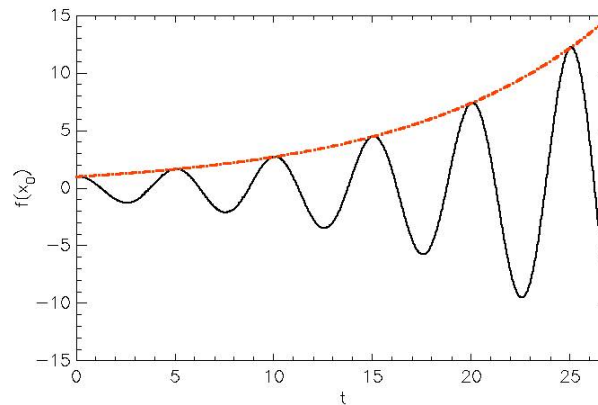
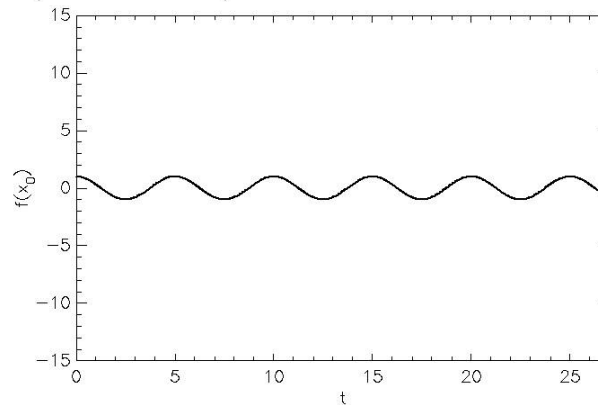
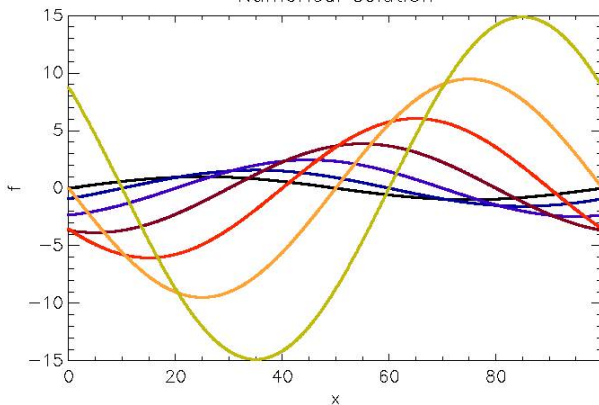
$$\frac{f_i^{n+1} - f_i^n}{\Delta t} = -v \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right)$$

Exact solution



Different times in sequence
(black, blue, ..., yellow) →
(t_0, t_1, \dots, t_N)

Numerical solution



Something weird is going on, numerical solution is out of control!



➤ Let us perform an analysis of the finite difference scheme by expressing the solution as a Fourier series

➤ Since the equation is linear, we only examine the behavior of a single mode. Consider a trial solution of the form

$$f^n(x_i) = \sum \hat{f}^n(k) e^{jkx_i} = \hat{f}^n(k) e^{ijk\Delta x} \quad \longrightarrow \quad f_i^n = \hat{f}^n e^{ijk\Delta x}$$

➤ This is a spatial Fourier expansion. Plugging in the difference formula:

$$\hat{f}^{n+1} = \hat{f}^n - \frac{C}{2} \hat{f}^n (e^{jik\Delta x} - e^{-jik\Delta x}); \quad C = \frac{v\Delta t}{\Delta x}$$

➤ Let us define the amplification Factor as:

$$A = \left| \frac{\hat{f}^{n+1}}{\hat{f}^n} \right|^2$$

A method is well-behaved or stable when $A \leq 1$

Von Neumann stability analysis (2)

For our “forward-central scheme” one gets

$$A = [1 - jC \sin(k\Delta x)]^2 = 1 + C^2 \sin^2(k\Delta x) \leq 1$$

Independently of the CFL number, all Fourier modes increase in magnitude as time advances

This method is unconditionally unstable...

We are in trouble!

Let us play a bit with our scheme

$$f_i^{n+1} = \underbrace{f_i^n}_{\text{Replace by average}} - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right), \quad \forall i = 1, \dots, N_x$$

Replace by average

$$f_i^{n+1} = \frac{(f_{i+1}^n + f_{i-1}^n)}{2} - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right)$$

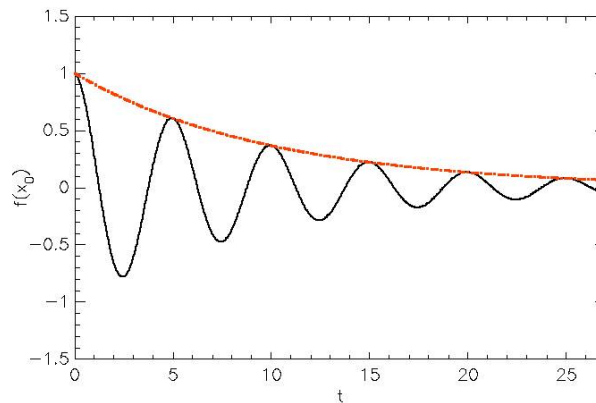
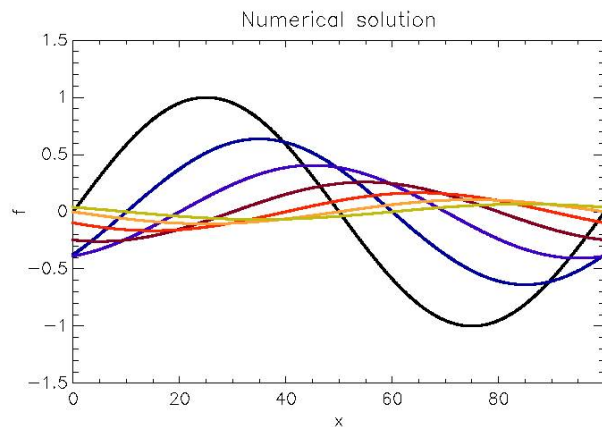
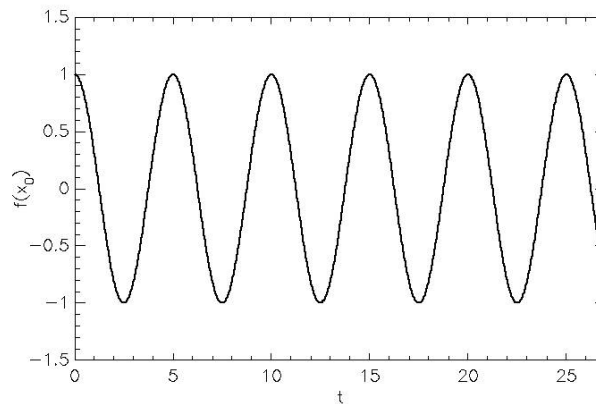
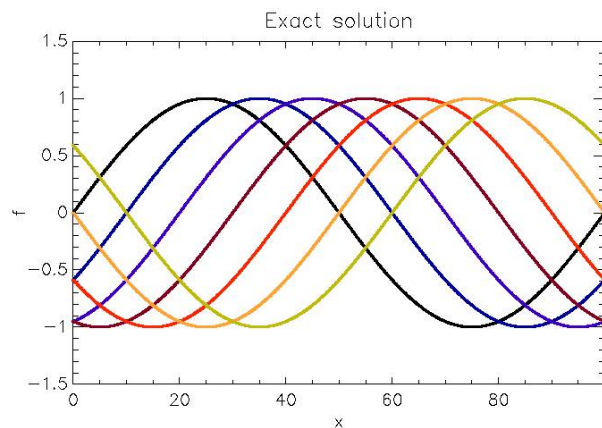
Lax-Wendroff method

$$A \leq 1 \Rightarrow \Delta t \leq \frac{\Delta x}{|v|} \quad \text{CFL stability condition}$$

Check the CFL condition before running a simulation!

Lax-Wendroff solution

$$f_i^{n+1} = \frac{(f_{i+1}^n + f_{i-1}^n)}{2} - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right)$$



**Signals are slowly decaying:
What's happening now?**



$$f_i^{n+1} = \frac{(f_{i+1}^n + f_{i-1}^n)}{2} - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right)$$

Rearranging the RHS

$$f_i^{n+1} = \frac{(f_{i+1}^n + f_{i-1}^n)}{2} - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right) + f_i^n - f_i^n \Rightarrow$$

$$\Rightarrow f_i^{n+1} = \frac{(f_{i+1}^n + f_{i-1}^n - 2f_i^n)}{2\Delta x^2} \Delta x^2 - v\Delta t \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right) + f_i^n \Rightarrow$$

$$\Rightarrow \frac{f_i^{n+1} - f_i^n}{\Delta t} + v \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right) - \frac{(f_{i+1}^n + f_{i-1}^n - 2f_i^n)}{2\Delta x^2} \frac{\Delta x^2}{\Delta t} = 0$$

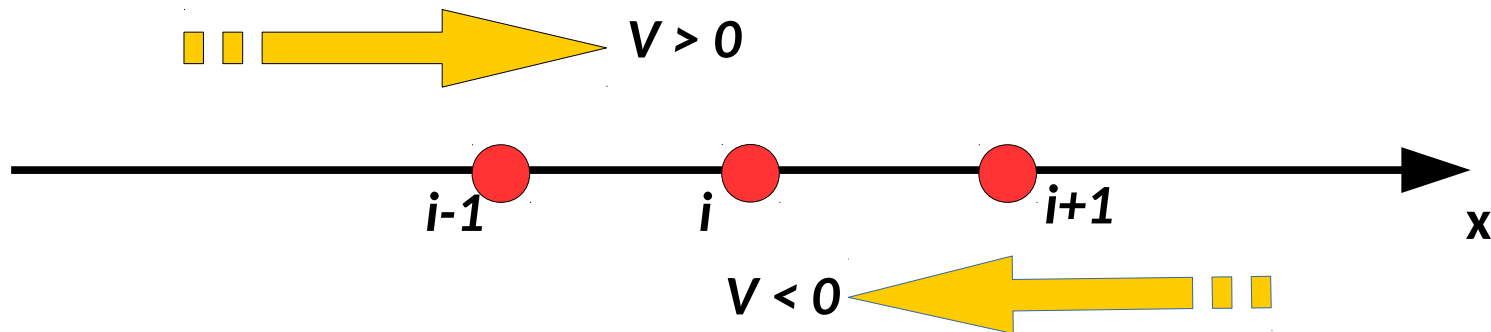


$\Delta t, \Delta x \rightarrow 0$

$$\frac{\partial f}{\partial t} + v \frac{\partial f}{\partial x} - \nu \frac{\partial^2 f}{\partial x^2} = 0; \quad \nu = \frac{\Delta x^2}{2\Delta t}$$

We have an additional term!
We are not solving anymore
the original equation

Check the consistency of an algorithm before running a simulation



$$f_i^{n+1} = f_i^n - |v| \Delta t \left(\frac{f_i^n - f_{i-1}^n}{\Delta x} \right), \quad v > 0$$

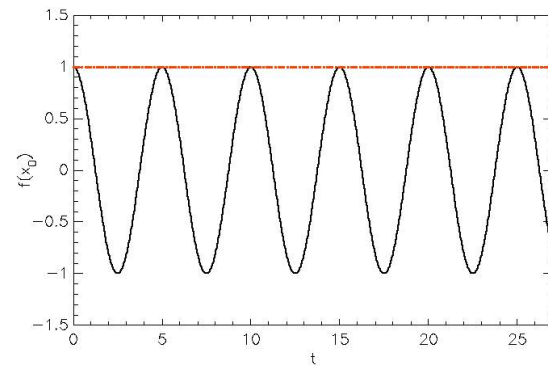
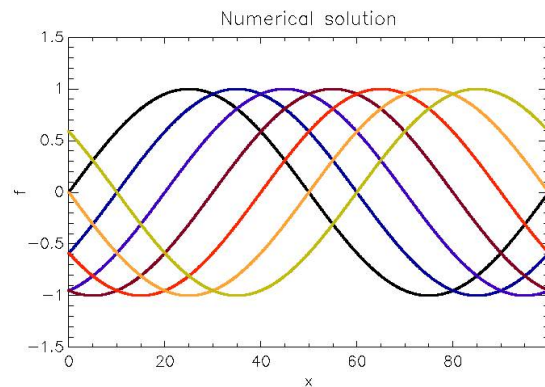
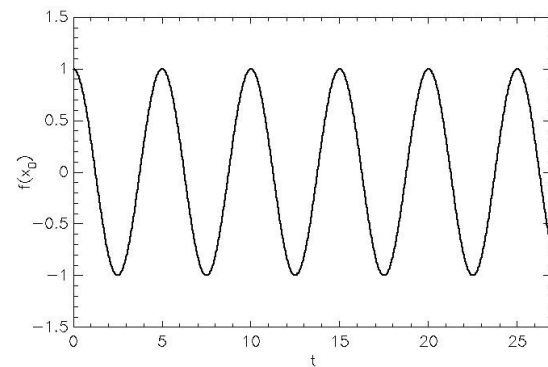
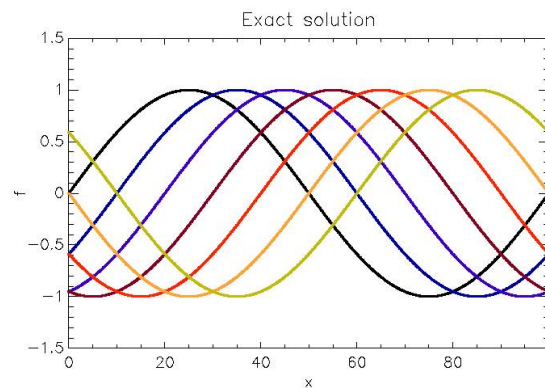
$$f_i^{n+1} = f_i^n + |v| \Delta t \left(\frac{f_{i+1}^n - f_i^n}{\Delta x} \right), \quad v < 0$$

$$A \leq 1 \Rightarrow \Delta t \leq \frac{\Delta x}{|v|}$$

CFL stability condition

$$f_i^{n+1} = f_i^n - |v| \Delta t \left(\frac{f_i^n - f_{i-1}^n}{\Delta x} \right), \quad v > 0$$

$$f_i^{n+1} = f_i^n + |v| \Delta t \left(\frac{f_{i+1}^n - f_i^n}{\Delta x} \right), \quad v < 0$$



**Upwind
schemes work
quite well!**

Widely adopted for integration of the Vlasov equation

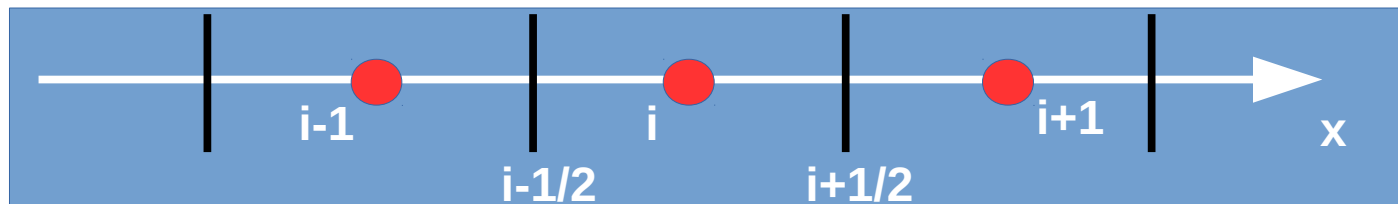
Van Leer, JCP, 1974, 1977a, 1977b, 1979; Mangeney+, JCP, 2000

$$\bar{f}_i(t) = \frac{1}{\Delta x} \int_{x_i - \frac{\Delta x}{2}}^{x_i + \frac{\Delta x}{2}} f(x, t) dx$$

The unknowns are the spatial averages of the function itself



Averaging and integrating advection equation in time gives:

$$\begin{aligned} \bar{f}_i(t + \Delta t) &= \bar{f}_i(t) - \frac{v\Delta t}{\Delta x} \int_t^{t+\Delta t} dt' \int_{x_i - \frac{\Delta x}{2}}^{x_i + \frac{\Delta x}{2}} \frac{\partial f(x, t')}{\partial x} dx = \\ &= \bar{f}_i(t) - \frac{v\Delta t}{\Delta x} \int_t^{t+\Delta t} dt' \left[f\left(x_i + \frac{\Delta x}{2}, t'\right) - f\left(x_i - \frac{\Delta x}{2}, t'\right) \right] \end{aligned}$$




Implicit scheme

$$\frac{f_i^{n+1} - f_i^n}{\Delta t} = -v \left(\frac{f_{i+1}^n - f_{i-1}^n}{2\Delta x} \right) \quad \text{Explicit scheme}$$

$$\left. \frac{\partial f}{\partial t} \right|_{t_n} \quad \text{forward} \qquad \left. \frac{\partial f}{\partial t} \right|_{t_{n+1}} \quad \text{backward}$$



$$f_i^{n+1} = f_i^n - v\Delta t \left(\frac{f_{i+1}^{n+1} - f_{i-1}^{n+1}}{2\Delta x} \right) \Rightarrow$$

$$\left(-\frac{v\Delta t}{2\Delta x} \right) f_{i-1}^{n+1} + f_i^{n+1} - \left(\frac{v\Delta t}{2\Delta x} \right) f_{i+1}^{n+1} = f_i^n$$

LHS: Linear combination of unknowns: **fully implicit schemes**

Implicit schemes

$$-\alpha f_{i-1}^{n+1} + f_i^{n+1} + \alpha f_{i+1}^{n+1} = f_i^n; \quad \forall i = 1, \dots, N_x; \quad \alpha = \frac{v\Delta t}{\Delta x}$$

$$\begin{pmatrix} M \end{pmatrix} \begin{pmatrix} f_i^{n+1} \end{pmatrix} = \begin{pmatrix} f_i^n \end{pmatrix}$$

The solution for f at step $n+1$ is obtained by solving this linear system, through standard linear algebra routines

Example: $N_x=6$ and periodic boundary conditions

$$M = \begin{bmatrix} 1 & \alpha & 0 & 0 & 0 & -\alpha \\ -\alpha & 1 & \alpha & 0 & 0 & 0 \\ 0 & -\alpha & 1 & \alpha & 0 & 0 \\ 0 & 0 & -\alpha & 1 & \alpha & 0 \\ 0 & 0 & 0 & -\alpha & 1 & \alpha \\ \alpha & 0 & 0 & 0 & -\alpha & 1 \end{bmatrix}$$

M is a tridiagonal matrix with full corners

Fully implicit schemes are unconditionally STABLE!
... but there is a big matrix to invert!

Let's go back to our 1D-1V Vlasov equation

$$\frac{\partial f}{\partial t} + v \frac{\partial f}{\partial x} + a \frac{\partial f}{\partial v} = \frac{\partial f}{\partial t} + \frac{\partial(vf)}{\partial x} + \frac{\partial(af)}{\partial v} = 0$$

$$f = f(x, v, t); \quad a = a(x)$$

Now we know how to solve advection equations. Let's split the evolution in 2 parts:

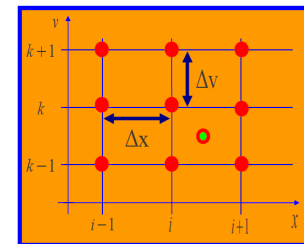
$$\frac{\partial f_x}{\partial t} + v \frac{\partial f_x}{\partial x} = 0$$

$$\frac{\partial f_v}{\partial t} + a \frac{\partial f_v}{\partial v} = 0$$



$$f_x(t + \Delta t) = \Lambda_x(\Delta t) f_x(t)$$

$$f_v(t + \Delta t) = \Lambda_v(\Delta t) f_v(t)$$



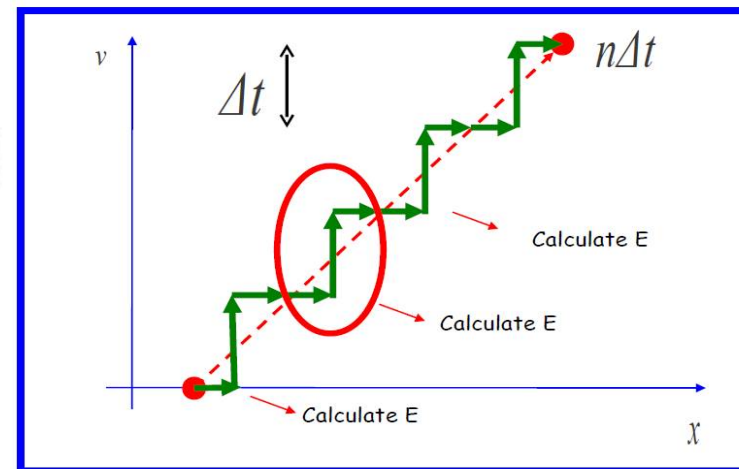
Couple $f_x(x, t)$ and $f_v(x, t)$ to get a solution for $f(x, v, t)$:

$$f(n\Delta t) = \{\Lambda_x(\Delta t/2) \Lambda_v(\Delta t) \Lambda_x(\Delta t/2)\}^n f_0 + o(\Delta t^3)$$

The splitting scheme

Cheng & Knorr, JCP, 1976;

Generalized to 6D in Mangeney et al. JCP, 2000



Take-home messages from Part I

- Equations that describe the plasma in a self-consistent way are very complex and computationally demanding
- There are two approaches to study plasma dynamics: the Eulerian (solve Eq.s for the distribution function) and the Lagrangian approach (solve equations for particles)
- Basic numerical scheme do not work properly. Test your scheme BEFORE running “important simulations”
- Advanced, high order methods give satisfactory results
- Numerical simulations are complementary to observational data. Understanding the reality cannot rely on simulations or observation alone, comprehension is given by a right balance among the two.



*“Calculators can only calculate -
they cannot do mathematics.”*

J. A. Van de Walle