

# New formulations of the semi-Lagrangian method for Vlasov-type equations

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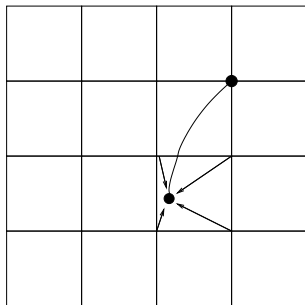
*In collaboration with*  
N. Crouseilles, G. Latu, M. Mehrenberger, T. Respaud

# Numerical methods for gyrokinetic model

- Most codes still based on Particle-In-Cell method and its variants.
- More codes now use phase-space grid. Such codes exist in Japan, USA and Europe.
- Most use eulerian methods from fluid dynamics.
- We have contributed to GYSELA code developed by V. Grandgirard which is based on semi-Lagrangian method.

- 1 Classical semi-Lagrangian method
- 2 Conservative semi-Lagrangian method
- 3 Forward semi-Lagrangian method

# The backward semi-Lagrangian Method



- $f$  conserved along characteristics
- Find the origin of the characteristics ending at the grid points
- Interpolate old value at origin of characteristics from known grid values → High order interpolation needed

- Typical interpolation schemes.
  - Cubic spline (Cheng-Knorr)
  - Cubic Hermite with derivative transport (Nakamura-Yabe)

# History of semi-Lagrangian schemes for Vlasov

- Cheng-Knorr (JCP 1976): splitting method for 1D Vlasov-Poisson. Cubic spline interpolation.
- ES-Roche-Bertrand-Ghizzo (JCP 1998): general semi-Lagrangian framework for Vlasov-type equations.
- Nakamura-Yabe (CPC 1999): semi-Lagrangian CIP method with Hermite interpolation
- Filbet-ES-Bertrand (JCP 2001) : semi-Lagrangian PFC method: positive and conservative
- N. Besse - ES (JCP 2003) : semi-Lagrangian solver on unstructured grids.
- Crouseilles-Mehrenberger-ES (2008): Equivalence of point based and conservative methods for Vlasov-Poisson + new class of positive filters.
- Crouseilles-Respaud-ES (2008): Forward semi-Lagrangian method.

# Convergence of semi-Lagrangian schemes

- Filbet (SINUM 2001): PFC for Vlasov-Poisson
- N. Besse (SINUM 2003): semi-Lagrangian method with linear for Vlasov-Poisson.
- Campos Pinto - Mehrenberger (Numer. Math. 2008): adaptive SL method for Vlasov-Poisson
- N. Besse (SINUM 2008): Convergence of semi-Lagrangian method with cubic Hermite interpolation.
- N. Besse - Mehrenberger (Math of Comp 2008): SL method for different classes of high-order interpolators.

- Transport equation

$$\frac{\partial f}{\partial t} + \mathbf{a} \cdot \nabla f = 0,$$

- Characteristics

$$\frac{dX}{dt} = \mathbf{a}$$

- Computation of the origin of the characteristics :
  - Explicit solution if  $\mathbf{a}$  does not depend on  $x$
  - Else, numerical algorithm needed.

# Splitting for exact computation of characteristics

- In many cases splitting can enable to solve a constant coefficient advection at each split step.
- E.g. separable Hamiltonian  $H(\mathbf{q}, \mathbf{p}) = U(\mathbf{q}) + V(\mathbf{p})$ .
  - Vlasov equation in canonical coordinates reads

$$\frac{\partial f}{\partial t} + \nabla_{\mathbf{p}} H \cdot \nabla_{\mathbf{q}} f - \nabla_{\mathbf{q}} H \cdot \nabla_{\mathbf{p}} f = 0.$$

- Split equations then become

$$\frac{\partial f}{\partial t} + \nabla_{\mathbf{p}} V \cdot \nabla_{\mathbf{q}} f = 0, \quad \frac{\partial f}{\partial t} - \nabla_{\mathbf{q}} U \cdot \nabla_{\mathbf{p}} f = 0,$$

where  $U$  does not depend on  $\mathbf{p}$  and  $V$  does not depend on  $\mathbf{q}$   $\rightarrow$  characteristics can be solved explicitly.

- Vlasov-Poisson falls into this category with  $\mathbf{q} = \mathbf{x}$ ,  $\mathbf{p} = \mathbf{v}$ ,  
 $H(\mathbf{x}, \mathbf{v}) = \frac{1}{2} m \mathbf{v}^2 + q \phi(\mathbf{x}, t)$ .



# First order computation of characteristics

- Consider general case. Characteristics defined by

$$\frac{d\mathbf{X}}{dt} = \mathbf{a}(\mathbf{X}, t).$$

- Backward solution:  $\mathbf{X}^n$  is known and  $\mathbf{a}^n$  known on the grid.
- Standard procedure to derive first order numerical method for EDO. Integrate on one time step and use quadrature formula for integral (left or right rectangle).

$$\mathbf{X}^{n+1} - \mathbf{X}^n = \Delta t \mathbf{a}^n(\mathbf{X}^n) \quad \text{or} \quad \mathbf{X}^{n+1} - \mathbf{X}^n = \Delta t \mathbf{a}^{n+1}(\mathbf{X}^{n+1}).$$

- No explicit solution:
  - Fixed point procedure needed in first case (e.g. Newton).
  - Predictor-corrector method on  $\mathbf{a}$  needed in second case.

## Second order method for Vlasov-Poisson (1/2)

Technique can be used for solving separable Hamiltonian Vlasov equations without splitting. We shall consider Vlasov-Poisson to illustrate this case.

- At time  $t_n$ :  $f^n$  and  $E^n$  are known at grid points.  $f^{n+1}$  and  $E^{n+1}$  need to be computed.
- Need to solve characteristics backward in time from  $t_{n+1}$  to  $t_n$

$$\frac{dV}{dt} = E(X(t), t), \quad \frac{dX}{dt} = V.$$

- Main problem  $E^{n+1}$  needed and not known.

## Second order method for Vlasov-Poisson (2/2)

A second order in time algorithm predictor-corrector algorithm is defined as follows:

- 1 Predict  $\bar{E}^{n+1}$  using continuity equation.
- 2 For all grid points  $x_i = X^{n+1}, v_j = V^{n+1}$  compute
  - $V^{n+1/2} = V^{n+1} - \frac{\Delta t}{2} \bar{E}^{n+1}(X^{n+1}),$
  - $X^n = X^{n+1} - \Delta t V^{n+1/2},$
  - $V^n = V^{n+1/2} - \frac{\Delta t}{2} E^n(X^n).$
  - Interpolate  $f^n$  at point  $(X^n, V^n).$
- 3 Yields first approximation of  $f^{n+1}(x_i, v_j) = f^n(X^n, V^n)$  that can be used to correct  $\bar{E}^{n+1}.$

Iterate until  $\bar{E}^{n+1}$  does not vary anymore.

At most a couple iterations required.

# A two step second order method

- Solve characteristics defined by  $\frac{d\mathbf{X}}{dt} = \mathbf{a}(\mathbf{X}, t)$ .
- Centered quadrature on two time steps:

$$\mathbf{X}^{n+1} - \mathbf{X}^{n-1} = 2\Delta t \mathbf{a}^n(\mathbf{X}^n), \quad \mathbf{X}^{n+1} + \mathbf{X}^{n-1} = 2\mathbf{X}^n + O(\Delta t^2).$$

- Use fixed point procedure to compute  $\mathbf{X}^{n-1}$  such that

$$\mathbf{X}^{n+1} - \mathbf{X}^{n-1} = \Delta t \mathbf{a}^n\left(\frac{\mathbf{X}^{n+1} + \mathbf{X}^{n-1}}{2}\right).$$

- **Problem:** compute  $f^{n+1}$  from  $f^{n-1}$ . Even and odd order time approximations become decoupled after some time. Artificial coupling needs to be introduced.

# A one step predictor-corrector second order method

- Solve characteristics defined by  $\frac{d\mathbf{X}}{dt} = \mathbf{a}(\mathbf{X}, t)$ .
- Centered quadrature on one time step:

$$\mathbf{X}^{n+1} - \mathbf{X}^n = \Delta t \mathbf{a}^{n+\frac{1}{2}}(\mathbf{X}^{n+\frac{1}{2}}), \quad \mathbf{X}^{n+1} + \mathbf{X}^n = 2\mathbf{X}^{n+\frac{1}{2}} + O(\Delta t^2).$$

- Now  $\mathbf{a}^{n+\frac{1}{2}}$  is unknown. Predictor-corrector procedure needed.
- Use fixed point procedure to compute  $\mathbf{X}^n$  such that

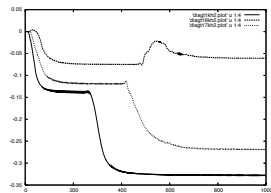
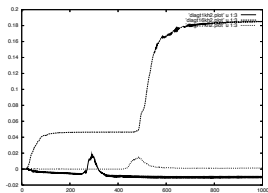
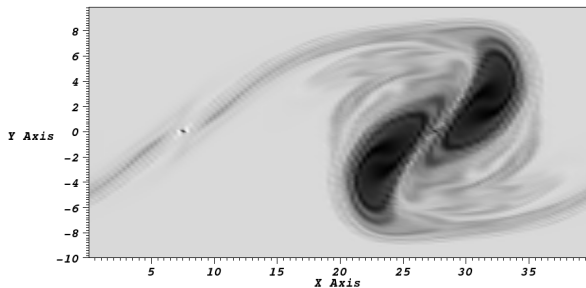
$$\mathbf{X}^{n+1} - \mathbf{X}^n = \Delta t \bar{\mathbf{a}}^{n+\frac{1}{2}} \left( \frac{\mathbf{X}^{n+1} + \mathbf{X}^n}{2} \right).$$

- Both predictor-corrector and fixed point iterations needed.

- 1 Classical semi-Lagrangian method
- 2 Conservative semi-Lagrangian method**
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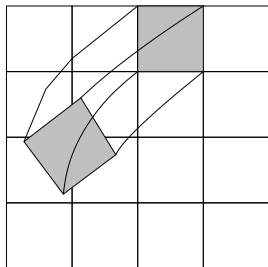
- When non conservative splitting is used for the numerical solver, the solver is not exactly conservative.
- Does generally not matter when solution is smooth and well resolved by the grid. The solver is still second order and yields good results.
- However: Fine structures develop in non linear simulations and are at some point locally not well resolved by the phase space grid.
- In this case a non conservative solvers can exhibit a large numerical gain or loss of particles which is totally unphysical.
- **Lack of robustness.**

# Vortex in Kelvin-Helmholtz instability





# Conservative semi-Lagrangian method



- Start from conservative form of Vlasov equation

$$\frac{\partial f}{\partial t} + \nabla \cdot (f \mathbf{a}) = 0.$$

- $\int_V f \, dx \, dv$  conserved along characteristics
- Three steps:
  - High order polynomial reconstruction.
  - Compute origin of cells
  - Project (integrate) on transported cell.
- Efficient with splitting in 1D conservative equations as cells are then defined by their 2 endpoints. A lot more complex for 2D (or more) transport.
- Splitting on conservative form: **always conservative**.

# High order polynomial reconstruction

- We only use the method with 1D splitting with equations in conservative form.
- Unknowns are cell averages:  $f_j = \frac{1}{\Delta x} \int f(x) dx$ .
- At time step  $t_n$  let  $f_j^n$  known average value of  $f^n$  on cell  $[x_{j-\frac{1}{2}}, x_{j+\frac{1}{2}}]$  of length  $h_j = x_{j+\frac{1}{2}} - x_{j-\frac{1}{2}}$ .
- Construct polynomial  $p_m(x)$  of degree  $m$  such that

$$\frac{1}{h_j} \int_{x_{j-\frac{1}{2}}}^{x_{j+\frac{1}{2}}} p_m(x) dx = f_j^n.$$

- Reconstruction by primitive.

# Choice of interpolation

- What interpolation should be chosen for primitive?
- Lagrange interpolation with centered stencil (used in PFC Filbet, ES, Bertrand JCP 2001).
- ENO type interpolation. Lagrange with varying stencil. Not efficient for Vlasov.
- Cubic spline interpolation: cubic polynomial on each cell, globally  $C^2$   
→ reconstructed function is then locally a quadratic polynomial and globally  $C^1$ . **Linked to cubic spline interpolation for classical semi-Lagrangian method.**

- Compute cell origins:
  - In 1D cell and its origin determined by end points. **Compute origin of end points** like in classical semi-Lagrangian method.
  - Need to make sure end points do not cross  $\rightarrow$  restriction on time step.
- Compute average value of  $f^{n+1}$  on cells using

$$\int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} f^{n+1}(x) dx = \int_{X(t_n; x_{i-\frac{1}{2}}, t_{n+1})}^{X(t_n; x_{i+\frac{1}{2}}, t_{n+1})} f^n(x) dx,$$

where  $f^n(x)$  is the high order reconstruction.

# Link between classical and conservative semi-Lagrangian methods

- For constant coefficient advections it can be shown that

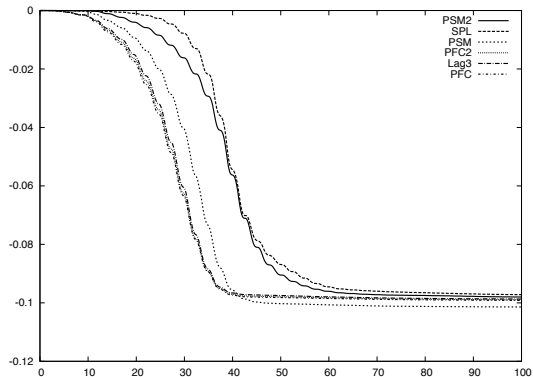
$$\begin{aligned} \text{C-Lag}(2d) &\iff \text{SL-Lag}(2d+1) \\ \text{PSM} &\iff \text{SPL} \end{aligned}$$

- Consequences :

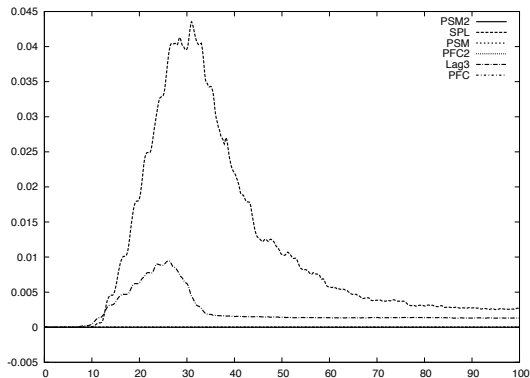
- 1 Classical and conservative semi-Lagrangian methods equivalent for constant coefficients split equations.
- 2 The PFC method (Filbet-ES-Bertrand, JCP 2001) corresponds for the Vlasov-Poisson (or Vlasov-Maxwell) systems to a classical semi-Lagrangian method with cubic Lagrange interpolation.

- Physical distribution function is always positive.
- High-order interpolation can lead to negative values in some zones.
- Reconstructed polynomial can be locally modified to remain positive.
- Performed for Lagrange reconstruction in PFC method.
- Introduces a little more dissipativity, but far less as monotonic reconstructions performed in fluid dynamics.

Evolution of  $L^2$  norm for  $N = 128$  for SLD

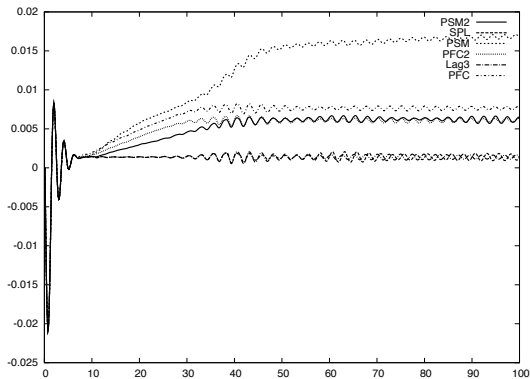


Evolution of  $L^1$  norm for  $N = 128$  for SLD

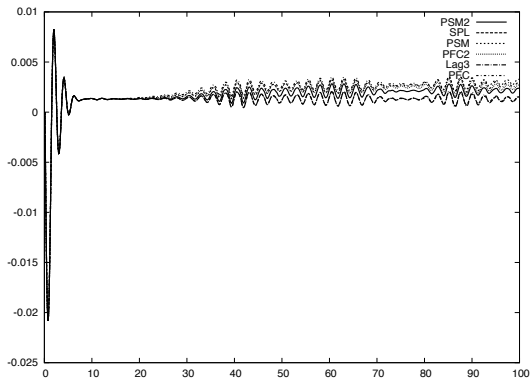




Evolution of total energy for  $N = 128$  for SLD

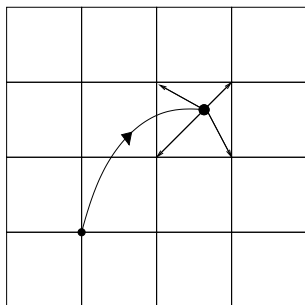


Evolution of total energy for  $N = 512$  for SLD



- 1 Classical semi-Lagrangian method
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# The forward semi-Lagrangian method



- $f$  conserved along characteristics
  - Characteristics advanced with same time schemes as in PIC method.
  - Leap-Frog Vlasov-Poisson
  - Runge-Kutta for guiding-center or gyrokinetic
- 
- Values of  $f$  deposited on grid of phase space using convolution kernel.
  - Identical to PIC deposition scheme but in whole phase space instead of configuration space only.
  - Similar to PIC method with reconstruction introduced by Denavit (JCP 1972).

# Discrete distribution function

- Function projected on partition of unity basis for conservativity.
- Linear B-splines very diffusive. Not useful in practice.
- Good choice is cubic B-splines.
- $f$  is reprojected on mesh at each time step.
- Between  $t_n$  and  $t_{n+1}$

$$f_h(x, v, t) = \sum_{i,j} w_{i,j} B(x - X(t; x_i, v_j, t_n)) B(v - X(t; x_i, v_j, t_n)).$$

Weight  $w_{i,j}$  associated to the particle starting from grid point  $(x_i, v_j)$  at  $t_n$  is coefficient of spline satisfying interpolation conditions

$$f_h(x_k, v_l, t_n) = \sum_{i,j} w_{i,j} B(x_k - x_i) B(v_l - v_j).$$

- Projection on phase space mesh is obtained with formula

$$f^{n+1}(x_k, v_l) = \sum_{i,j} w_{i,j} B(x_k - X(t_{n+1}, x_i, v_j, t_n)) B(v_l - V(t_{n+1}, x_i, v_j, t_n)).$$

# Time advance for Vlasov-Poisson

- As opposite to BSL, trajectories are advanced forward in time.
- Advection field is known at initial time. Standard ODE algorithms can be applied.
- For Vlasov-Poisson, we have  $z(t^n) = (x^n, v^n)$ , and  $E(t^n, z^n) = E(t^n, x^n)$ . **Separable Hamiltonian.**
- Natural scheme is Verlet algorithm, which is second order accurate in time

$$\text{Step1 : } v^{n+\frac{1}{2}} - v^n = \frac{\Delta t}{2} E(t^n, x^n),$$

$$\text{Step2 : } x^{n+1} - x^n = \Delta t v^{n+1/2},$$

$$\text{Step3 : } v^{n+1} - v^{n+\frac{1}{2}} = \frac{\Delta t}{2} E(t^{n+1}, x^{n+1}).$$

# Time advance for guiding-center

- Explicit Euler and Runge Kutta have been implemented.
- Second-order Runge Kutta method

$$\text{Step1 : } X^{n+1} - X^n = \Delta t E^\perp(t^n, X^n)$$

$$\text{Step2 : Compute } E^\perp(t^{n+1}, X^{n+1})$$

$$\text{Step3 : } X^{n+1} - X^n = \frac{\Delta t}{2} \left[ E^\perp(t^n, X^n) + E^\perp(t^{n+1}, X^{n+1}) \right]$$

- Fourth order Runge-Kutta

$$\text{Step1 : } k_1 = E^\perp(t^n, X^n)$$

$$\text{Step2 : Compute } k_2 = E^\perp(t^{n+\frac{1}{2}}, X^n + \frac{\Delta t}{2} k_1)$$

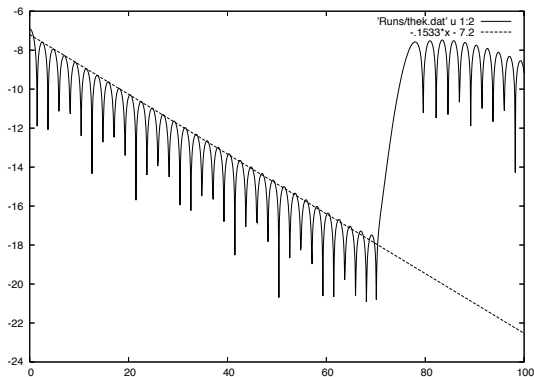
$$\text{Step3 : Compute } k_3 = E^\perp(t^{n+\frac{1}{2}}, X^n + \frac{\Delta t}{2} k_2)$$

$$\text{Step4 : Compute } k_4 = E^\perp(t^{n+1}, X^n + \Delta t k_3)$$

$$\text{Step5 : } X^{n+1} - X^n = \frac{\Delta t}{6} [k_1 + 2k_2 + 2k_3 + k_4]$$

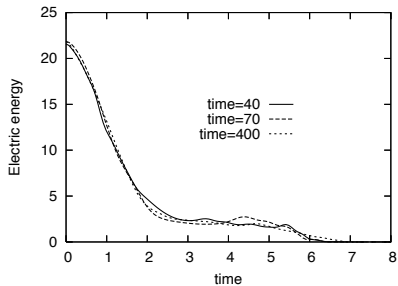
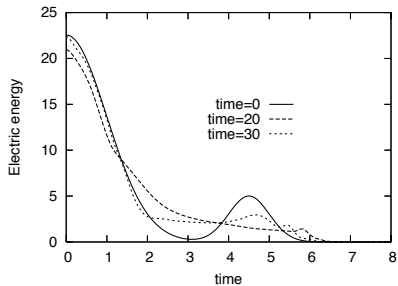
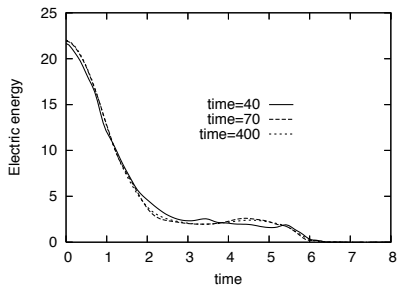
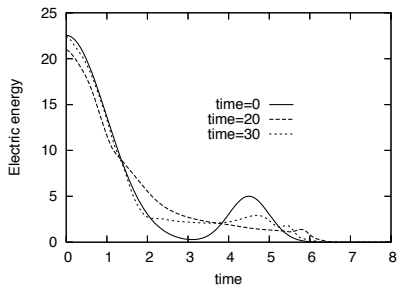
# Landau damping

$$f_0(x, v) = (1 + 0.001 \cos(kx)) \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}}, \quad L = 4\pi.$$

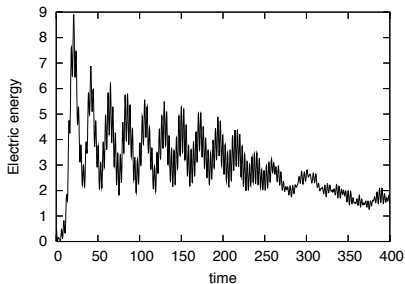
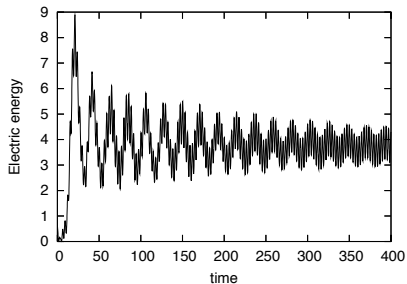




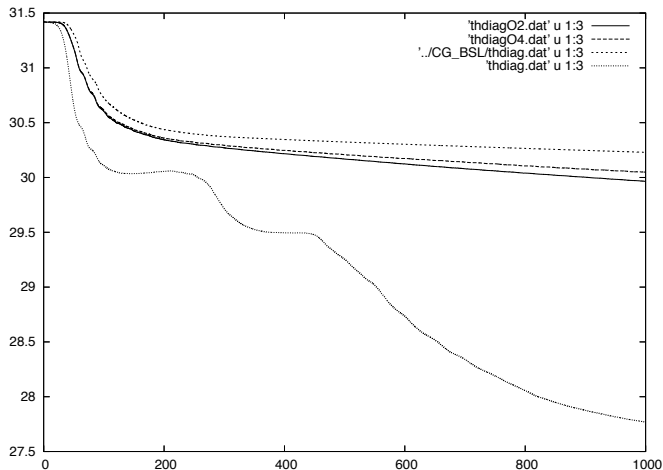
# Bump on tail (BSL (top) vs. FSL (bottom))



# Bump on tail: potential energy (BSL (left) vs. FSL (right))



# Energy conservation for Kelvin-Helmholtz instability



- Forward semi-Lagrangian method very promising.
- Accurate description of whole phase-space in particular tail of distribution function and small perturbations.
- A little more diffusive than BSL. Better with smaller time steps.
- Some advantages: classical explicit EDO solver can be used, in particular high order if needed. No need for predictor-corrector or fixed point algorithm.
- Can benefit from charge conserving PIC algorithms (Villasenor-Buneman).