

# Adaptive multiresolution semi-Lagrangian discontinuous Galerkin methods for the Vlasov equations

Nicolas Besse<sup>†</sup>

joint work with Eric Madaule and Erwan Deriaz

<sup>†</sup>Observatoire de la Côte d'Azur, Nice, France

Nicolas.Besse@oca.eu

# Contents

- 1 Vlasov equations
- 2 Adaptive multiresolution Semi-Lagrangian discontinuous Galerkin methods
- 3 Numerical results

# Vlasov equations

## The Vlasov equation (plasma setting)

- $(x, \xi) \in \mathbb{R}^d \times \mathbb{R}^d$ ,  $d = 1, \dots, 3$  : the phase-space
- $f(t, x, \xi)$  : the statistical distribution function of particles
- Vlasovian regime:  $(\nu_{coll}/\omega_p) \sim g = (n_0 \lambda_D^3)^{-1} \ll 1$ :

Individual and short-range interactions (collisions) are neglected. Collective and long-range Coulombian interactions are dominant and modeled by mean-fields:

$$\frac{\partial f}{\partial t} + v(\xi) \cdot \nabla_x f + (F_{\text{self}}(t, x, \xi) + F_{\text{applied}}(t, x, \xi)) \cdot \nabla_\xi f = 0$$

The relativistic velocity  $v(\xi)$  is given by:

$$v(\xi) = \frac{\xi/m}{\sqrt{1 + |\xi|^2/(mc)^2}}.$$

$F_{\text{self}}(t, x, \xi)$  is the Lorentz force given by:

$$F_{\text{self}}(t, x, \xi) = q(E(t, x) + v(\xi) \times B(t, x)).$$

The charge and current density are given by the two first moments of  $f$

$$\rho(t, \mathbf{x}) = q \int_{\mathbb{R}^d} f(t, \mathbf{x}, \xi) d\xi, \quad \mathbf{j}(t, \mathbf{x}) = q \int_{\mathbb{R}^d} \mathbf{v}(\xi) f(t, \mathbf{x}, \xi) d\xi.$$

The Vlasov-Poisson system (VP) (plasma case):

$$\mathbf{B} = 0, \quad \mathbf{E} = -\nabla\phi, \quad -\Delta\phi = \rho/\epsilon_0$$

The Vlasov-Quasistatic system (VQS):

$$\mathbf{E} = -\nabla\phi, \quad -\Delta\phi = \rho/\epsilon_0, \quad \mathbf{B} = \nabla \times \mathbf{A}, \quad -\Delta\mathbf{A} = \mu_0 \mathbf{j}.$$

The Vlasov-Darwin system (VD):

$$\left\{ \begin{array}{l} \frac{\partial \mathbf{E}_{irr}}{\partial t} - c^2 \nabla \times \mathbf{B} = -\mu_0 c^2 \mathbf{j}, \quad \frac{\partial \mathbf{B}}{\partial t} + \nabla \times \mathbf{E}_{sol} = 0, \\ \nabla \cdot \mathbf{E}_{irr} = \rho/\epsilon_0, \quad \nabla \times \mathbf{E}_{irr} = 0, \quad \nabla \cdot \mathbf{E}_{sol} = 0, \quad \nabla \mathbf{B} = 0, \\ \mathbf{E} = \mathbf{E}_{irr} + \mathbf{E}_{sol} \end{array} \right.$$

## The Vlasov-Maxwell system (VM):

$$\begin{cases} \frac{\partial E}{\partial t} - c^2 \nabla \times B = -\mu_0 c^2 j, & \frac{\partial B}{\partial t} + \nabla \times E = 0, \\ \nabla \cdot E = \rho / \epsilon_0, & \nabla \cdot B = 0. \end{cases}$$

For (VQS), (VD) and (VM) we have the compatibility condition for the source terms:

$$\frac{\partial \rho}{\partial t} + \nabla \cdot j = 0.$$

## Other Vlasov systems:

- Vlasov-gravitational:  
*Vlasov-Poisson, Vlasov-Nordström, Vlasov-Einstein, ...*
- Vlasov-plasma, classic and quantum versions:  
*Vlasov-waves, Vlasov-gyrokinetic, Vlasov-Wigner, Vlasov-Dirac-Benney, ...*

## Invariants

- $f_0(x, \xi) \geq 0 \implies f(t, x, \xi) \geq 0, \forall t > 0$
- If  $f$  and  $\beta$  enough regular,

$$\int \beta(f(t, x, \xi)) dx d\xi$$

- Especially the norm  $L^p, 1 \leq p \leq \infty$ 
  - $L^\infty \implies$  maximum principle
  - $L^1 \implies$  mass conservation
  - $L^2 \implies$  numerical dissipation
- $\beta(r) = r \ln r \implies$  “Boltzmann” kinetic entropy conservation

$$H(t) = \int f(t, x, \xi) \ln f(t, x, \xi) dx d\xi.$$

- Energy conservation:

$$(\text{VP}) : \quad \frac{1}{2m} \int f(t, x, \xi) |\xi|^2 dx d\xi + \frac{\varepsilon_0}{2} \int |E(t, x)|^2 dx,$$

(VM):

$$\int mc^2(\gamma(\xi) - 1)f(t, x, \xi)|\xi|^2 dx d\xi + \epsilon_0 \int \frac{|E(t, x)|^2 + c^2|B(t, x)|^2}{2} dx$$

(VD):

$$\int mc^2(\gamma(\xi) - 1)f(t, x, \xi)|\xi|^2 dx d\xi + \epsilon_0 \int \frac{|E_{irr}(t, x)|^2 + c^2|B(t, x)|^2}{2} dx$$

- momentum conservation:

(VP):

$$\int f(t, x, \xi)\xi dx d\xi$$

(VM):

$$\int f(t, x, \xi)\xi dx d\xi + \int dx \epsilon_0 E(t, x) \times B(t, x)$$

(VD):

$$\int f(t, x, \xi)\xi dx d\xi + \int dx \epsilon_0 E_{irr}(t, x) \times B(t, x)$$



## Characteristic curves equations

If  $\mathbf{a}(t, \mathbf{x}, \xi) = (v(\xi), F(t, \mathbf{x}, \xi))^T$  is enough regular (Lipschitz), we can introduce characteristic curves  $(X(t; \mathbf{s}, \mathbf{x}, \xi), \Xi(t; \mathbf{s}, \mathbf{x}, \xi))$  associated to the first differential operator

$$\frac{\partial}{\partial t} + \mathbf{a} \cdot \nabla_{(\mathbf{x}, \xi)}$$

which solves the classical ODEs equations

$$\left\{ \begin{array}{l} \frac{dX}{dt}(t; \mathbf{s}, \mathbf{x}, \xi) = v(\Xi(t; \mathbf{s}, \mathbf{x}, \xi)), \\ \frac{d\Xi}{dt}(t; \mathbf{s}, \mathbf{x}, \xi) = F(t, X(t; \mathbf{s}, \mathbf{x}, \xi), \Xi(t; \mathbf{s}, \mathbf{x}, \xi)), \\ X(\mathbf{s}; \mathbf{s}, \mathbf{x}, \xi) = \mathbf{x}, \quad \Xi(\mathbf{s}; \mathbf{s}, \mathbf{x}, \xi) = \xi, \end{array} \right.$$

In the case of (VP), (VQS), (VD) and (VM), as

$$\nabla_{(x,\xi)} \cdot a = 0$$

the jacobian of the map

$$(x, \xi) \mapsto (X, \Xi) = \varphi_t(x, \xi) = (X(t; s, x, \xi), \Xi(t; s, x, \xi)),$$

remains constant and the Lagrangian flow  $\varphi_t(x, \xi)$  preserves the measure and the connexity of the phase-space volume during time (phase-space incompressibility). This property implies the equivalence between the advective and the conservation form of the Vlasov Eq.:

$$\partial_t f + a \cdot \nabla_{(x,\xi)} f = 0 \iff \partial_t f + \nabla_{(x,\xi)} \cdot (af) = 0,$$

which traduces conservation of the matter (continuity equation in phase-space). The advective form traduces the fact that  $f$  is constant along characteristic curves

$$f(t, x, \xi) = f(s, X(s; t, x, \xi), \Xi(s; t, x, \xi)).$$

**RK: When variables are non-canonical  $\varphi_t$  is no more a volume-preserving map; the defect is compensated by the jacobian of the map "canonical  $\leftrightarrow$  non-canonical", but the flow is intrinsically volume-preserving (Liouville Th.)**

# Adaptive multiresolution Semi-Lagrangian discontinuous Galerkin methods

## The principle of discontinuous-Galerkin method

- 0) We consider a 1D scalar conservation law, because the multi-dimensional case is solved by using splitting methods.

- 1) 1D scalar conservation law,  $f = f(t, x)$ ,

$$\partial_t f + \partial_x(a(t, x)f) = 0 \quad + \text{initial conditions} + \text{boundary conditions}$$

- 2) We multiply the equation by a test function  $\varphi \in V$ , and we integrate on *each* cell  $I_i$

$$\int_{I_i} (\partial_t f + \partial_x(af)) \varphi \, dx = 0, \quad \forall \varphi \in V.$$

- 3) Weak formulation in space of the equation: integration by parts (IBP).

$$\int_{I_i} (\partial_t f \varphi - af \partial_x \varphi) \, dx + \int_{\partial I_i} af \cdot n \varphi \, d\gamma = 0, \quad \forall \varphi \in V.$$

- 4) Since  $V$  is a space of discontinuous functions, we replace fluxes by “numerical fluxes” at cell interfaces

$$\int_{I_i} (\partial_t f \varphi - af \partial_x \varphi) \, dx - (\widehat{af}\varphi)_{i-1/2} + (\widehat{af}\varphi)_{i+1/2} = 0, \quad \forall \varphi \in V$$

- 5) The choice of  $V$  (polynomial basis which can differ from a cell to another one) and numerical fluxes determine the properties of the numerical scheme (stability, consistency, convergence, high-order accuracy, conservation)

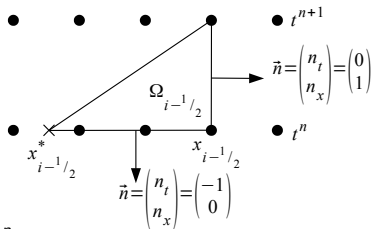
## The semi-Lagrangian discontinuous Galerkin method (SLDG)\*

After time integration between  $t^n$  and  $t^{n+1}$  we get, with  $f_h(t, x) = S_t(f_h^n)$  (where  $S_t(f_h^n)$  is the exact evolution operator starting with the initial condition  $f_h^n$ ),

$$\int_{I_i} f_h^{n+1} \cdot \varphi \, dx = \int_{I_i} f_h^n \cdot \varphi \, dx + \int_{t^n}^{t^{n+1}} \int_{I_i} a(t, x) S_t(f_h^n) \cdot \partial_x \varphi \, dx \, dt - \int_{t^n}^{t^{n+1}} \left( a(t, x) S_t(f_h^n) \cdot \varphi \Big|_{x_{i+1/2}^-} - a(t, x) S_t(f_h^n) \cdot \varphi \Big|_{x_{i-1/2}^+} \right) dt.$$

A space-time integration by parts (Divergence theorem) of the conservation law on the domain  $\Omega_{i-1/2}$  (cf. figure), transforms the time integration between  $t^n$  and  $t^{n+1}$  into a space integration (at time  $t^n$ ) which uses origin of characteristic curves at time  $t^n$  (noted  $x^*$ ). This integral is evaluated by using a Gauss-Legendre quadrature formula:

$$\int_{I_i} f_h^{n+1} \cdot \varphi \, dx = \int_{I_i} f_h^n \cdot \varphi \, dx + \Delta x_i \sum_{i_g} w_{i_g} \int_{\tilde{x}_{i_g}^*}^{\tilde{x}_{i_g}} f_h(\xi, t^n) \, d\xi \cdot \partial_x \varphi \Big|_{\tilde{x}_{i_g}} - \int_{x_{i+1/2}^*}^{x_{i+1/2}} f_h(\xi, t^n) \, d\xi \cdot \varphi(x_{i+1/2}^-) + \int_{x_{i-1/2}^*}^{x_{i-1/2}} f_h(\xi, t^n) \, d\xi \cdot \varphi(x_{i-1/2}^+)$$



## The characteristic discontinuous Galerkin method (CDG)\*

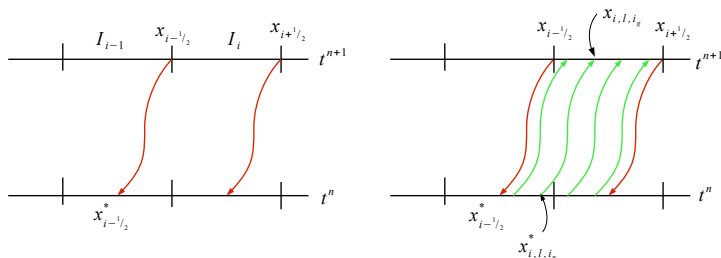
- Principle: we solve a dual problem. Indeed a test function  $\varphi(x) \in V$  (taken as final condition at time  $t^{n+1}$ ) is advected backward to the time  $t^n$  by the advection field  $a(t, x)$ .

We then get:

$$\int_{I_i} f_h^{n+1}(x) \varphi(x) dx = \int_{I_i^*} f_h(t^n, x) \varphi(t^n, x) dx, \quad I_i^* = [x_{i-1/2}^*, x_{i+1/2}^*]$$

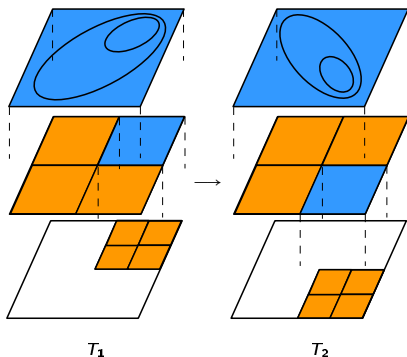
$$\approx \sum_{\ell} \sum_{i_g} w_{i_g} f_h(x_{i,\ell,i_g}^*, t^n) \varphi(x_{i,\ell,i_g}) \Gamma(I_{i,\ell}^*)$$

- $x_{i\pm 1/2}^*$ : origin of characteristics at time  $t^n$ , coming from mesh points  $x_{i\pm 1/2}$  at  $t^{n+1}$ .
  - Index “ $\ell$ ” labels intersected cells.
  - Index “ $i_g$ ” labels quadrature points (Gauss-Legendre).
  - $\Gamma(I_{i,\ell}^*)$  is the measure of the intersection between cells  $I_i^*$  and  $I_\ell$ .
- The method is illustrated by the following figure:



\*

# Principle of AMR



Example adaptive.

**Black:** level lines of the distribution function and mesh.

**Blue:** inactive cells, created but not used for computation.

**Orange:** active cells, created and used for computation.

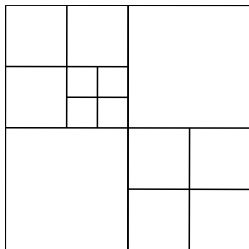
Here, we see that 3 levels are created, two of which are used for computation.

The mesh is more refined where variations of level lines are important.

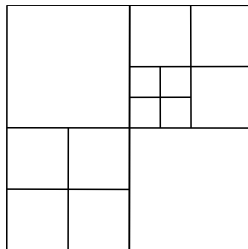
The typical loop of the algorithm:

- We suppose that we have an initial multi-wavelet representation of the distribution function associated to an initial (adaptive) mesh.
- Predict a new adaptive mesh by advecting forward (with a low-order scheme) cells of the initial mesh: Apply **cell creation & cell refinement**.
- Apply **SLDG or CDG schemes** by using the initial multi-wavelet representation of the distribution function projected on the predicted mesh. We then get a new multi-scale distribution function.
- Use a multi-wavelet decomposition of the new distribution function to discard details smaller than prescribed threshold: **coarsening**.

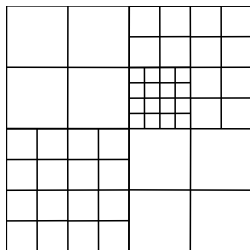
## An example of adaptive mesh prediction



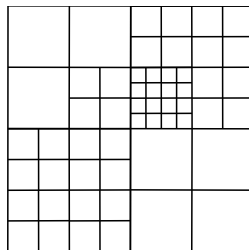
Initial mesh  $M^n$ .



Predicted mesh after forward advection.



Predicted mesh after refinement.



Merged mesh  $M^{n+1}$ .



## Multi-scale representation by means of multi-wavelets basis

- We consider the spaces  $V_n^k$  of  $\text{Dim}(V_n^k) = 2^n(k+1)$  and defined by

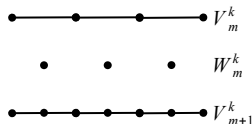
$$V_n^k = \{f : f|_{[2^{-n}l, 2^{-n}(l+1)]} \in \mathbb{P}^k, l = 0, \dots, 2^n - 1, f \text{ elsewhere}\}, k \in \mathbb{N}, n \in \mathbb{N}.$$

- We have inclusions

$$V_0^k \subset V_1^k \subset \dots \subset V_n^k \subset \dots \subset L^2.$$

- The multi-wavelets space  $W_n^k$  is defined such that

$$V_{n+1}^k = V_n^k \oplus W_n^k \quad \text{and} \quad V_n^k \perp W_n^k.$$



- We then have the multi-scale decomposition:

$$V_n^k = V_0^k \oplus W_0^k \oplus W_1^k \oplus \dots \oplus W_{n-1}^k,$$

or in other words the multi-scale representation of  $f$ :

$$f(t, x) = \sum_{j=0}^{k-1} \left( s_{j,0}^0(t) \phi_j(x) + \sum_{m=0}^{n-1} \sum_{l=0}^{2^m-1} d_{j,l}^m(t) \psi_{j,l}^m(x) \right)$$

- If  $\{\phi_j\}_{j=0..k-1}$  is a basis for  $V_0^k$  then  $\phi_{j,l}^n(x) = 2^{n/2} \phi_j(2^n x - l)$  is a basis for  $V_n^k$ .  
If  $\{\psi_j\}_{j=0..k-1}$  is a basis for  $W_0^k$  then  $\psi_{j,l}^n(x) = 2^{n/2} \psi_j(2^n x - l)$  is a basis for  $W_n^k$ .
- $s_{j,0}^0$  : scale coefficient.  $d_{j,l}^k$  : wavelet coefficient.
- Threshold criterion: if  $\|d_{j,l}^m\|_{\ell^2} < \epsilon_m(\epsilon_0, m)$ , then ignore details of level  $\geq m$

## Numerical results\*

\* Besse-Deriaz-Madaule JCP 332 2017

## Error for linear transport: rotation

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	4.47	2.57	-	0.401	-	0.210	-
0.01	2.09	0.0523	5.10	0.0149	4.32	0.0173	3.27
0.001	1.14	5.98 E-3	4.43	1.39 E-3	4.14	1.52 E-3	3.60
1 E-4	0.529	6.31 E-4	3.89	1.46 E-4	3.71	2.94 E-4	3.08
1 E-5	0.236	6.53 E-5	3.60	1.42 E-5	3.48	2.84 E-5	3.03

Error obtained with the AMW-SLDG scheme for polynomials of degree 2.

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	4.47	0.786	-	0.179	-	0.357	-
0.01	2.43	0.0118	6.87	3.72 E-3	6.33	6.81 E-3	6.47
0.001	2.09	5.85 E-3	6.42	1.34 E-3	6.42	1.72 E-3	6.99
1 E-4	1.23	1.38 E-3	4.92	3.03 E-4	4.95	4.63 E-4	5.16
1 E-5	0.625	7.11 E-5	4.73	1.76 E-5	4.69	4.22 E-5	4.59

Error obtained with the AMW-SLDG scheme for polynomials of degree 3.

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	4.47	2.57	-	0.401	-	0.210	-
0.01	2.09	0.0538	5.07	0.0149	4.31	0.0184	3.19
0.001	1.14	7.87 E-3	4.23	1.77 E-3	3.97	1.67 E-3	3.54
1 E-4	0.527	6.90 E-4	3.85	1.45 E-4	3.71	2.81 E-4	3.09
1 E-5	0.236	1.31 E-4	3.36	2.36 E-5	3.31	2.84 E-5	3.03

Error obtained with the AMW-CDG scheme for polynomials of degree 2.

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	4.47	0.787	-	0.179	-	0.357	-
0.01	2.43	0.0118	6.87	3.72 E-3	6.33	6.81 E-3	6.47
0.001	2.09	5.88 E-3	6.42	1.34 E-3	6.42	1.72 E-3	6.99
1 E-4	1.24	1.45 E-3	4.90	3.08 E-4	4.95	4.63 E-4	5.17
1 E-5	0.626	1.18 E-4	4.47	2.71 E-5	4.49	5.57 E-5	4.46

Error obtained with the AMW-CDG scheme for polynomials of degree 3.

## Error for nonlinear transport: Burgers equation

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	1.25	0.204	-	0.0860	-	0.0839	-
0.01	0.313	0.0105	2.14	3.80 E-3	2.25	3.54 E-3	2.28
0.001	0.157	1.30 E-3	2.43	4.78 E-4	2.50	4.65 E-4	2.50
1 E-4	0.0869	2.49 E-4	2.51	9.00 E-5	2.57	1.34 E-4	2.41
1 E-5	0.0412	2.40 E-5	2.65	8.28 E-6	2.71	1.00 E-5	2.64

Error obtained with the AMW-SLDG scheme for polynomials of degree 2.

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	1.25	0.187	-	0.0612	-	0.0454	-
0.01	0.627	8.94 E-3	4.38	2.99 E-3	4.36	2.88 E-3	3.98
0.001	0.313	5.26 E-4	4.24	1.89 E-4	4.17	2.09 E-4	3.88
1 E-4	0.198	1.78 E-4	3.77	7.46 E-5	3.64	1.09 E-4	3.27
1 E-5	0.157	3.28 E-5	4.16	1.19 E-5	4.11	1.38 E-5	3.89

Error obtained with the AMW-SLDG scheme for polynomials of degree 3.

threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	1.25	0.204	-	0.0860	-	0.0839	-
0.01	0.313	0.0105	2.14	3.80 E-3	2.25	3.54 E-3	2.28
0.001	0.157	1.30 E-3	2.43	4.78 E-4	2.50	4.65 E-4	2.50
1 E-4	0.0869	2.48 E-4	2.52	9.00 E-5	2.57	1.34 E-4	2.41
1 E-5	0.0412	2.30 E-5	2.66	8.16 E-6	2.71	1.03 E-5	2.64

Error obtained with the AMW-CDG scheme for polynomials of degree 2.

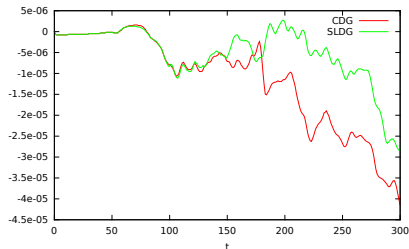
threshold	$\langle h \rangle$	$L^1$		$L^2$		$L^\infty$	
		error	order	error	order	error	order
0.1	1.25	0.187	-	0.0612	-	0.0454	-
0.01	0.627	8.94 E-3	4.38	2.99 E-3	4.36	2.88 E-3	3.98
0.001	0.313	5.26 E-4	4.24	1.89 E-4	4.17	2.08 E-4	3.88
1 E-4	0.198	1.78 E-4	3.77	7.46 E-5	3.64	1.08 E-4	3.27
1 E-5	0.157	3.24 E-5	4.16	1.19 E-5	4.11	1.34 E-5	3.91

Error obtained with the AMW-CDG scheme for polynomials of degree 3.

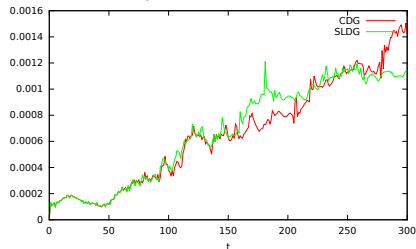
## Plasma case: bump-on-tail 1/2

$$\text{I.C.: } f_0(x, v) = \frac{(1 + 0.04 \cos(0.5x))}{10\sqrt{2\pi}} \left( 9 \exp\left(\frac{-v^2}{2}\right) + 2 \exp\left(-2(v - 4.5)^2\right) \right), \quad (x, v) \in [0, 20\pi] \times [-9, 9].$$

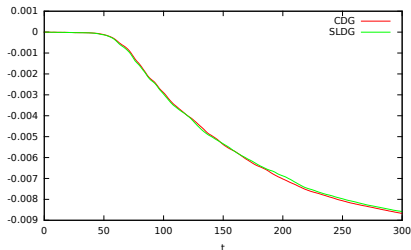
$\Delta t = 0.1$ . Maximum level of refinement is 8. Polynomial degree is 2. Threshold  $\epsilon_0 = 3 \times 10^{-3}$ .



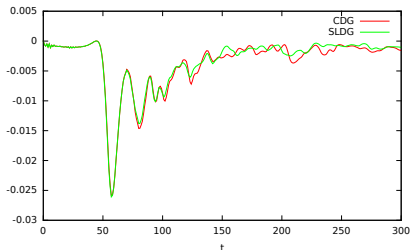
Relative variation of mass.



Relative variation of  $L^1$ -norm.

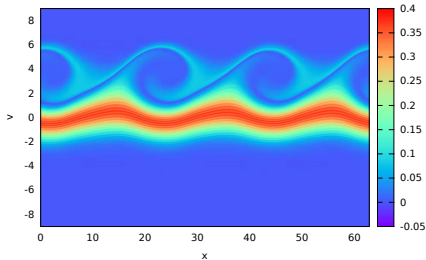


Relative variation of  $L^2$ -norm.

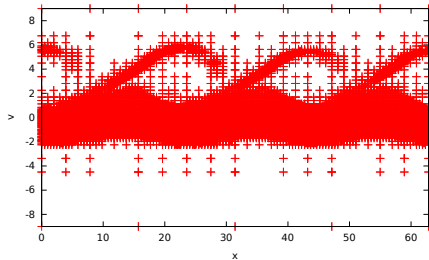


Relative variation of total energy.

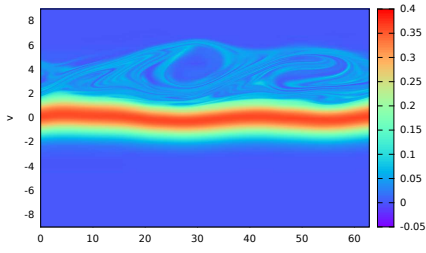
## Plasma case: bump-on-tail 2/2



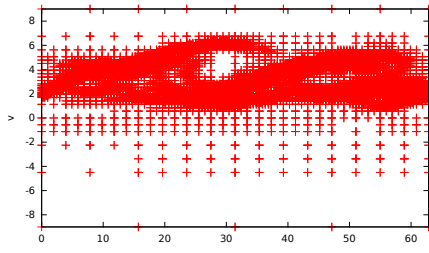
Distribution function at  $t = 60$  pp.



Mesh at  $t = 60$  pp.



Distribution function at  $t = 200$  pp.



Mesh at  $t = 200$  pp.

Distribution function and Mesh for bump on tail with the AMW-SLDG scheme.

## Plasma case: focusing beam 1/2

Vlasov-Poisson equation with cylindrical symmetry:

$$\partial_t f(r, v, t) + \partial_r \left( \frac{v}{\varepsilon} f(r, v, t) \right) + \partial_v \left( (E_\varepsilon(r, t) + F_\varepsilon(r, t)) f(r, v, t) \right) = 0,$$

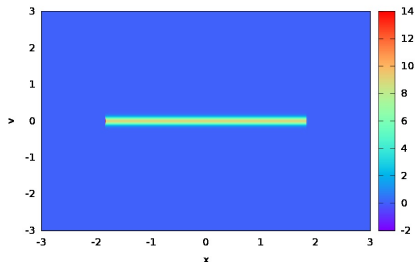
$$\frac{1}{r} \partial_r (r E_\varepsilon(r, t)) = \rho(r, t).$$

$$\text{Initial Condition: } f_0(r, v) = \frac{3}{4v_{th}} \exp\left(\frac{-v^2}{2v_{th}^2}\right) \mathbb{1}_{[-1.8, 1.8]}(r), \quad (r, v) \in [-3, 3]^2.$$

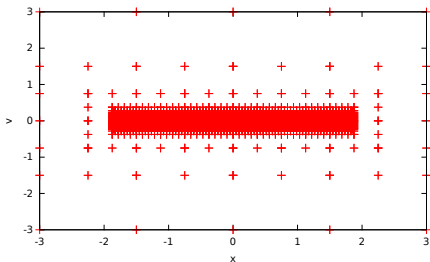
$$\text{Focusing external field: } F_\varepsilon(r, t) = r \left( \frac{-1}{\varepsilon} + \cos^2\left(\frac{t}{\varepsilon}\right) \right).$$

$\varepsilon = 0.1$ .  $v_{th} = 0.07$ .  $\Delta t = 0.02$ . Polynomials of degree up to 2.

Maximum level of refinement is 8. Threshold  $\epsilon_0 = 10^{-2}$ .

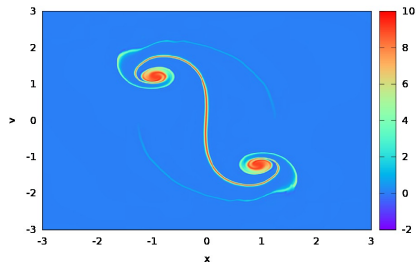


Initial distribution function.

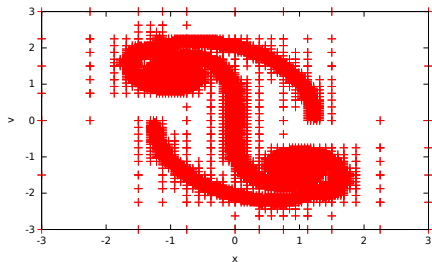


Initial mesh.

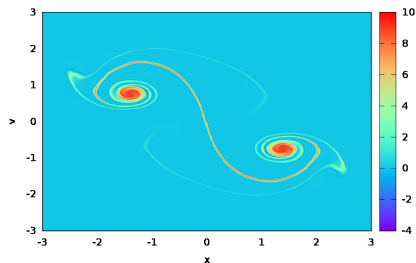
## Plasma case: focusing beam 2/2



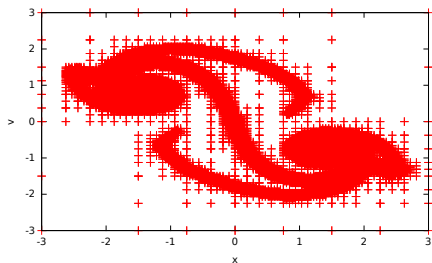
Distribution function at  $t = 16$ .



Mesh at  $t = 16$ .



Distribution function at  $t = 20$ .



Mesh at  $t = 20$ .

Distribution function and mesh for focusing beam with the AMW-SLDG scheme.



## Astrophysic case: cold layer 1/2

$$f_0(x, v) = \frac{1}{0.15\sqrt{2\pi}} \exp\left(-\frac{(v - u(x))^2}{2 \times 0.15^2}\right), \quad (x, v) \in [0, 2\pi] \times [-10, 10],$$

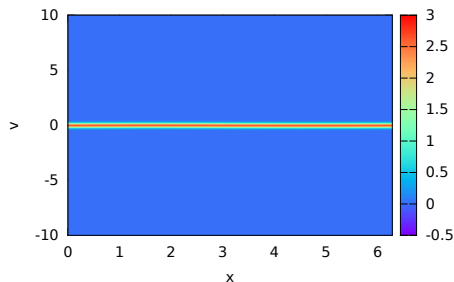
$$u(x) = 0.01 \sin(x).$$

$\Delta t = 0.02$ .

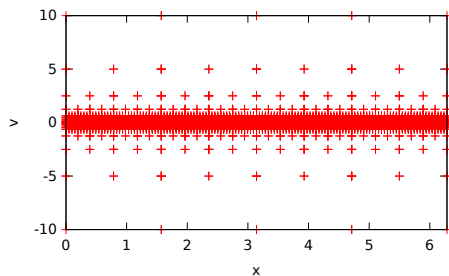
Polynomials of degree 2.

Maximum level of refinement is 8.

Threshold is  $\epsilon_0 = 3 \times 10^{-3}$ .

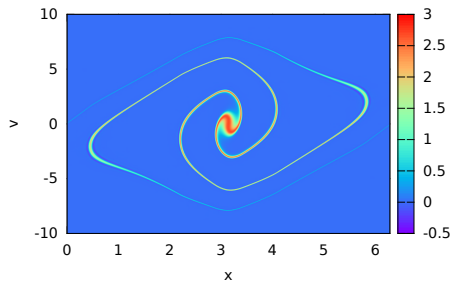


Initial distribution function.

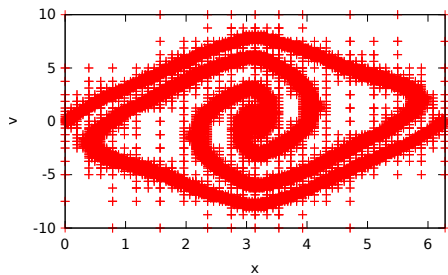


Initial mesh.

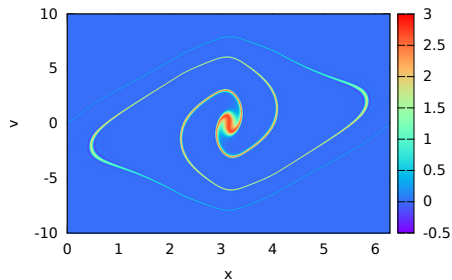
## Astrophysic case: cold layer 2/2



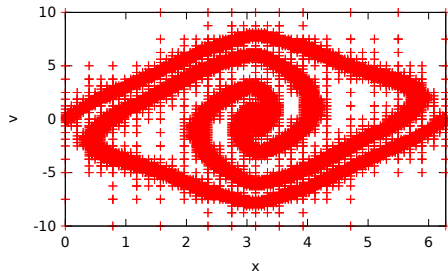
AMW-CDG



AMW-CDG



AMW-SLDG



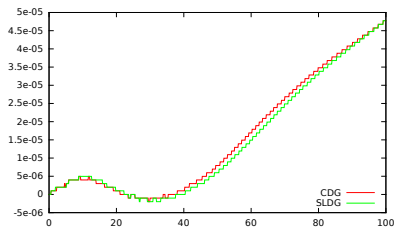
AMW-SLDG

Distribution function and mesh at  $t = 3$ .

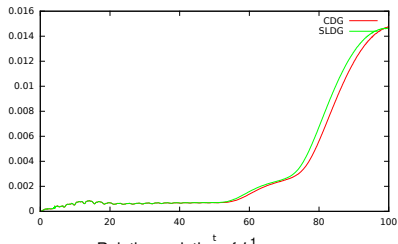
## Astrophysic case: gaussian distribution 1/3

$$\text{Initial condition: } f_0(x, v) = 4 \exp\left(-\frac{(x^2 + v^2)}{0.08}\right), \quad (x, v) \in [-2, 2]^2$$

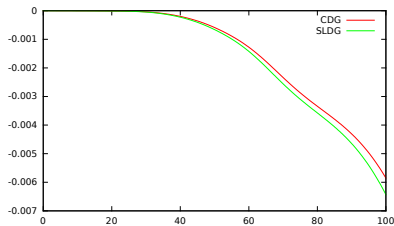
$\Delta t = 0.1$ . Polynomials of degree up to 3. Maximum level of refinement is 9. Threshold is  $\epsilon_0 = 3 \times 10^{-3}$ .



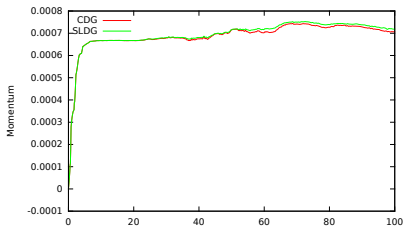
Relative variation of mass.



Relative variation of  $L^1$ -norm.

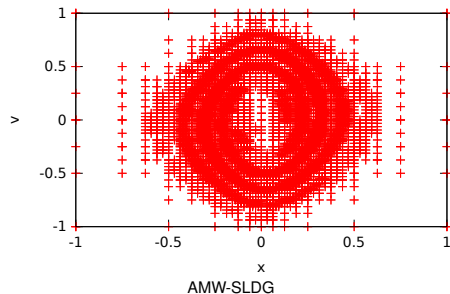
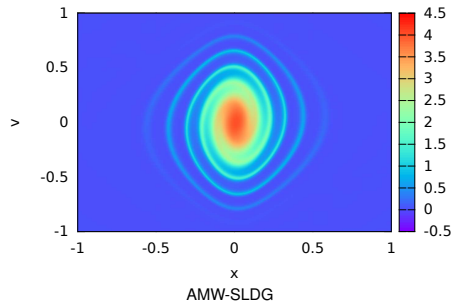
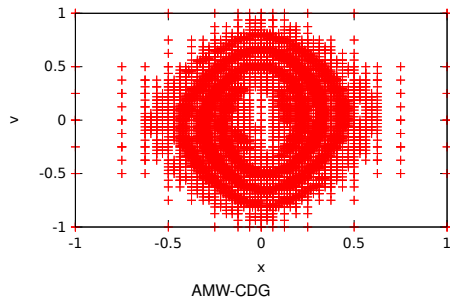
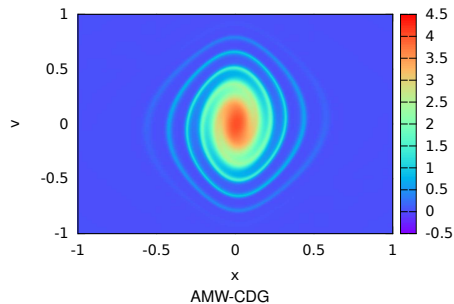


Relative variation of  $L^2$ -norm.



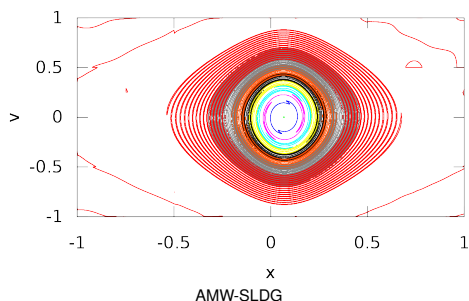
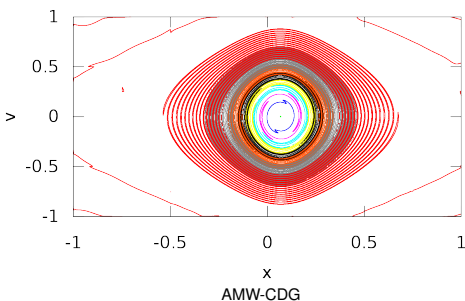
Variation of momentum.

## Astrophysic case: gaussian distribution 2/3

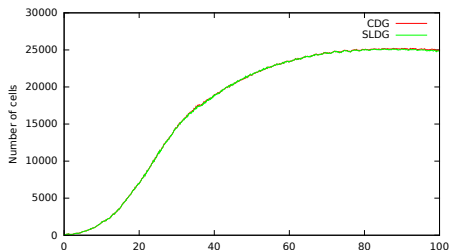


Distribution function and mesh at  $t = 15$ .

## Astrophysic case: gaussian distribution 3/3



Level lines of the distribution function at  $t = 100$ .



### ◀ Number of cells

- Adaptive mesh: 25000 cells
- Uniform mesh on  $[-2, 2]^2$ : 262000 cells
- Uniform mesh on  $[-1, 1]^2$ : 65500 cells

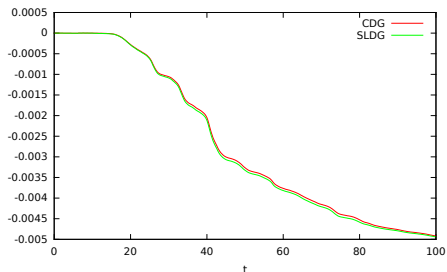
## Astrophysic case: Jeans instability and return to BGK stationary state

$$f(x, v, 0) = \frac{\exp\left(\frac{-v^2}{2}\right)}{\sqrt{2\pi}} (1 - A \cos(kx)), \quad (x, v) \in [0, 2\pi/k] \times [-V_c, V_c].$$

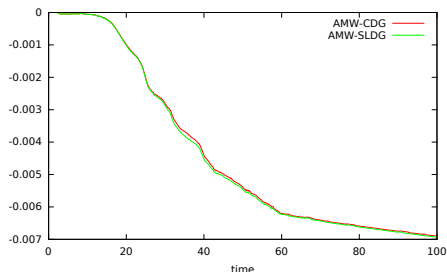
$A = 0.01$ ,  $k = 0.8$  and  $V_c = 6$ .  $\Delta t = 0.1$ .

Polynomials of degree 2.

Maximum level of refinement is 8.



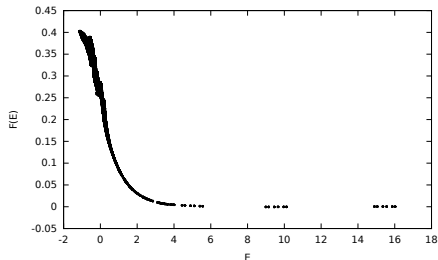
$\epsilon_0 = 0.003$ , 8 refinement levels



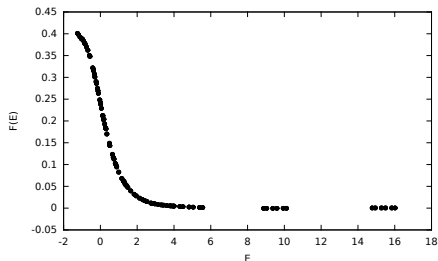
$\epsilon_0 = 0.01$ , 7 refinement levels

Time evolution of  $L^2$ -norm

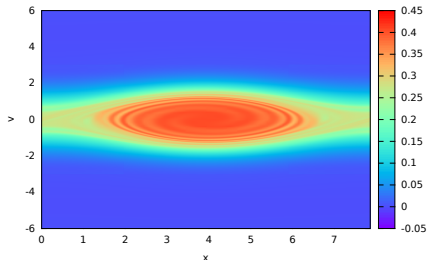
# Astrophysic case: Jeans instability and return to BGK stationary state



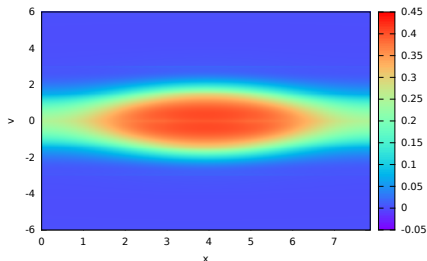
$E = v^2/2 - \phi(t = 100, x)$ ,  $\epsilon_0 = 0.003$ , 8 refinement levels



$E = v^2/2 - \phi(t = 100, x)$ ,  $\epsilon_0 = 0.01$ , 7 refinement levels



$f(t = 100, x, v)$ ,  $\epsilon_0 = 0.003$ , 8 refinement levels



$f(t = 100, x, v)$ ,  $\epsilon_0 = 0.01$ , 7 refinement levels

AMW-CDG scheme



THANK YOU FOR YOUR ATTENTION