Semi-Lagrangian particles or 6D Vlasov on single core

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A brief and biased history of particle methods for Vlasov-Poisson and flow simulations

A /	Numerics	Vlasov	Incompressible flows	Compressible flows
80's		Numerical analysis (PIC, VIC,)		Design of SPH
	Numerical issues Noise / accuracy	Random init n _{part} >> N _{cell}	Location processing technique	Renormalization h << ε
	Field solve	Grid/FFT	Grid-free/Biot-Savart Fast N-body solvers	no field
90's	Accuracy		Particle remeshing	
2000	Field solve		FFT based	
2005-	Accuracy		-High order remeshing -Directional splitting -S.L particles -Multi-resolution particles -GPU implementation	



Vlasov-Maxwell equations

Distribution function for one specie of ions (or electrons) subject to electric and magnetic fields satisfy

$$f = f(\mathbf{x}, \mathbf{v}, \mathbf{t}) \in [\mathbf{0}, \mathbf{1}]$$
 $\mathbf{E} = \mathbf{E}(\mathbf{x}, \mathbf{t})$ $\mathbf{B} = \mathbf{B}(\mathbf{x}, \mathbf{t})$

Conservation of charge:

$$\frac{\partial f}{\partial t} + (\mathbf{v} \cdot \nabla_{\mathbf{x}})\mathbf{f} + ((\mathbf{E} + \mathbf{v} \times \mathbf{B}) \cdot \nabla_{\mathbf{v}})\mathbf{f} = \mathbf{0}$$

+ Maxwell equations coupling E and B to moments of f (density and charge):

$$\rho(\mathbf{x}, t) = q \int f(\mathbf{x}, \mathbf{v}, t) \, dv$$
$$i(\mathbf{x}, t) = q \int \mathbf{v} f(\mathbf{x}, \mathbf{v}, t) \, dv$$



Transport equation in phase space (x,v) of dimension up to 6

Advection field given by

$$\mathbf{U} = \begin{bmatrix} \mathbf{v} \\ E + \mathbf{v} \times \mathbf{B} \end{bmatrix}$$

satisfies

$${\rm div}_{{\bf x},{\bf v}}{\bf U}={\bf 0}$$

conservative advection equation for f with velocity field U

-> conservation of all L^p norms of f



Computational complexity lies in

- space dimension (up to 6)
- transport equation where one wishes to conserve physical invariants and f∈[0,1]



however support of f occupies in general small part of phase space

These features make natural the use of Lagrangian particle methods :

- replace f by macro-particles in the phase space
- follow them with local velocities
- compute E, B fields in self-consistent way with integral formulas or FFT-based grid solvers



Advantages and drawback of Lagrangian (grid-free) particle methods

Pros:

- calculations restricted to the support of f
- norms of distribution function f, bounds of f and entropy ok
- large time-steps (see later)

Cons :

convergence analysis (and practice) shows that it is important to assure $\Delta x \ll \epsilon$ (or $n_{part} >> N_{cells}$)

→ need many particles to limit numerical noise in field evaluation

 \rightarrow expensive



Can Eulerian methods provide reasonable alternative in d>1 ?

Yes, if used in multi-resolution modes

Recent work by Deriaz and Periani (SIAM MMS 2018) : multi-resolution method based on wavelet analysis and third order finite-difference methods (despite lack of conservativity, numerical diffusion, and CFL conditions)

Enables simulations of 6D gravitational systems with acceptable memory and CPU times requirements

Suffers difficulties related to finite-difference solvers (in particular numerical diffusion)

Is there room for methods in-between grid-free particle methods and eulerian methods with (bonus) multi-resolution capabilities ?



Semi-Lagrangian methods are good candidates :

- they are well-adapted to advection-dominated problems (low numerical diffusion, even for low order methods)
- not constrained by CFL conditions



Classical semi-Lagrangian methods for transport and Vlasov (Cruseilles et al 2009, Sonnendrucker et al 2010 ..):



Forward Semi-Lagrangian : push trajectories, deposit mass though interpolation reconstruct grid values through B-splines



Backward Semi-Lagrangian : go backward on trajectories, interpolate from grid values

$$f(t, x, y) = \sum_{k,l} \omega_{k,l}^n S(x - X_1(t; x_k, y_l, t^n)) S(y - X_2(t; x_k, y_l, t^n)).$$



In (2D) FSL methods, to advance from time step t^n to t^{n+1} , the solution is represented on a B-spline basis :

$$f(t, x, y) = \sum_{k,l} \omega_{k,l}^n S(x - X_1(t; x_k, y_l, t^n)) S(y - X_2(t; x_k, y_l, t^n))$$

and weights ω^{n+1} at time t^{n+1} are recovered by solving linear system :

$$f_{i,j}^{n+1} = \sum_{k,l} \omega_{k,l}^{n+1} S(x_i - x_k) S(y_j - y_l).$$

Conservative methods can be constructed along similar lines

Drawbacks:

- cost (linear system)
- memory requirements (like Eulerian methods, need full grids)
- high order ?

So far, used mostly (only ?) for low-dimensional (1+1D) systems



Other approach to Semi-Lagrangian methods: remeshed particle methods

- try to combine natural adaptivity of particles with accuracy of Semi-Lagrangian methods
- well established for 3D level-set methods and complex flow calculations but (almost) never used so far for Vlasov-Poisson

Principle:

- initialize particles in support of f and push them with local velocities (like regular particles)
- remesh at each time-step with high order interpolation formulas
- retain after remeshing only particles with strength above a given cut-off



Remeshed particle methods

Idea goes back to the 80's: Krasny's 2D vortex sheet, Meiburg's 3D jets, and Chorin's and Leonard's hairpin removal Insert fresh particles «in between» old particles when needed

Specific to problems with topology control

More generic approach : remesh particles on regular grids through standard 3D **interpolation** formulas.

Criterium for remeshing schemes :

conservation of the moments of the particle distribution: $\int f d\mathbf{x}$, $\int \mathbf{x} f d\mathbf{x}$, $\int \mathbf{x}^2 f d\mathbf{x} \cdots$

Allowed first high resolution DNS of flow past cylinders at high Reynolds numbers (Koumoutsakos & Leonard, JFM 1995), ... before spectral element calculations



Traditionally, work with tensor products of ID formulas

Typical interpolation formulas :

•conservation of 3 moments (third order truncation error) use 3 points in each direction smooth version uses an additional grid point -> 4 grid points

•conservation of 5 moments (5th order truncation error) use 5 points smooth version spread particle on 6 nearest grid points

•resulting stencils in 3D : 27, 64, 125, 216 points

if advection of particles is split direction by direction, reduces to onedimensional stencils

Until recently particle remeshing was considered as an ad-hoc fix, and momentum conservation properties as safeguard



Particle methods with remeshing at every time-step can be viewed and analyzed as **forward semi-lagrangian** methods (C. et al, M2AN 2014)

How they work:

- I) particles on a grid
- 2) push particles with local velocity values

3) remesh particles on the grid, through interpolation

In I D for advection equation $\theta_t + u \theta_x = 0$

method can be described by the following equations:

$$x_i^{n+1} = x_i + \tilde{u}_i^n \Delta t \qquad \theta_i^{n+1} = \sum_j \theta_j^n \Gamma\left(\frac{x_j^{n+1} - x_i}{\Delta x}\right), i \in \mathbb{Z}^d$$

where \tilde{u}_i^n depend on the time-stepping scheme and Γ is a piecewise polynomial kernel



 Γ is defined by regularity and moment properties :

$$\bigstar \text{ moment properties}: \qquad \sum_{k \in \mathbb{Z}} k^{\alpha} \Gamma(x - k) = x^{\alpha}, \ 0 \le \alpha \le p, \ x \in \mathbb{R}$$

$$\bigstar \text{ regularity}: \qquad \Gamma \text{ is of class } C^{r} \quad \text{and } \Gamma \in \mathcal{C}^{\infty}\left(\left]l, l + 1\right[\right), \ l \in \mathbb{Z}$$

$$\bigstar \text{ interpolation property}: \qquad \Gamma(i) = \begin{cases} 1 \text{ if } i = 0, \\ 0 \text{ otherwise.} \end{cases}$$

Remark : $\Delta t \leq \|ec{
abla} ec{u}\|_{\infty}^{-1}$ is sometimes called a Lagrangian CFL condition (LCFL) with LCFL 1

If at all times, each cell contains exactly one particle, order = p



Examples of remeshing kernels (2nd and 6th order)

$$\Lambda_{4,2}(x) = \begin{cases} 1 - \frac{5}{4}|x|^2 - \frac{35}{12}|x|^3 + \frac{21}{4}|x|^4 - \frac{25}{12}|x|^5 & 0 \le |x| < 1\\ -4 + \frac{75}{4}|x| - \frac{245}{8}|x|^2 + \frac{545}{24}|x|^3 - \frac{63}{8}|x|^4 + \frac{25}{24}|x|^5 & 1 \le |x| < 2\\ 18 - \frac{153}{4}|x| + \frac{255}{8}|x|^2 - \frac{313}{24}|x|^3 + \frac{21}{8}|x|^4 - \frac{5}{24}|x|^5 & 2 \le |x| < 3\\ 0 & 3 \le |x| \end{cases}$$

$$\Lambda_{6,6}(x) = \begin{cases} 1 - \frac{49}{36} |x|^2 + \frac{7}{18} |x|^4 - \frac{1}{36} |x|^6 - \frac{46109}{144} |x|^7 + \frac{81361}{48} |x|^8 - \frac{544705}{144} |x|^9 + \frac{655039}{144} |x|^{10} \\ - \frac{223531}{72} |x|^{11} + \frac{81991}{72} |x|^{12} - \frac{6307}{36} |x|^{13}, \qquad 0 \leqslant |x| < 1 \\ - \frac{44291}{5} + \frac{1745121}{20} |x| - \frac{15711339}{12} |x|^2 + \frac{32087377}{30} |x|^3 - \frac{7860503}{80} |x|^4 + \frac{38576524}{15} |x|^5 \\ - \frac{24659323}{10} |x|^6 + \frac{84181657}{48} |x|^7 - \frac{74009313}{80} |x|^8 + \frac{17159513}{48} |x|^9 \\ - \frac{7870247}{80} |x|^{10} + \frac{438263}{24} |x|^{11} - \frac{81991}{80} |x|^{12} + \frac{6307}{60} |x|^{13}, \qquad 1 \leqslant |x| < 2 \\ 3905497 - \frac{424679647}{20} |x| + \frac{3822627865}{72} |x|^2 - \frac{2424839767}{30} |x|^3 + \frac{3009271097}{36} |x|^4 \\ - \frac{930168127}{15} |x|^5 + \frac{305535494}{20} |x|^6 - \frac{9998313437}{72} |x|^7 + \frac{203720335}{36} |x|^8 - \frac{137843153}{144} |x|^9 \\ + \frac{22300663}{144} |x|^{10} - \frac{6126883}{360} |x|^{11} + \frac{81991}{72} |x|^{12} - \frac{6307}{180} |x|^{13}, \qquad 2 \leqslant |x| < 3 \\ - \frac{255622144}{5} + \frac{971097344}{5} |x| - \frac{15295867328}{45} |x|^2 + \frac{5442932656}{15} |x|^3 - \frac{2372571796}{9} |x|^4 \\ + \frac{2064517469}{15} |x|^5 - \frac{9563054381}{180} |x|^6 + \frac{2210666335}{144} |x|^7 - \frac{796980541}{240} |x|^8 \\ + \frac{76474979}{144} |x|^9 - \frac{43946287}{720} |x|^{10} + \frac{343721}{72} |x|^{11} - \frac{81991}{360} |x|^{12} + \frac{901}{180} |x|^{13}, \qquad 3 \leqslant |x| < 4 \\ 0, \qquad 4 \le |x| \end{cases}$$



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Refinement study : case of a rotating patch in an off-center vorticity field



$$x,t) = 2\cos(\pi t/T) \left(\begin{array}{c} -\sin^2(\pi x)\sin(\pi y)\cos(\pi y) \\ \sin^2(\pi y)\sin(\pi x)\cos(\pi x) \end{array} \right)$$

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Kernel	Order of convergence
$\Lambda_{2,1}$	1.87
$\Lambda_{4,2}$	3.17
$\Lambda_{6,4}$	5.92

3D case : comparisons with Weno and VOF methods

Implementation of grid-based methods with particles for corrections



Enright et al. JCP 2002	Vincent et al, JCP 2010	N=100, CFL=8	N=160, CFL=12
3r order Weno	VOF		CPU time :
N=100 + 64 ppc	N=64 + 9 ppc		1 s per iteration
CFL=1 (?)	CFL=0.1	remeshed part	icle method, 4th order
CPU =??		remeshing,	
		2nd c	order in time





N = 256 and CFL = 30. Left picture: kernel $\Lambda_{2,1}$; right picture: kernel $\Lambda_{8,4}$.



Semi-Lagrangian (or remeshed) particle methods for Vlasov-Poisson

First attempt: Myers, Colella, Van Straalen (SIAM J. Sci. Comput., 2017) :

- Successful application to 2+2D Landau damping
- Use of 2nd and 4th order kernels
- roadblock for higher dimension : need a full grid to remesh particles
- still n>N, and remeshing frequency to adjust

roadblock can be removed by using link-lists of particles in lower dimensional spaces (C., JCP 2018)



2D example : particles in (x,y) space, directional splitting of advection

1st sweep : horizontal advection



sort particles by horizontal lines :

each particle on the line gets an address in the original list



line I=1

```
-> i=l:n(l)=3,
push
xp(l(i))=xp(l(i))+dt * vx(1)
```

```
remesh
i=int(xp):int(xp)+1, ug(i)=ug(i)+up*Γ(xp-i*dh) (for a 2 points formula)
```

Good news : Only 1d array for values on the grid ! Bad news : Need to label lines in a 6D space : 5d arrays



Good compromise between sizes of label and grid value arrays :

sort particles in 3D spaces : (x,y,z) spaces then (u,v,w) spaces

Remarks :

- accumulation of charges to compute density in a given (x,y,z) plane and calculation of field are done simultaneously with particle sorting
- push-remesh line by line can still be (and is) done inside each 3D space to reduce computational cost of high order remeshing kernels. Important for high order (large stencils) kernels.
- uniform velocity in each line -> order of method given by number of moments of cut-off (no need of regularity)

Memory requirements for 6D algorithm with Np particles :

- 7 main arrays of size Np for positions, velocity, distribution function
- 7 auxiliary arrays for same quantities
- 2 arrays of size Np to store particles addresses in link-list algorithm
- several 3D arrays for E, density, link-list



Goal of simulations :

test accuracy / efficiency of SL particles on uniform grids against MRA Eulerian methods and regular particle methods

Example 1: 2+2D plasma two beams instability, SL with 4th order kernel

$$f_0(x, y, u, v) = \frac{7}{4\pi} \exp\left(-\frac{u^2 + 4v^2}{8}\right) \sin^2\left(\frac{u}{3}\right) (1 + 0.05\cos(0.3x))$$
$$\Omega = \left[-\frac{10\pi}{3}, \frac{10\pi}{3}\right]^2 \times \left[-3\pi, 3\pi\right]^2$$

+ periodic boundary conditions for E.



MRA grid (Deriaz-Periani) 324-2564



number of particles on a 128⁴ grid



Qualitative comparison : cuts in (x,u) plane at t=12 (time of peak potential energy)



SL isolines of f MRA isolines of f MRA grid



Conservation properties :

total density, L2 norm, entropy and total energy $\int v^2 f \, dx \, dv + \int E^2 \, dx$



SL particle method grid with 128⁴ grid

MRA grid (Deriaz-Periani) 324-2564





Comparison of potential energy $\int E^2$ obtained by MRA and SL particles at two resolutions



First conclusions on this case :

- Accuracy of SL on uniform grids comparable to finite-difference MRA at higher local resolution
- Diffusive effects visible on Eulerian scheme, even at high (local) resolution, not seen on SL methods



Comparison with pure PIC method on 2+2D Landau Damping

$$f_0(x, y, u, v) = \frac{1}{2\pi} \exp\left(-u^2 - v^2\right) \left(1 + 0.01 \cos\frac{x}{2} \cos\frac{y}{2}\right)$$



SL particles on 32⁴ grid

Classical PIC on 256² cells, with varying particle/cell ratio (from Y. Barsamian PhD thesis)



Good fit for longer times with higher resolution for SL particles



SL particles on 32⁴ grid

SL particles on 64⁴ grid





density isosurfaces







In that example, using a higher order SL particle method can be useful





number of particles for SL methods on 96⁶ grid

For comparison the 326-5126 MR method of Deiraz-Peirani used up to 5 109 active grid points



Comparisons of kinetic energy : SL particles vs MRA vs reference grid-free particles (code GADGET) using 500 M particles (fixed number)





Conclusion for this case :

-as in the plasma case for a SL particle methods using roughly the same number of particles on a uniform grid, profiles of invariants and kinetic energy are similar to Eulerian MRA code (with less numerical dissipation at late times)

-results compare well with grid-free particle code using roughly same number of particles

So what ?



	/ +12	4th order SL PM	8th order SL PM	Wavelet MRA [7]	GADGET [7]
?	Effective grid resolution	96	96	32 to 512	N.A.
	Maximum number of active grid-points / particles	1.810^{8}	2.510^{8}	510^{9}	510^{8}
	Number of time-steps	100	100	1349	N.A.
	Wall clock CPU time	3.5 hours	5.8 hours	120 days	1 week
	Hardware	1 Intel Xeon E5-2640 2.5 GHz	1 Intel Xeon E5-2640 2.5 GHz	32 Intel Xeon X5650 2.66GHz	500 cores





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link-list
 particle-grid interpolation
 particle assignment

SL particles

MRA

breakdown of main steps of SL algorithm

The dark side :

- only very under-resolved simulations (can be seen on cuts of f)
- bounds of f ok for plasma case
 but (very) bad in astro case!





Conclusion :

There is room for simple methods on single cores even for complex problems

Combining SL particles with wavelets (already done for Navier-Stokes, Bergdorf et al SIAM MMS 2006) should be a killer



Wavelet-based multi-resolution particle methods (Bergdorf & Koumoutsakos, 2006)

At each time-step, wavelet-based MRA of (grid) quantities, based on interpolating wavelets :

$$q(\boldsymbol{x}) = \sum_{\boldsymbol{k}\in\mathcal{K}^0} c_{\boldsymbol{k}}^0 \varphi_{\boldsymbol{k}}^0(\boldsymbol{x}) + \sum_{l=0}^{L-1} \sum_{\boldsymbol{k}\in\mathcal{K}^l} \sum_{\mu=1}^{2^a-1} d_{\boldsymbol{k}}^{l,\mu} \psi_{\boldsymbol{k}}^{l,\mu}(\boldsymbol{x})$$

where d is the dimension and scaling functions and wavelets are recursively given by filter operations

$$\varphi_{\boldsymbol{j}}^{l} = \sum_{\boldsymbol{k}} H_{\boldsymbol{j},\boldsymbol{k}}^{l} \varphi_{\boldsymbol{k}}^{l+1}, \qquad \psi_{\boldsymbol{j}}^{l,\mu} = \sum_{\boldsymbol{k}} G_{\boldsymbol{j},\boldsymbol{k}}^{l,\mu} \varphi_{\boldsymbol{k}}^{l+1}$$



Nested grids and grid adaptation based on thresholding detail coefficient (Liandrat & Tchamitchian, 1990, Vasilyev 2003)

Particle method advects/remeshes scale solution at the successive scales, level by level, then add up results to reconstruct solution and perform MRA for next iteration

Nested grids, for wavelet coefficient above given threshold :

 $\{ \boldsymbol{\mathcal{K}}_{>}^{l} \}_{l=0}^{L}$

An additional buffer is created around particles activated at level l, with values obtained by interpolation form level l-1, to allow consistent remeshing

$$\mathcal{B}^{l} = \left\{ \mathbf{k}' \mid \min_{\mathbf{k} \in \mathcal{K}^{l}_{>}} |\mathbf{k}' - \mathbf{k}| \leq \lceil \frac{1}{2} \operatorname{supp}(\zeta) + \operatorname{LCFL}
ceil
ight\}$$

Finally, like for grid-based methods, need to allow levels l+1 to appear from level 1 during advection



Algorithm for time advancement of particles at a given level \boldsymbol{l}



additional trick : to allow levels l+1 to appear through advection of level l, remesh level l particles on grid l+1 consistent with lagrangian CFL for time-step



Illustration of MRA SL particles for flow around a wind turbine (ETH group of Koumoutsakos)

Ingredients : wavelet-based particles for vorticity transport and

Brinkman penalization for non-slip boundary conditions (Angot et al., 1999, Coquerelle & Cottet, 2008)







Extension to Vlasov-Poisson « should » be straightforward

with (at least) two questions :

1) splitting (x,y,z)/(u,v,w) does not allow fine scales to appear dynamically

- -> each level passively transported
- -> is it a problem ?
- -> should one choose other splitting strategy e.g. (x,u)/(y,v)/(z,w) ?

2) kinetic equations vs fluid models : worth it ?

