



# From particle methods to hybrid semi- Lagrangian schemes

Martin Campos-Pinto<sup>1</sup>, Frédérique Charles<sup>1</sup>

<sup>1</sup>Laboratoire Jacques-Louis Lions  
Sorbonne universite, CNRS UMR 7598

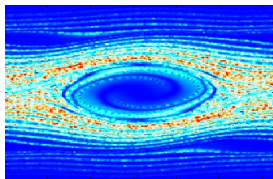
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# Introduction

## Particle methods

Widely used for transport equations, especially for kinetic equations

- Quite easy to implement, even for high dimensions
- Lower computational cost than Eulerian methods (DG, Backward or Forward semi-Lagrangien schemes...)
- **Main drawback : noisy solutions**



## Outline

- 1 Particle methods
- 2 LTP method
- 3 FBL method

# Particle methods

## Transport equation (conservative form)

$$\partial_t \rho + \nabla \cdot (a\rho) = 0, \quad \rho(0, \cdot) = \rho^0 \quad (1)$$

with  $a : \mathbb{R}^+ \times \mathbb{R}^d \rightarrow \mathbb{R}^d$  smooth ( $a_i \in L^\infty(0, T; W^{1,\infty})$ ).

## Principle of particle methods

If  $\rho^0(x) = \delta(x - x_0)$ , then the measure solution of (1) is given by

$$\rho(t, x) = \delta(x - X_{0,x_0}(t))$$

where  $X_{s,x_0}(t) = F^{s,t}(x_0)$  is the characteristic line starting from  $x_0$  :

$$\begin{cases} \frac{d}{dt} X_{s,x_0}(t) &= a(t, X_{s,x_0}(t)) \\ X_{s,x_0}(s) &= x_0 \end{cases}$$

# Particle methods

Consequence: discretization of (1)

- We choose  $(\omega_k^0, x_k^0)_{1 \leq k \leq N}$  such that  $\rho^0(x) \sim \rho_h^0 = \sum_{k=1}^N \omega_k^0 \delta(x - x_k^0)$

For example (deterministic initialization)

$$x_k^0 = h\mathbf{k} \quad (\mathbf{k} \in \mathbb{Z}^d), \quad \omega_k^0 = \rho^0(x_k^0)$$

- Then the solution  $\rho(t, x)$  is approximated by

$$\rho_h(t, x) = \sum_{k=1}^N \omega_k^0 \delta(x - X_k(t)) \quad \text{with} \quad \begin{cases} X_k'(t) = a(t, X_k(t)) \\ x_k(0) = x_k^0 \end{cases}$$

Remark

For equation  $\partial_t \rho + \nabla \cdot (a\rho) + a_0 \rho = 0$  the weights evolve according to

$$\omega_k'(t) + a_0(t, X_k(t))\omega_k(t) = 0, \quad \omega(0) = \omega_k^0$$



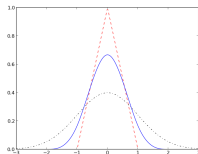
# Regularization

## Convolution kernel

Let  $\varepsilon > 0$  and  $\varphi_\varepsilon$  such that  $\int_{\mathbb{R}^d} \varphi_\varepsilon(x) dx = 1$ ,  $\varphi_\varepsilon \xrightarrow{\varepsilon \rightarrow 0} \delta$ ,  $\varphi_\varepsilon$  even.

Typically, we take  $\varphi_\varepsilon(y) = \frac{1}{\varepsilon^d} \varphi\left(\frac{y}{\varepsilon}\right)$ , with

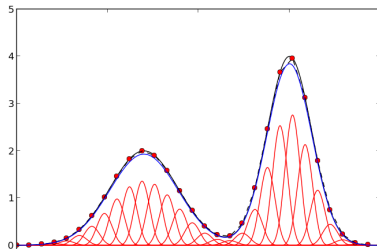
- $\varphi$  a spline function (B1 or B3)
- or  $\varphi$  a truncated Gaussian



## Smooth particle approximation of $\rho$

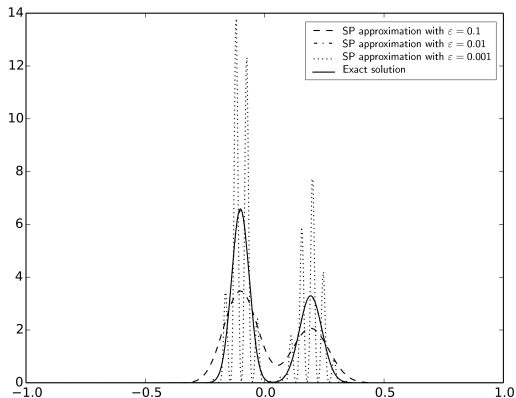
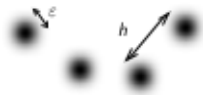
$$\rho_{h,\varepsilon}(t, x) = \sum_{k=1}^N \omega_k \varphi_\varepsilon(x - x_k(t))$$

$$(\omega_k \simeq \int_{x_k^0 - h/2, x_k^0 + h/2} \rho^0(t) dt)$$



# Smooth Particles methods

Choice of  $\varepsilon$  ?



# Convergence of smooth Particles methods

## Error estimate for a given velocity $a$ [Raviart]

If  $\varphi \in \mathcal{W}^{m,1}$  is such that  $\int \varphi = 1$ ,  $\int |x|^r |\varphi(x)| dx < +\infty$  and for a certain  $m > 0$ ,  $r > 0$

$$\int x_1^{k_1} \dots x_d^{k_d} \varphi(x_1, \dots, x_d) dx = 0, \quad |k| = k_1 + \dots + k_d \leq r - 1$$

then the following error estimate hold :

$$\|\rho_{h,\varepsilon} - \rho\|_{L^\infty(0,T) \times \mathbb{R}^d} \leq C_T \left( \varepsilon^r |\rho^0|_{W^{r,\infty}} + \left(\frac{h}{\varepsilon}\right)^m |\rho^0|_{W^{m,\infty}} \right)$$

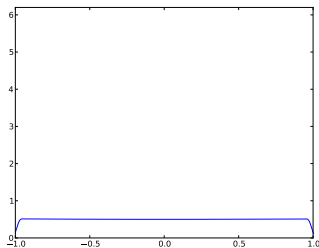
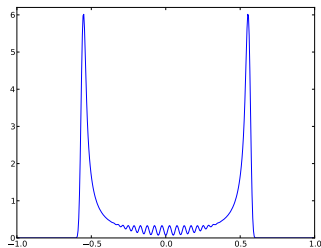
- With  $h \sim \varepsilon$ , **weak** convergence of the density only
- For a given  $\varepsilon$ , **strong** convergence requires  $h \ll \varepsilon$  ( $h \sim \varepsilon^{1/q}$ ,  $q < 1$ )
  - $\Rightarrow$  require a **huge number of particles**  $N$  ( $N \sim \frac{1}{h^d}$ )
  - $\Rightarrow$  implies **extended particle overlapping**

# Exemple of a "noisy" Smooth Particle simulation

Numerical resolution with a SP method of the 1D-aggregation equation

$$\begin{cases} \partial_t \rho + \frac{\partial}{\partial x} (\rho u) = 0 \\ u(t, x) = -W' \star \rho \\ \rho(0, x) = \rho^0(x) \end{cases} \quad (2)$$

where  $W(x) = \frac{|x|^a}{a} - \frac{|x|^b}{b}$ ,  $(a, b) = (4, 2.5)$ , with  $N = 100$  particles,  $\varepsilon = h$ .


 $\rho_{h,\varepsilon}^0$ 

 $\rho_{h,\varepsilon}^n$  at steady state, with SP method

# Denoising particle methods

## Periodical remapping with $\varepsilon \sim h$

- re-initialize the particles on a phase-space grid.
- new particles = regular nodes

$$\rho_h^n(x) = \sum_k \omega_k \varphi_h(x - X_k^n) \quad \longrightarrow \quad \rho_h^{n, \text{remap}}(x) = A_h(\rho_h^n) = \sum_k \omega_k^{\text{remap}} \varphi_h(x - x_k^0)$$

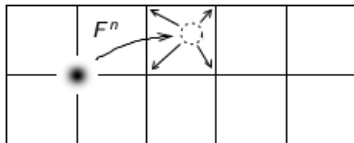
with  $\omega_k^{\text{remap}} = a_{k,h}(\rho_h^n)$  depends of values of  $\rho_h^n$  on a local stencil around  $x_k^0$ .

$\Rightarrow$  oscillations are smoothed out

... but **numerical diffusion** is introduced

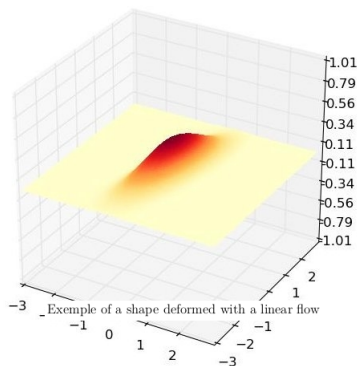
## Same idea : Remeshed Particle Method/ Forward Semi-Lagrangien schemes

- Crouseilles, Respaud, Sonnendrücker 2008
- Cottet, Etancelin, Perignon, Picard, 2014 ; Cottet 2018



# LTP (LTPIC) method

- Original idea of Perthame and Cohen (2000) : transform the shape  $\varphi_\varepsilon$  of particles in order to better follow the flow
- LTP (and QTP) method developed in the context of the Vlasov-Poisson system (divergence free flow) by M. Campos-Pinto (JSC 2014, JCP 2014).



# Idea of LTP method

- The solution of equation (1) is given by  $\rho(t) = F_{\text{ex}}^{0,t} \# \rho^0$ , i.e :

$$\rho(t, x) = j^{t,0}(x) \rho^0(F_{\text{ex}}^{t,0}(x))$$

with

$$\rho^0 \sim \rho_h^0 = \sum_k \omega_k \varphi_{h,k}^0, \quad \varphi_{h,k}^0(x) = \frac{1}{h^d} \varphi\left(\frac{x - x_k^0}{h}\right)$$

- Introduce a linearization of the forward flow around  $x_k^0$

$$F_{\text{lin},k}^{0,t} : x \rightarrow x_k(t) + \bar{J}_k^t(x - x_k^0), \quad \bar{J}_k^t = J^{0,t}(x_k^0)$$

and transform the shape of each particle according to the linearized flow

$$\varphi_{h,k}^t = F_{\text{lin},k}^{0,t} \# \varphi_h^0$$

Then

$$\varphi_{h,k}^t(y) = \frac{1}{\det(\bar{J}_k^t)} \varphi_h\left(\underbrace{(\bar{J}_k^t)^{-1}}_{\text{deformation}}(y - x_k(t))\right)$$

LTP representation of  $\rho(t^n, \cdot)$ 

$$\rho_h^n(x) = \sum_k \omega_k \varphi_{h,k}^n(x) \quad \text{with} \quad \varphi_{h,k}^n(x) = \frac{1}{h_k^n} \varphi \left( \frac{D_k^n}{h} (x - x_k^n) \right)$$

- $D_k^n$  : "Deformation matrix";  $D_k^0 = I_{d \times d}$
- $h_k^n$  : "volume of particles";  $h_k^0 = h^d$

## Computation of the approximated Jacobian matrix

- Compute  $D_k^n = D_k^{n-1} (J_k^{n-1})^{-1}$ , where  $J_k^{n-1} \simeq J^{t^{n-1}, t^n}(x_k^{n-1})$ ,

$$\text{using for example} \quad \frac{J^{s,t}}{dt}(x) = du(t, F^{s,t}(x)) J^{s,t}(x)$$

- or directe FD using current positions of particles approximation

$$D_k^n = \left[ \left( \frac{x_{k+e_j}^n - x_{k-e_j}^n}{2h} \right)_{1 \leq i, j \leq d} \right]^{-1},$$



# Passiv flow

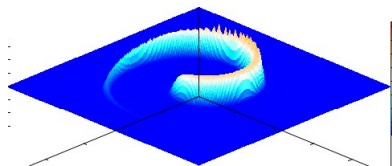
Theorem [M. Campos-Pinto, 2014]

If  $\rho^0 \in W^{1,\infty}(\mathbb{R}^d)$ ,  $a \in L^\infty([0, T], W^{2,\infty})$  and  $\operatorname{div}(a) = 0$ , with  $x_k^n = F_{\text{ex}}^{0,t_n}(x_k^0)$  then

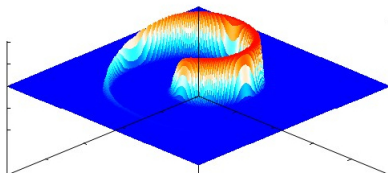
$$\|\rho_h^n - \rho_{\text{ex}}^n\|_\infty \leq Ch \quad (C = C(T, \|\rho^0\|_{W^{1,\infty}},))$$

Example : reversible "swirling" velocity field (LeVeque)

$$u_{SW,T}(t, x) = \cos\left(\frac{\pi t}{T}\right) \operatorname{curl}(\phi_{SW}(x)), \quad \phi_{SW}(x) = \frac{-1}{\pi} \sin^2(\pi x_1) \sin^2(\pi x_2)$$



SP solution



LTP solution

$$N_{\text{part}} = 4 \cdot 10^4, \quad t = T/2, \quad \Delta_{\text{remap}} = 10\Delta t$$

# Passiv flow : idea of the proof

- Approximation operator :  $A_h : \rho^0 \mapsto \rho_h^0(x) = \sum_k \omega_k \varphi_{h,k}^0(x)$  such that

$$\|A_h \rho^0 - \rho^0\|_{L^\infty} \leq Ch |\rho^0|_{W^{1,\infty}}$$

- Let denote  $S_{h,k}^0 = \text{Supp}(\varphi_{h,k}^0)$  and

$$\Sigma_{h,k}^n = F_{\text{ex}}^{0,t_n}(S_{h,k}^0) \cup F_{h,k}^{0,t_n}(S_{h,k}^0)$$

- Overlapping ( $\mathcal{K}_n(x)$  : set of overlapping particles at location  $x$ ) :

$$\kappa_n := \sup_{x \in \mathbb{R}^d} \#\mathcal{K}_n(x), \quad \mathcal{K}_n(x) := \{k \in \mathbb{Z}^d, x \in \Sigma_{h,k}^n\}.$$

- Error on the Forward flow :

$$\bar{e}_F^n = \sup_k \|F_{\text{ex}}^{0,t_n} - F_{h,k}^{0,t_n}\|_{L^\infty(S_{h,k}^0)}$$

# Vlasov-Poisson system

$$\frac{\partial f}{\partial t} + v \cdot \partial_x f + E \cdot \partial_v f = 0, \quad \partial_x E(t, x) = 1 - \int_{\mathbb{R}} f(t, x, v) dv$$

$f(t, x, v)$  : density function in electron,  $t \in [0, T]$

## Hypothesis

- $E$  given by

$$E(t, x) = \int_0^L K(x, y) \left( 1 - \int_{\mathbb{R}} f(t, y, v) dv \right) dy,$$

- bounded support in  $v$ -dimension,  $L$ -periodic with respect to  $x$
- $f^0 \in \mathcal{W}^{2, \infty}(\mathbb{R}^2)$
- global neutrality relation

$$\int_0^L \left( \int_{\mathbb{R}} f^0(x, v) dv - 1 \right) dx = 0.$$

# Vlasov-Poisson system

Theorem [GH Cottet, PA Raviart, 1984]

Using SP method, one has

$$\sup_{t \in [0, T]} \|E_h(t) - E(t)\|_{L^\infty} + \sup_k \|Z_k^h(t) - F_{\text{ex}}^{0,t}(z_k^0)\| \leq C_T h$$

Theorem [M. Campos-Pinto, F.C, 2014]

Using LTP method, one has

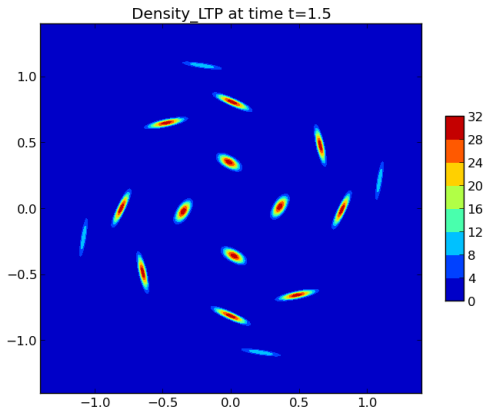
$$\sup_{0 \leq n \leq T/\Delta t} \|E_h^n + 1/2 - E(t_{n+1/2})\|_{L^\infty} + \sup_k \|z_k^n - F_{\text{ex}}^{0,t_n}(z_k^0)\| \leq C_T (h^2 + \Delta t^2)$$

and provided  $\Delta t \lesssim \sqrt{h}$ , the particle approximation of the phase space density satisfies

$$\|f_h^n - f_{\text{ex}}^n\|_{L^\infty(\mathbb{R}^2)} \lesssim h + \frac{\Delta t^2}{h}$$

# Loss of locality in LTP method

Evolution of the shapes of particles



⇒ necessity of periodical remappings

# FBL method

We consider here the equation

$$\partial_t \rho + a \cdot \nabla \rho = 0, \quad \rho(0, \cdot) = \rho^0 \quad (3)$$

The exact solution is given by

$$\rho(t, x) = \rho^0 (F^{t,0}(x)) \quad (4)$$

## Idea

Use a approximation of the backward flow  $F^{t,0}(x)$  (for all  $x$ ) directly in (4) !!

# FBL method

- "Particles" are pushed forward like in Particles methods :  $(x_k^n)_k$
- Reconstruct of the backward flow at time  $t_n$  on a grid  $(\xi_i)_{i \in \mathbb{Z}^d}$  :

$$F^{t_n, 0}(\xi_i) \simeq B_i^n := F_{h, k^*(n, i)}^{t_n, 0}$$

where

$$k^*(n, i) := \operatorname{argmin}_k \|x_k^n - \xi_i\|$$

and

$$F_{h, k^*(n, i)}^{t_n, 0} : x \mapsto x_k^0 + D_k^n(x - x_k^n), \quad D_k^n \simeq (J_{F_{\text{ex}}^{0, t_n}}(x_k^0))^{-1}$$

- (optional) Reconstruct a backward flow for all  $x$  with a partition of unity  $\sum_i S_h(x - \xi_i) = 1$ , and

$$B_h^n(x) = \sum_i B_i^n(x) S_h(x - \xi_i)$$

- Define a **FBL approximation** of  $\rho^n$  by

$$\rho_h^n(x) = \rho^0(B_h^n(x))$$

# FBL method for a passiv flow

Theorem [M. Campos-Pinto, F.C, 2016]

The FBL approximation of order 1 (LFBL) satisfy

$$\|\rho_h^n - \rho_{\text{ex}}^n\|_\infty \leq Ch^2$$

Key argument

LTP estimate is based on the bound

$$|\omega_k| \|\varphi_h(F_{h,k}^{t_n,0}(x)) - \varphi_h(F_{\text{ex}}^{t_n,0}(x))\| \leq |\omega_k| \|\varphi_h\|_{\text{lip}} \|F_{h,k}^{t_n,0} - F_{\text{ex}}^{t_n,0}\|_{L^\infty(S_{h,k}^n)} \lesssim \frac{1}{h} \bar{e}_F^n$$

whereas for FBL we have

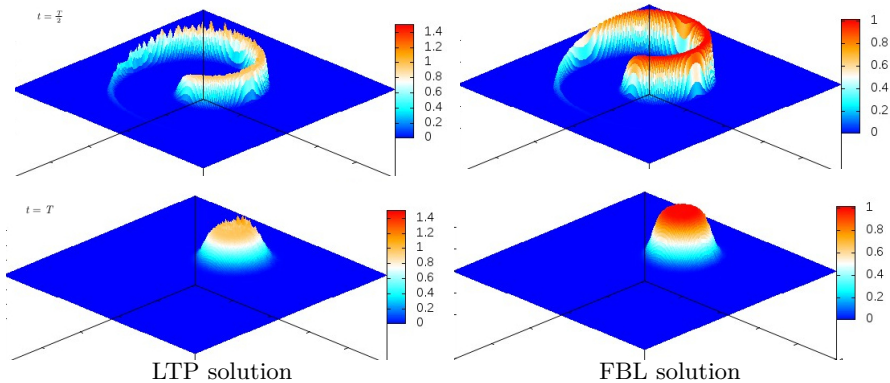
$$\|\rho_0(B_h^n(x)) - \rho_0(F_{\text{ex}}^{t_n,0}(x))\| \lesssim \|\rho^0\|_{\text{lip}} \bar{e}_F^n$$



# FBL method

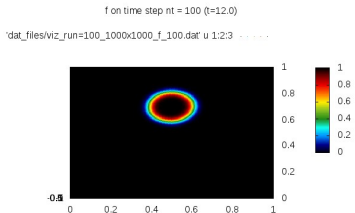
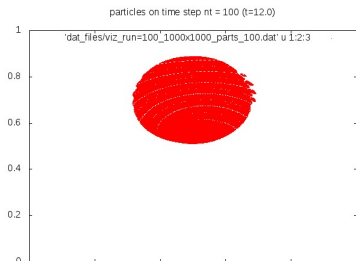
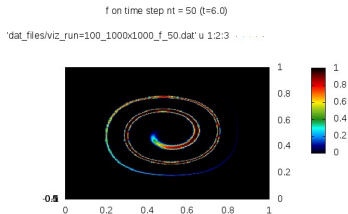
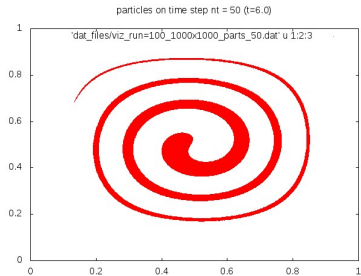
Example : reversible "swirling" velocity field  $u_{SW,T}$

$N_{part} = 4 \cdot 10^4$ ,  $T = 5$ ,  $\Delta t = 0.05$ ,  $\Delta t_{remap} = 1.5$



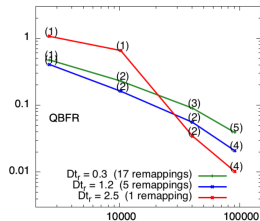
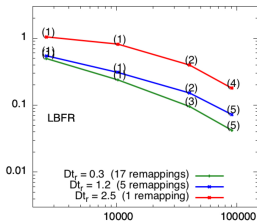
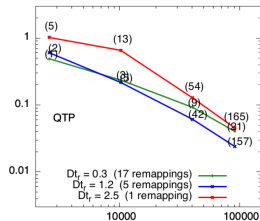
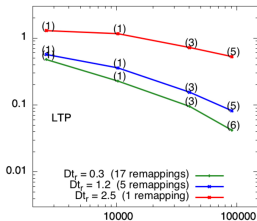
# FBL method

$$N_{part} = 10^6, T = 12, \Delta t = 0.12, \Delta t_{remap} = 2$$



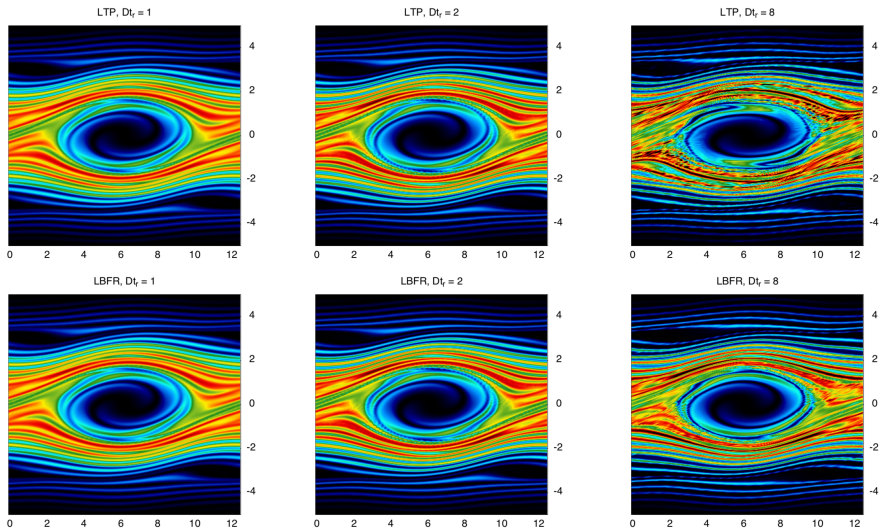
# FBL method

Error estimates for the reversible "swirling" velocity field  $u_{SW,T}$



# FBL for Vlasov-Poisson system

Two stream instability  $f_0(x, v) = (1 + \epsilon \cos(kx)) v^2 \frac{1}{\sqrt{2\pi}} e^{-v^2/2}$ ,  $\epsilon = 0.5$ ,  $T = 45$ .



## Conclusion

- LTP methods improve accuracy ... but still need remappings
- QTP even better ... but very costly
- FBL methods: **enhanced locality**, **less remappings** are needed

## Next prospects

- Multiscale method
- Extension to diffusion equation and Vlasov Fokker-Planck
  - ▶ linear Fokker Planck-Planck-Landau operator
 
$$PF(f)(v) = \nabla_v \cdot (\nabla_v f(v) + vf(v))$$
  - ▶ non linear kernel  $L_\phi(f)(v) = \nabla_v \cdot ((a_\phi \star f)\nabla_v f(v) - (b_\phi \star f)f(v))$

## References

- M. Campos-Pinto, Towards smooth particles methods without smoothing, JSC 2014.
- M. Campos-Pinto, E. Sonnendrücker, A. Friedman, S. Lund, D. Grote, Noiseless Vlasov-Poisson simulations with linearly transformed particles, JCP 2014.
- M. Campos-Pinto, FC, Uniform convergence of a linearly transformed particle method for the Vlasov-Poisson system, SINUM 2016.
- M. Campos-Pinto, J. Carrillo, FC, Y.-P. Choi Convergence of a linearly transformed particle method for aggregation equations, Num. Math., 2018.
- M. Campos-Pinto, FC, From particle methods to hybrid semi-Lagrangian schemes. SMAI-JCM, 2018.

Thank you for your attention!