

# Meshfree particle model for kinetic plasma simulations

John M. Finn (Tibbar); Evstati G. Evstatiev (SNL\*)

September 28, 2022

\*Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

## Abstract

We revisit a meshfree particle model for kinetics of a 1D electrostatic plasma, using kernel density estimation and a similar method for the electric field  $E$ . The translationally invariant kernel  $K(x - y)$  represents the macroparticle charge distribution. Two length scales enter, the width  $w$  of  $K$  and the interparticle spacing  $l$ . This model conserves momentum and energy. Similarly, continuity is satisfied exactly, and the Gauss's law and Ampere's law formulations are exactly equivalent. A unified analysis is used for numerical stability and noise properties. The force can be computed directly using the convolution  $K_2 = K * K$ , and  $K_2$  is positive definite. We discuss the analogy in the presence of a grid. We can specify a single kernel  $K_2$ , related to the 'kernel trick' of machine learning. Numerical instability can occur unless  $K_2$  is positive definite, related to a breakdown in energy conservation. For the noise analysis, the covariance matrix for the electric field shows a plasma dispersion function modified by  $w$  and  $l$ . The number of particles per cell does not enter, and the noise is characterized by the number of particles per kernel width, i.e.  $w/l$ . We present the bias-variance optimization (BVO) for the electric field, and compare it to the density BVO.

## Outline

- Meshfree Vlasov-Gauss and Vlasov-Ampère formulations
- Moment equations; conservation properties
- ‘Kernel trick’ – preventing instabilities and ensuring energy conservation
- Linearized equations for numerical stability and noise response for a cold plasma
- Linearized equations for numerical stability and noise response for a warm plasma
- Bias-variance optimization for the electric field

## The model

One-dimensional electrostatic particle method, immobile ions.

Periodic boundary conditions  $0 \leq x \leq 1$

Kernel density estimation to obtain the electron density

The kernel (particle shape)  $K(x - y)$  is the local charge density within the macroparticle

As in E. G. Evstatiev, J. M. Finn, B. A. Shadwick, N. Hengartner, “Noise and error analysis and optimization in particle-based kinetic plasma simulations”, JCP 440, 110394 (2021).

Related kernel-like method to obtain the estimated electric field  $E(x)$  and the electrostatic potential  $\phi(x)$

The force on a macroparticle: integrate  $F(x) = \int E(y)K(y - x)dy$   
*Same kernel* for particle positions  $\rightarrow E(x)$  and for  $E(x) \rightarrow F(x)$ .

Source and target – charge distribution same. Leads to  $K * K$  and a positive definite kernel

## The model

$$f(x, v, t) = \sum_{\alpha=1}^N q_\alpha K(x - \xi_\alpha(t)) \delta(v - v_\alpha(t))$$

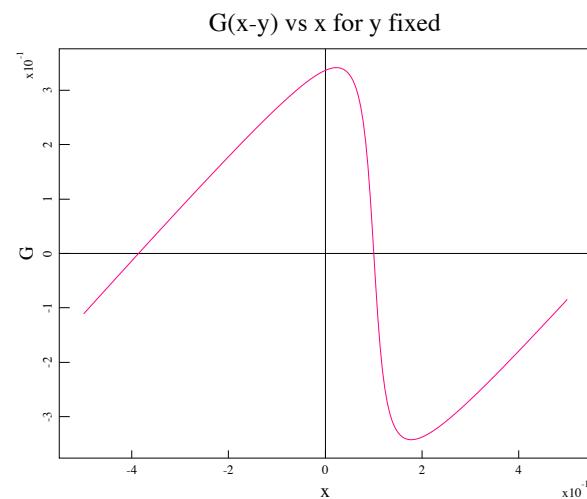
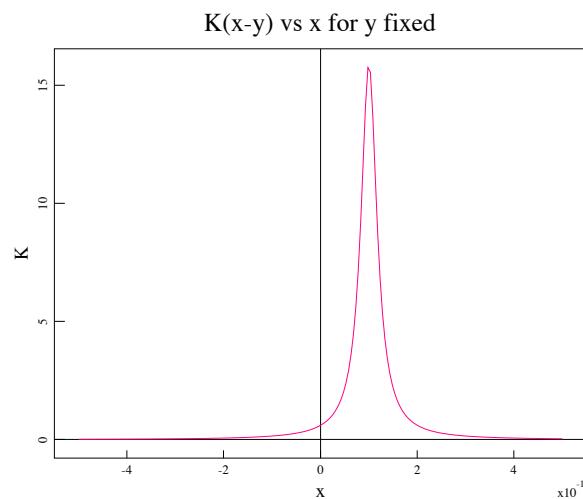
$$K(x) = \frac{1}{w} K_f \left( \frac{x}{w} \right) \quad K_{per}(x) = \sum_{n=-\infty}^{\infty} K(x - n)$$

$$\rho(x, t) = \sum_{\alpha} q_\alpha K(x - \xi_\alpha(t))$$

$$E(x, t) = \sum_{\alpha} q_\alpha G(x - \xi_\alpha(t)); \quad G'(x) = 1 - K(x)$$

$$\phi(x, t) = \sum_{\alpha} q_\alpha \Phi(x - \xi_\alpha(t)); \quad \Phi'(x) = -G(x)$$

# The kernel $K(x - y)$ and $G(x - y)$



## The model

$G'(x) = 1 - K(x)$  (Gauss);  $\int_0^1 G(x)dx = 0$ . (zero potential difference.)

$q_\alpha G(x - \xi_\alpha)$ ...electric field due to single (macro)particle at  $\xi_\alpha$ .  
Length scales:  $w$ ,  $\lambda = 1/N$ . Ratio  $w/\lambda = Nw$  – particles per kernel width.

$$\frac{d\xi_\alpha}{dt} = v_\alpha$$

$$\frac{dv_\alpha}{dt} = -F(\xi_\alpha, t), \quad F(x, t) = \int E(y)K(y - x)dy$$

Same kernel  $K$ .

## Note

It is possible to calculate the force directly by

$$F(x, t) = E(x, t) = \sum_{\alpha} q_{\alpha} G_2(x - \xi_{\alpha}(t))$$

$$G'_2 = 1 - K'_2 \text{ and } K_2(x) = \int K(y)K(x - y)dy.$$

This is a positive definite kernel! More later.

***The kernel trick:*** just choose a positive definite kernel for  $K_2$   
(Then  $K = \sqrt{K_2}$ , but you do not need  $K$ .)

## Meshfree Vlasov-Gauss/Vlasov-Poisson and Vlasov-Ampere

$$\rho(x, t)u(x, t) = \sum_{\alpha} q_{\alpha} K(x - \xi_{\alpha}(t))v_{\alpha}(t) \quad f(v|x, t) = f(x, v, t)/\rho(x, t)$$

and (conditional expectation)  $u(x, t) = \int f(v|x, t)v dv$ .  
 $\partial_t \rho(x, t) + \partial_x (\rho(x, t)u)$

$$= - \sum_{\alpha} q_{\alpha} K'(x - \xi_{\alpha}(t))\dot{\xi}_{\alpha}(t) + \sum_{\alpha} q_{\alpha} K'(x - \xi_{\alpha}(t))v_{\alpha}(t) = 0$$

The continuity equation is satisfied exactly

$$u(x, t) = \frac{1}{\rho} \sum_{\alpha} q_{\alpha} K(x - \xi_{\alpha}(t))v_{\alpha}(t) = \sum_{\alpha} q_{\alpha} W(x, t)v_{\alpha}(t)$$

Nadaraya-Watson non-parametric regression

$$W(x, \xi_{\alpha}(t), t) = \frac{K(x - \xi_{\alpha}(t))}{\rho(x, t)} = \frac{K(x - \xi_{\alpha}(t))}{\sum_{\beta} q_{\beta} K(x - \xi_{\beta}(t))} \text{ Partition of unity}$$

## Conservation properties

- ▶ Momentum is conserved exactly because, with no mesh, translational invariance is exact
- ▶ Energy is conserved exactly – related to the  $K_2 = K * K$  issue above. (Time step  $h \rightarrow 0$ )
- ▶ Vlasov-Ampère  $\partial_t E = -j = \rho u$ . V-A and V-G (V-P) are equivalent because the continuity equation holds exactly
- ▶ The model consists of  $N$  macroparticles with charge distributed according to  $K$ , with the force computed from  $K_2$

## Linearized equations, cold plasma – numerical stability

Introduce a lattice *just for the linearized calculations*

$$x_\alpha = (\alpha - 1)\Delta, \xi_\alpha = x_\alpha + \delta\xi_\alpha$$

$$\tilde{E}(x, t) = -\Delta \sum_{\alpha} G'(x - x_{\alpha}) \delta\xi_{\alpha}(t) = \Delta \sum_{\beta} (K(x - x_{\beta}) - 1) \delta\xi_{\beta}(t)$$

$$\ddot{\delta\xi_{\alpha}} = -\omega^2 \delta\xi_{\alpha} = \tilde{F}(\xi_{\alpha}(t), t) = -\Delta \sum_{\beta} (K_2(x_{\alpha} - x_{\beta}) \cancel{- 1}) \delta\xi_{\beta}(t)$$

$$\omega_k^2 = \hat{K}_f(kw), \quad \omega_k^2 \rightarrow_{w \rightarrow 0} \hat{K}_f(0) = 1 (= \omega_{pe}^2)$$

$\omega_k^2 > 0$  if  $K(x - y)$  is positive definite. If using the kernel trick (choosing  $K_2$ ), **be sure that  $K_2$  is positive definite.**

## Positive definite kernels

Energy conservation also requires  $K_2$  to be positive definite.

If  $B(x)$  is the boxcar kernel,  $B * B =$  tent or linear is pos def.  
 $B * B * B =$  quadratic spline is *not*.

This numerical instability seems to be the first connection  
between kernels for density estimation and positive definite  
(reproducing) kernels.

## Linearized equations, cold plasma – noise

$$\langle \hat{E}(k_1, \omega_1)^* \hat{E}(k_2, \omega_2) \rangle = \frac{\sigma^2 I_c}{2\pi} \frac{\hat{K}(k_1)}{\hat{K}_2(k_1) - \omega_1^2} \frac{\hat{K}(k_1)}{\hat{K}_2(k_1) - \omega_2^2} (1 + \omega_1 \omega_2) \delta(k_1 - k_2)$$

$1/(\hat{K}_2(k_1) - \omega_1^2)$  propagator for plasma oscillation

- ▶ Includes decay of initial conditions (ballistic term) – terms off  $D(k, \omega) = 0$
- ▶ Shows that a plasma oscillation ( $D(k, \omega) \approx 0$ ) persists after the ballistic term decays

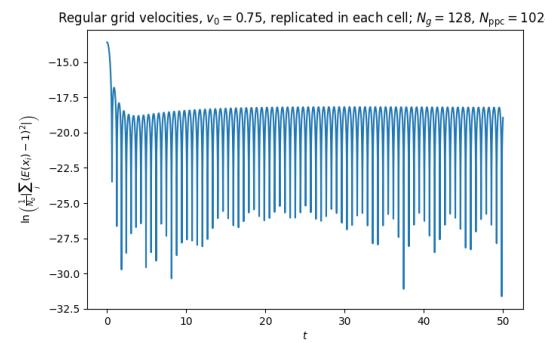
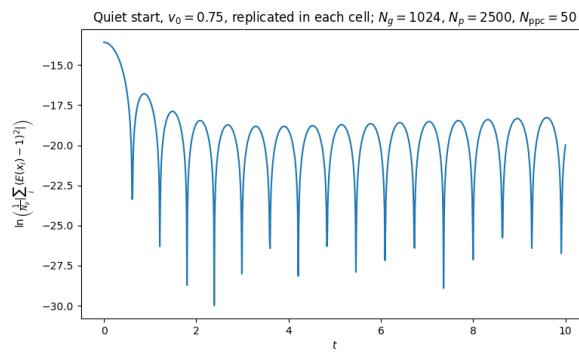
## Warm plasma

Linear stability for  $f_0(v) = \Theta(v_0^2 - v^2)/2v_0$ : Langmuir wave (without Landau damping) has

$$\omega^2 = \hat{K}_2(k) + k^2 v_0^2$$

Thermal term ameliorates the instability if the kernel trick is used and  $K_2$  is not positive definite

$\ln \langle \hat{E}(k_1, \omega_1)^* \hat{E}(k_2, \omega_2) \rangle$ : propagator for the Langmuir wave  
 $1/(\hat{K}_2(k_1) - \omega_1^2 - k^2 v_0^2)$  w/o Landau damping. Ballistic term damps, leaving (non-Landau damped) Langmuir waves.



## Bias-variance optimization for $E(x, t)$ and $F(x, t)$

$\partial_x E = 1 - \rho$ , so  $E(x)$  is smoother – variance should be lower.

BVO for the density was computed in E. G. Evstatiev, J. M. Finn, B. A. Shadwick, N. Hengartner, “Noise and error analysis and optimization in particle-based kinetic plasma simulations”, JCP 440, 110394 (2021). Density: diagnostic.  $F_e(x) = \sum_{\alpha} q_{\alpha} G_2(x - \xi_{\alpha})$  enters

$$\text{Bias: } \langle F_e(x) \rangle = F(x) + B(x) + O(w^4),$$

$$B(x) = \frac{C_2 w^2}{2} F''(x), \quad C_2 = \int \zeta^2 K_f(\zeta) d\zeta$$

Error is  $Q = B^2 + V$

$$Q(w) \approx \frac{\rho(x)}{12N_p} - \frac{2w\rho(x)C_3}{N_p} + \frac{C_2^2 w^4}{4} E''(x)^2$$

## Bias-variance optimization

(MISE)

$$Q_2(w) = \frac{1}{12N_p} - \frac{2wC_3}{N_p} + \frac{C_2^2 w^4}{4} R$$

Minimize:

$$w = \left( \frac{2C_3}{C_2^2 R N_p} \right)^{1/3}, \quad Q_{2,min} = \frac{1}{12N_p} - \frac{3}{4(C_2^2 R)^{1/3}} \left( \frac{2C_3}{N_p} \right)^{4/3}.$$

$$w_{min} \sim N_p^{-1/5} \text{ (density); } \quad w_{min} \sim N_p^{-1/3} \text{ (force)}$$

For non-compact support kernels (e.g. Gaussian), computation time scales as  $N^2$ . With compact support, like  $N(Nw_{min}) \sim N^{5/3}$  (density:  $\sim N^{9/5}$  – worse)

## Conclusions

- ▶ Particles per cell does not enter (meshfree.) *Particles per kernel width* replaces it. Comparison with a PIC code with  $\Delta \rightarrow 0$  gives agreement.
- ▶ Exact energy conservation occurs for  $h \rightarrow 0$ : for finite  $h$ , use a symplectic integrator like leapfrog (symplectic Euler) – good energy conservation properties.
- ▶ With  $K$  in source of  $E(x)$  and in integrating  $E(x)$  over target particle, the same  $K$  should be used and  $K_2 = K * K$  can be used to compute  $F(x, t)$  directly, without computing  $E(x)$ . The *kernel trick* means that a positive definite kernel can be used in place of  $K_2$ .
- ▶ If a non-positive definite kernel is used in the kernel trick, it can give numerical stability as well as lack of energy conservation.
- ▶ BVO: If the optimal  $w_{min}$  is chosen, the computation time scales as  $N^{5/3}$  – better than  $N^2$  but not  $N \log N$ .