Asymptotic preserving scheme for Lévy Fokker Planck equation with fractional diffusion limit

Li Wang

School of Mathematics University of Minnesota

Kinetic equations: from modeling computation to analysis

Centre International de Rencontres Mathématiques

March 22-26, 2021

Joint with Wuzhe Xu

Outline



2 Numerical methods

3 Numerical examples



Outline



2 Numerical methods

3 Numerical examples



Introduction

Lévy Fokker Planck equation

$$\begin{cases} \partial_t f + v \cdot \nabla_x f = \underbrace{\nabla_v \cdot (vf) - (-\Delta_v)^s f}_{:=\mathcal{L}^s(f)}, \quad s \in (0,1), \\ f(0,x,v) = f_{in}(x,v), \end{cases}$$

• fractional Laplacian:

$$(-\Delta_{v})^{s} f(v) := C_{s,d} \text{P.V.} \int_{\mathbb{R}^{d}} \frac{f(v) - f(w)}{|v - w|^{d + 2s}} \mathrm{d}w, \qquad C_{s,d} = \frac{4^{s} \Gamma(d/2 + s)}{\pi^{d/2} |\Gamma(-s)|} (-\Delta_{v})^{s} f(v) := \mathcal{F}\left[|k|^{2s} \mathcal{F}[f](k)\right](v)$$

- s = 1 reduces to the classical Fokker-Planck operator
- application: nonlocal effect in plasma turbulence

< ロ > (四) (四) (三) (三) (

Properties of \mathcal{L}^s

- conservation of mass
- equilibrium

$$\mathcal{L}^{s}(\mathcal{M}) = 0, \quad \int_{\mathbb{R}^{d}} \mathcal{M}(v) dv = 1, \quad \mathcal{M}(v) \sim \frac{C_{0}}{|v|^{d+2s}} \text{ as } |v| \to \infty.$$

• entropy dissipation: $\partial_t f = \mathcal{L}^s f^{-1}$

$$G_{\mathcal{M}}^{\Phi}(f) := \int_{\mathbb{R}^d} \Phi(f) \mathcal{M} dv - \Phi\left(\int_{\mathbb{R}^d} f \mathcal{M} dv\right)$$
$$E_{\Phi}(t) = G_{\mathcal{M}}^{\Phi}\left(\frac{f}{\mathcal{M}}\right), \qquad E_{\Phi}(t) \le e^{-\frac{t}{C}} E_{\Phi}(0), \quad t \ge 0$$

¹Gentil, Imbert, Asymptotic Analysis 2008.

イロト イロト イヨト イヨト 三日

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v) \,. \end{cases} \end{split}$$

Q: why anomalous scaling?

$$\begin{split} \varepsilon^{2}\partial_{t}f^{\varepsilon} + \varepsilon v \cdot \nabla_{x}f^{\varepsilon} &= \nabla_{v} \cdot (vf^{\varepsilon}) + \Delta_{v}f^{\varepsilon} \\ \varepsilon \to 0 \Longrightarrow f \to \rho e^{-\frac{\|v\|^{2}}{2}} \\ \int_{\mathbb{R}^{N}} dv \Longrightarrow \partial_{t}\rho + \frac{1}{\varepsilon}\nabla_{x} \cdot j &= 0 \\ \int_{\mathbb{R}^{N}} v \cdot dv \Longrightarrow \varepsilon \partial_{t}j + \nabla_{x} \cdot \int_{\mathbb{R}^{N}} v \otimes vfdv &= -\frac{1}{\varepsilon}j \\ \Longrightarrow \quad \partial_{t}\rho + \nabla_{x} \cdot (D\nabla_{x}\rho) &= 0, \quad D = \int v \otimes ve^{-\frac{\|v\|^{2}}{2}} dv \end{split}$$

(日) (문) (문) (문) (문)

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v) \,. \end{cases} \end{split}$$

 $\mathbf{Q}:$ why anomalous scaling?

$$\begin{split} \varepsilon^{2}\partial_{t}f^{\varepsilon} + \varepsilon v \cdot \nabla_{x}f^{\varepsilon} &= \nabla_{v} \cdot (vf^{\varepsilon}) + \Delta_{v}f^{\varepsilon} \\ \varepsilon \to 0 \Longrightarrow f \to \rho e^{-\frac{\|v\|^{2}}{2}} \\ \int_{\mathbb{R}^{N}} dv \Longrightarrow \partial_{t}\rho + \frac{1}{\varepsilon}\nabla_{x} \cdot j &= 0 \\ \int_{\mathbb{R}^{N}} v \cdot dv \Longrightarrow \varepsilon \partial_{t}j + \nabla_{x} \cdot \int_{\mathbb{R}^{N}} v \otimes vfdv &= -\frac{1}{\varepsilon}j \\ \Longrightarrow \quad \partial_{t}\rho + \nabla_{x} \cdot (D\nabla_{x}\rho) &= 0, \quad D = \int v \otimes ve^{-\frac{\|v\|^{2}}{2}}dv \end{split}$$

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v) \,. \end{cases} \end{split}$$

Q: why anomalous scaling?

$$\begin{split} \varepsilon^{2}\partial_{t}f^{\varepsilon} + \varepsilon v \cdot \nabla_{x}f^{\varepsilon} &= \nabla_{v} \cdot (vf^{\varepsilon}) + \Delta_{v}f^{\varepsilon} \\ \varepsilon \to 0 \Longrightarrow f \to \rho e^{-\frac{|v|^{2}}{2}} \\ \int_{\mathbb{R}^{N}} dv \Longrightarrow \partial_{t}\rho + \frac{1}{\varepsilon}\nabla_{x} \cdot j &= 0 \\ \int_{\mathbb{R}^{N}} v \cdot dv \Longrightarrow \varepsilon \partial_{t}j + \nabla_{x} \cdot \int_{\mathbb{R}^{N}} v \otimes vfdv &= -\frac{1}{\varepsilon}j \\ \Longrightarrow \quad \partial_{t}\rho + \nabla_{x} \cdot (D\nabla_{x}\rho) &= 0, \quad D = \int v \otimes ve^{-\frac{|v|^{2}}{2}}dv \end{split}$$

Li Wang (Minnesota)

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^\varepsilon + \varepsilon v \cdot \nabla_x f^\varepsilon = \nabla_v \cdot (v f^\varepsilon) - (-\Delta_v)^s f^\varepsilon, \\ f^\varepsilon(0, x, v) = f_{in}(x, v) \,. \end{cases} \end{split}$$

Theorem (Cesbron, Mellet, Trivisa 2012)

Assume that $f_0 \in L^2(\mathbb{R}^N, \mathcal{M}(v)^{-1} dv dx)$, where $\mathcal{M}(v)$ is the unique normalized equilibrium distribution. Then, up to a subsequence, the solution f^{ε} converges weakly in $L^{\infty}(0,T; L^2(\mathbb{R}^{2d}, \mathcal{M}(v)^{-1} dv dx))$ to $\rho(t, x)\mathcal{M}(v)$ as $\varepsilon \to 0$, where $\rho(t, x)$ solves

$$\begin{cases} \partial_t \rho + (-\Delta_x)^s \rho = 0, \\ \rho(0, x) = \rho_{in}(x) := \int_{\mathbb{R}^d} f_{in}(x, v) \mathrm{d}v. \end{cases}$$

《曰》 《圖》 《臣》 《臣》 三臣

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v). \end{cases} \\ \bullet \ f^{\varepsilon}(t, x, v) \rightharpoonup \rho(t, x) \mathcal{M}(v) \quad \text{weakly in } L^2_{\mathcal{M}^{-1}}(\mathbb{R}^N \times (0, \infty)), \qquad \rho = \int f \mathrm{d}v \end{split}$$

• consider a test function $\phi(t, x, v) = \varphi(t, x + \varepsilon v)$

$$\int f^{\varepsilon} \left[\varepsilon^{2s} \partial_t \phi + \varepsilon v \cdot \nabla_x \phi - v \cdot \nabla_v \phi + (-\Delta_v)^s \phi \right] dx dv dx + \varepsilon^{2s} \int f_0(x, v) \phi(0, x, v) dx dv = 0$$

<ロ> (四) (四) (三) (三) (三) (三)

 $\begin{aligned} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v). \end{cases} \end{aligned}$

• $f^{\varepsilon}(t, x, v) \rightharpoonup \rho(t, x)\mathcal{M}(v)$ weakly in $L^{2}_{\mathcal{M}^{-1}}(\mathbb{R}^{N} \times (0, \infty)), \qquad \rho = \int f dv$

 $\bullet \ {\rm consider} \ {\rm a} \ {\rm test} \ {\rm function} \ \phi(t,x,v) = \varphi(t,x+\varepsilon v)$

$$\int f^{\varepsilon} \left[\varepsilon^{2s} \partial_t \phi + \varepsilon v \cdot \nabla_x \phi - v \cdot \nabla_v \phi + (-\Delta_v)^s \phi \right] dx dv dt + \varepsilon^{2s} \int f_0(x, v) \phi(0, x, v) dx dv = 0$$

◆ロ → ◆母 → ◆臣 → ◆臣 → ○ ● ○ ○ ○ ○

$$\begin{split} x \mapsto \varepsilon x, \quad t \mapsto \varepsilon^{2s} t \\ \begin{cases} \varepsilon^{2s} \partial_t f^{\varepsilon} + \varepsilon v \cdot \nabla_x f^{\varepsilon} = \nabla_v \cdot (v f^{\varepsilon}) - (-\Delta_v)^s f^{\varepsilon}, \\ f^{\varepsilon}(0, x, v) = f_{in}(x, v). \end{cases} \end{split}$$

• $f^{\varepsilon}(t, x, v) \rightarrow \rho(t, x)\mathcal{M}(v)$ weakly in $L^{2}_{\mathcal{M}^{-1}}(\mathbb{R}^{N} \times (0, \infty)), \qquad \rho = \int f dv$

 $\bullet \ {\rm consider} \ {\rm a} \ {\rm test} \ {\rm function} \ \phi(t,x,v) = \varphi(t,x+\varepsilon v)$

Outline



2 Numerical methods

3 Numerical examples



æ

< ロ > (四) (四) (三) (三) (

Numerical challenges

$$\begin{cases} \varepsilon^{2s} \partial_t f + \varepsilon v \cdot \nabla_x f = \nabla_v \cdot (vf) - (-\Delta_v)^s f \\ f(0, x, v) = f_{in}(x, v) \end{cases}$$

$$\begin{cases} \partial_t \rho + (-\Delta_x)^s \rho = 0\\ \rho(0, x) = \int f_{in}(x, v) dv \end{cases}$$

・ロト ・ 日 ・ ・ ヨ ・ ・ ヨ ・ ・

• Nonlocality $(-\Delta_v)^s \frac{1}{|v|^{d+2s}}$

• finite difference/finite element + domain truncation

[Huang, Oberman, 2014] [Zhang, Deng, Karniadakis, 2018]

• spectral method

[Mao, Shen, 2017] Hermite polynomial

[Cayama, Cuesta, Hoz, 2019] mapped Chebyshev polynomial

[Sheng, Shen, Tang, Wang, Yuan, 2020] Donford-Taylor formula

review

[Bonito et al., 2018] [Lishke et al., 2020]

Stiffness

[W., Yan, 2016], [Crouseilles, Hivert, Lemou, 2016], [W., Yan, 2019]

Numerical challenges

$$\begin{cases} \varepsilon^{2s} \partial_t f + \varepsilon v \cdot \nabla_x f = \nabla_v \cdot (vf) - (-\Delta_v)^s f \\ f(0, x, v) = f_{in}(x, v) \end{cases}$$

$$\begin{cases} \partial_t \rho + (-\Delta_x)^s \rho = 0\\ \rho(0, x) = \int f_{in}(x, v) dv \end{cases}$$

《曰》 《圖》 《臣》 《臣》 三臣

- Nonlocality $(-\Delta_v)^s \frac{1}{|v|^{d+2s}}$
 - finite difference/finite element + domain truncation
 - [Huang, Oberman, 2014] [Zhang, Deng, Karniadakis, 2018]
 - spectral method

[Mao, Shen, 2017] Hermite polynomial

[Cayama, Cuesta, Hoz, 2019] mapped Chebyshev polynomial

[Sheng, Shen, Tang, Wang, Yuan, 2020] Donford-Taylor formula

• review

[Bonito et al., 2018] [Lishke et al., 2020]

Stiffness

[W., Yan, 2016], [Crouseilles, Hivert, Lemou, 2016], [W., Yan, 2019]

Numerical challenges

$$\begin{cases} \varepsilon^{2s} \partial_t f + \varepsilon v \cdot \nabla_x f = \nabla_v \cdot (vf) - (-\Delta_v)^s f \\ f(0, x, v) = f_{in}(x, v) \end{cases}$$

$$\begin{cases} \partial_t \rho + (-\Delta_x)^s \rho = 0\\ \rho(0, x) = \int f_{in}(x, v) dv \end{cases}$$

イロト スピト スヨト スヨト 三日

- Nonlocality $(-\Delta_v)^s \frac{1}{|v|^{d+2s}}$
 - finite difference/finite element + domain truncation

[Huang, Oberman, 2014] [Zhang, Deng, Karniadakis, 2018]

• spectral method

[Mao, Shen, 2017] Hermite polynomial

[Cayama, Cuesta, Hoz, 2019] mapped Chebyshev polynomial

[Sheng, Shen, Tang, Wang, Yuan, 2020] Donford-Taylor formula

review

[Bonito et al., 2018] [Lishke et al., 2020]

Stiffness

[W., Yan, 2016], [Crouseilles, Hivert, Lemou, 2016], [W., Yan, 2019]

[Cayama, Cuesta, Hoz, 2019]

• mapping $v \mapsto \xi$

$$\xi = \frac{v}{\sqrt{L_v^2 + v^2}} \in (-1, 1) \Longleftrightarrow v = \frac{L_v \xi}{\sqrt{1 - \xi^2}} \in (-\infty, \infty) \,,$$

• Chebyshev polynomials

$$T_k(\xi) = \cos(k \arccos(\xi)), \qquad \xi \in [-1, 1]$$

• further change of variable $q = \arccos(\xi) \in [0, \pi]$

$$T_k(\xi) = \cos(kq)$$

• In the new variable q $(v = \frac{L_v \xi}{\sqrt{1-\xi^2}} = L_v \cot(q))$

$$(-\Delta_q)^s f(q) = \begin{cases} -\frac{1}{L_v \pi} \int_0^{\pi} \frac{f'(p)}{\cot(q) - \cot(p)} dp, & s = \frac{1}{2}, \\ \frac{C_{s,d}}{2L_v^{2s} s(1-2s)} \int_0^{\pi} \frac{\sin^2(p) f''(p) + 2\sin(p) \cos(p) f'(p)}{|\cot(q) - \cot(p)|^{2s-1}} dp, & s \neq \frac{1}{2}. \end{cases}$$

[Cayama, Cuesta, Hoz, 2019]

• mapping $v \mapsto \xi$

$$\xi = \frac{v}{\sqrt{L_v^2 + v^2}} \in (-1, 1) \Longleftrightarrow v = \frac{L_v \xi}{\sqrt{1 - \xi^2}} \in (-\infty, \infty) \,,$$

• Chebyshev polynomials

$$T_k(\xi) = \cos(k \arccos(\xi)), \qquad \xi \in [-1, 1]$$

• further change of variable $q = \arccos(\xi) \in [0, \pi]$

 $T_k(\xi) = \cos(kq)$

• In the new variable q $\left(v = \frac{L_v \xi}{\sqrt{1-\xi^2}} = L_v \cot(q)\right)$

$$(-\Delta_q)^s f(q) = \begin{cases} -\frac{1}{L_v \pi} \int_0^{\pi} \frac{f'(p)}{\cot(q) - \cot(p)} dp, & s = \frac{1}{2}, \\ \frac{C_{s,d}}{2L_v^{s,s}(1-2s)} \int_0^{\pi} \frac{\sin^2(p)f''(p) + 2\sin(p)\cos(p)f'(p)}{|\cot(q) - \cot(p)|^{2s-1}} dp, & s \neq \frac{1}{2}. \end{cases}$$

[Cayama, Cuesta, Hoz, 2019]

• mapping $v \mapsto \xi$

$$\xi = \frac{v}{\sqrt{L_v^2 + v^2}} \in (-1, 1) \Longleftrightarrow v = \frac{L_v \xi}{\sqrt{1 - \xi^2}} \in (-\infty, \infty),$$

• Chebyshev polynomials

$$T_k(\xi) = \cos(k \arccos(\xi)), \qquad \xi \in [-1, 1]$$

• further change of variable $q = \arccos(\xi) \in [0, \pi]$

$$T_k(\xi) = \cos(kq)$$

• In the new variable q $(v = \frac{L_v \xi}{\sqrt{1-\xi^2}} = L_v \cot(q))$

$$(-\Delta_q)^s f(q) = \begin{cases} -\frac{1}{L_v \pi} \int_0^{\pi} \frac{f'(p)}{\cot(q) - \cot(p)} dp, & s = \frac{1}{2}, \\ \frac{C_{s,d}}{2L_v^{2s} s(1-2s)} \int_0^{\pi} \frac{\sin^2(p) f''(p) + 2\sin(p) \cos(p) f'(p)}{|\cot(q) - \cot(p)|^{2s-1}} dp, & s \neq \frac{1}{2}. \end{cases}$$

◆□ > ◆□ > ◆豆 > ◆豆 > ̄豆 = つへ⊙

[Cayama, Cuesta, Hoz, 2019]

• mapping $v \mapsto \xi$

$$\xi = \frac{v}{\sqrt{L_v^2 + v^2}} \in (-1, 1) \Longleftrightarrow v = \frac{L_v \xi}{\sqrt{1 - \xi^2}} \in (-\infty, \infty),$$

• Chebyshev polynomials

$$T_k(\xi) = \cos(k \arccos(\xi)), \qquad \xi \in [-1, 1]$$

• further change of variable $q = \arccos(\xi) \in [0, \pi]$

$$T_k(\xi) = \cos(kq)$$

• In the new variable q $(v = \frac{L_v \xi}{\sqrt{1-\xi^2}} = L_v \cot(q))$

$$(-\Delta_q)^s f(q) = \begin{cases} -\frac{1}{L_v \pi} \int_0^{\pi} \frac{f'(p)}{\cot(q) - \cot(p)} dp, & s = \frac{1}{2}, \\ \frac{C_{s,d}}{2L_v^{2s} s(1-2s)} \int_0^{\pi} \frac{\sin^2(p) f''(p) + 2\sin(p) \cos(p) f'(p)}{|\cot(q) - \cot(p)|^{2s-1}} dp, & s \neq \frac{1}{2}. \end{cases}$$

◆□ > ◆□ > ◆豆 > ◆豆 > ̄豆 = つへ⊙

• even extension

$$\tilde{f}(q) = \begin{cases} f(q), & q \in [0, \pi] \\ f(2\pi - q), & q \in [\pi, 2\pi] \end{cases}$$

• discrete Fourier transform

$$f(q_j) = \sum_{k=-N_v}^{N_v-1} \hat{f}_k e^{ikq_j}, \quad q_j = \frac{\pi(2j+1)}{2N_v}, \qquad 0 \le j \le N_v - 1$$

• compute $(-\Delta_v)^s f$

$$(-\Delta_q)^s f(q_j) = \sum_{k=-N_v}^{N_v-1} \hat{f}_k (-\Delta_q)^s e^{ikq_j}, \qquad j = 0, \cdots, 2N_v - 1.$$

• compute $(-\Delta_v)^s (e^{imq})$

• even extension

$$\tilde{f}(q) = \begin{cases} f(q), & q \in [0,\pi] \\ f(2\pi - q), & q \in [\pi, 2\pi] \end{cases}$$

• discrete Fourier transform

$$f(q_j) = \sum_{k=-N_v}^{N_v - 1} \hat{f}_k e^{ikq_j}, \quad q_j = \frac{\pi(2j+1)}{2N_v}, \qquad 0 \le j \le N_v - 1$$

• compute $(-\Delta_v)^s f$

$$(-\Delta_q)^s f(q_j) = \sum_{k=-N_v}^{N_v-1} \hat{f}_k (-\Delta_q)^s e^{ikq_j}, \qquad j = 0, \cdots, 2N_v - 1.$$

• compute $(-\Delta_v)^s (e^{imq})$

◆□▶ ◆御▶ ◆臣▶ ◆臣▶ ―臣 …のへ⊙

• even extension

$$\tilde{f}(q) = \left\{ \begin{array}{ll} f(q), & q \in [0,\pi] \\ f(2\pi - q), & q \in [\pi, 2\pi] \end{array} \right.$$

• discrete Fourier transform

$$f(q_j) = \sum_{k=-N_v}^{N_v-1} \hat{f}_k e^{ikq_j}, \quad q_j = \frac{\pi(2j+1)}{2N_v}, \qquad 0 \le j \le N_v - 1$$

• compute $(-\Delta_v)^s f$

$$(-\Delta_q)^s f(q_j) = \sum_{k=-N_v}^{N_v-1} \hat{f}_k (-\Delta_q)^s e^{ikq_j}, \qquad j = 0, \cdots, 2N_v - 1.$$

• compute $(-\Delta_v)^s (e^{imq})$

◆□ > ◆□ > ◆三 > ◆三 > ● ● ●

Computation of \mathcal{L}^s

• Examples: $(-\Delta_v)^s f$



• \mathcal{L} in new variable q

 $\mathcal{L}_q^s(f) := f - \cos(q)\sin(q)\partial_q f - (-\Delta_q)^s f$

Computation of \mathcal{L}^s

• Examples: $(-\Delta_v)^s f$



• \mathcal{L} in new variable q

$$\mathcal{L}_q^s(f) := f - \cos(q)\sin(q)\partial_q f - (-\Delta_q)^s f$$

• Micro-macro decomposition

$$f(t, x, v) = \frac{\eta(t, x, v)\mathcal{M}(v) + g(t, x, v)}{\eta(t, x, v) = h(t, x + \varepsilon v)}$$

• Plug the decomposition to the equation

$$\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x(\eta\mathcal{M}+g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta\mathcal{M})$$
$$= \mathcal{L}^s(g) + v\mathcal{M} \cdot \partial_v\eta - \mathcal{M}(-\Delta_v)^s\eta - I(\eta,\mathcal{M}) + \eta\mathcal{L}^s(\mathcal{M})$$

æ

・ロト ・日ト ・ヨト ・ヨト

• Micro-macro decomposition

$$f(t, x, v) = \eta(t, x, v)\mathcal{M}(v) + g(t, x, v)$$
$$\eta(t, x, v) = h(t, x + \varepsilon v)$$

• Plug the decomposition to the equation

$$\begin{split} \varepsilon^{2s} \partial_t (\eta \mathcal{M} + g) &+ \varepsilon v \partial_x (\eta \mathcal{M} + g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta \mathcal{M}) \\ &= \mathcal{L}^s(g) + v \mathcal{M} \cdot \partial_v \eta - \mathcal{M}(-\Delta_v)^s \eta - I(\eta, \mathcal{M}) + \eta \mathcal{L}^s(\mathcal{M}) \end{split}$$

æ

・ロト ・日ト ・ヨト ・ヨト

• Micro-macro decomposition

$$f(t, x, v) = \eta(t, x, v)\mathcal{M}(v) + g(t, x, v)$$
$$\eta(t, x, v) = h(t, x + \varepsilon v)$$

• Plug the decomposition to the equation

$$\begin{split} \varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) &+ \varepsilon v\partial_x(\eta\mathcal{M}+g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta\mathcal{M}) \\ &= \mathcal{L}^s(g) + v\mathcal{M} \cdot \partial_v \eta - \underbrace{\mathcal{M}(-\Delta_v)^s \eta}_{\varepsilon^{2s}\mathcal{M}(-\Delta_x)^s \eta} - I(\eta,\mathcal{M}) + \underbrace{\eta\mathcal{L}^s(\mathcal{M})}_{=0} \\ I(f,g) &= (-\Delta_v)^s(fg) - g(-\Delta_v)^s f - f(-\Delta_v)^s g \\ &= C_{1,s} \int_{\mathbb{R}^d} \frac{(f(v) - f(w))(g(w) - g(v))}{|v - w|^{d+2s}} \mathrm{d}w \,. \end{split}$$

æ

・ロト ・日ト ・ヨト ・ヨト

• Micro-macro decomposition

$$f(t, x, v) = \eta(t, x, v)\mathcal{M}(v) + g(t, x, v)$$
$$\eta(t, x, v) = h(t, x + \varepsilon v)$$

• Plug the decomposition to the equation

$$\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x(\eta\mathcal{M}+g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta\mathcal{M})$$

= $\mathcal{L}^s(g) + v\mathcal{M} \cdot \partial_v \eta - \underbrace{\mathcal{M}(-\Delta_v)^s \eta}_{\varepsilon^{2s}\mathcal{M}(-\Delta_x)^s \eta} - I(\eta,\mathcal{M}) + \underbrace{\eta\mathcal{L}^s(\mathcal{M})}_{=0}$
 $\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x g = \mathcal{L}^s(g) - \varepsilon^{2s}(-\Delta_x)^s \eta\mathcal{M} - I(\eta,\mathcal{M})$

• Split system

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}), \\ \partial_t \eta = -(-\Delta_x)^s \eta, \end{cases}$$

• Initial data

$$\eta_{in}(x,v) = \rho_{in}(x+\varepsilon v) \qquad g_{in}(x,v) = f_{in}(x,v) - \eta_{in}(x,v)\mathcal{M}(v) \,.$$

• Micro-macro decomposition

$$f(t, x, v) = \eta(t, x, v)\mathcal{M}(v) + g(t, x, v)$$
$$\eta(t, x, v) = h(t, x + \varepsilon v)$$

• Plug the decomposition to the equation

$$\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x(\eta\mathcal{M}+g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta\mathcal{M})$$

= $\mathcal{L}^s(g) + v\mathcal{M} \cdot \partial_v \eta - \underbrace{\mathcal{M}(-\Delta_v)^s \eta}_{\varepsilon^{2s}\mathcal{M}(-\Delta_x)^s \eta} - I(\eta,\mathcal{M}) + \underbrace{\eta\mathcal{L}^s(\mathcal{M})}_{=0}$
 $\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x g = \mathcal{L}^s(g) - \varepsilon^{2s}(-\Delta_x)^s \eta\mathcal{M} - I(\eta,\mathcal{M})$

• Split system

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}), \\ \partial_t \eta = -(-\Delta_x)^s \eta, \end{cases}$$

• Initial data

$$\eta_{in}(x,v) = \rho_{in}(x+\varepsilon v) \qquad g_{in}(x,v) = f_{in}(x,v) - \eta_{in}(x,v)\mathcal{M}(v) \,.$$

Li Wang (Minnesota)

• Micro-macro decomposition

$$\begin{aligned} f(t,x,v) &= \eta(t,x,v)\mathcal{M}(v) + g(t,x,v) \\ \eta(t,x,v) &= h(t,x+\varepsilon v) \end{aligned}$$

• Plug the decomposition to the equation

$$\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x(\eta\mathcal{M}+g) = \mathcal{L}^s(g) + \mathcal{L}^s(\eta\mathcal{M})$$

= $\mathcal{L}^s(g) + v\mathcal{M} \cdot \partial_v \eta - \underbrace{\mathcal{M}(-\Delta_v)^s \eta}_{\varepsilon^{2s}\mathcal{M}(-\Delta_x)^s \eta} - I(\eta,\mathcal{M}) + \underbrace{\eta\mathcal{L}^s(\mathcal{M})}_{=0}$
 $\varepsilon^{2s}\partial_t(\eta\mathcal{M}+g) + \varepsilon v\partial_x g = \mathcal{L}^s(g) - \varepsilon^{2s}(-\Delta_x)^s \eta\mathcal{M} - I(\eta,\mathcal{M})$

• Split system

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) \\ \partial_t \eta = -(-\Delta_x)^s \eta \end{cases}$$

• Initial data

$$\eta_{in}(x,v) = \rho_{in}(x+\varepsilon v) \qquad g_{in}(x,v) = f_{in}(x,v) - \eta_{in}(x,v)\mathcal{M}(v)$$

Li Wang (Minnesota)

Energy stability

Proposition

If (η, g) solves split system, then $f = \eta \mathcal{M} + g$ solves the original equation. Both system has the energy dissipation property. That is, define the total energy

$$E_f^2 = \int_{\mathbb{R}} \int_{\mathbb{R}} \frac{f^2}{\mathcal{M}} \mathrm{d}v \mathrm{d}x = \int_{\mathbb{R}} \int_{\mathbb{R}} \frac{(\eta \mathcal{M} + g)^2}{\mathcal{M}} \mathrm{d}v \mathrm{d}x,$$

then $\frac{dE_f}{dt} \leq 0$.

Proposition

$$E_{\eta}^{2} = \int_{\mathbb{R}} \int_{\mathbb{R}} \eta^{2} \mathcal{M} dx dv, \qquad E_{g}^{2} = \int_{\mathbb{R}} \int_{\mathbb{R}} \frac{g^{2}}{\mathcal{M}} dx dv$$

are both uniformly bounded in time.

3

・ロト ・ 日 ・ ・ ヨ ・ ・ ヨ ・ ・

To solve

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) \\ \partial_t \eta = -(-\Delta_x)^s \eta \end{cases}$$

we propose

$$\begin{cases} \frac{\varepsilon^{2s}}{\Delta t}(g^* - g^n) = \mathcal{L}^s(g^*) - \gamma g^* - I(\eta^n, \mathcal{M}) \\ \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^*) + \varepsilon v \partial_x g^{n+1} = \gamma g^{n+1} \\ \frac{1}{\Delta t}(\eta^{n+1} - \eta^n) = -(-\Delta_x)^s \eta^n \end{cases}$$

Compare to $\frac{\varepsilon^{2s}}{\Delta t}(g^{n+1}-g^n) + \varepsilon v \partial_x g^{n+1} = \mathcal{L}^s(g^{n+1}) - I(\eta^n, \mathcal{M})$ • alleviate ill-conditioning

	4.54	
$s = 0.8, \varepsilon = 1e - 3$		
$s = 0.8, \varepsilon = 1e - 5$		
	• • • • • •	

To solve

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) \\ \partial_t \eta = -(-\Delta_x)^s \eta \end{cases}$$

we propose

$$\begin{cases} \frac{\varepsilon^{2s}}{\Delta t}(g^* - g^n) = \mathcal{L}^s(g^*) - \gamma g^* - I(\eta^n, \mathcal{M}) \\ \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^*) + \varepsilon v \partial_x g^{n+1} = \gamma g^{n+1} \\ \frac{1}{\Delta t}(\eta^{n+1} - \eta^n) = -(-\Delta_x)^s \eta^n \end{cases}$$

Compare to $\frac{\varepsilon^{2s}}{\Delta t}(g^{n+1}-g^n) + \varepsilon v \partial_x g^{n+1} = \mathcal{L}^s(g^{n+1}) - I(\eta^n, \mathcal{M})$

• alleviate ill-conditioning

	4.54	
$s = 0.8, \varepsilon = 1e - 3$		
$s = 0.8, \varepsilon = 1e - 5$		

To solve

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) \\ \partial_t \eta = -(-\Delta_x)^s \eta \end{cases}$$

we propose

$$\begin{cases} \frac{\varepsilon^{2s}}{\Delta t}(g^* - g^n) = \mathcal{L}^s(g^*) - \gamma g^* - I(\eta^n, \mathcal{M}) \\ \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^*) + \varepsilon v \partial_x g^{n+1} = \gamma g^{n+1} \\ \frac{1}{\Delta t}(\eta^{n+1} - \eta^n) = -(-\Delta_x)^s \eta^n \end{cases}$$

Compare to $\frac{\varepsilon^{2s}}{\Delta t}(g^{n+1}-g^n) + \varepsilon v \partial_x g^{n+1} = \mathcal{L}^s(g^{n+1}) - I(\eta^n, \mathcal{M})$

• alleviate ill-conditioning

	$Cond(A_0)$	$Cond(A_{\frac{1}{2}})$	$Cond(A_1)$
$s = 0.4, \ \varepsilon = 1$	4.73	4.54	4.36
$s = 0.6, \epsilon = 1$	4.93	4.73	4.55
$s = 0.8, \varepsilon = 1$	13.28	12.66	12.11
$s = 0.4, \varepsilon = 1e - 3$	7.74e3	158	52
$s = 0.6, \varepsilon = 1e - 3$	1.24e5	187.42	63.44
$s = 0.8, \varepsilon = 1e - 3$	5.97e6	564.32	190.75
$s = 0.4, \varepsilon = 1e - 5$	3.57e5	181.66	55.84
$s = 0.6, \varepsilon = 1e - 5$	3.15e7	188.98	63.67
$s=0.8,\varepsilon=1e-5$	9.46e9	564.62	190.79

To solve

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) + \eta \mathcal{L}^s(\mathcal{M}), \\ \partial_t \eta = -(-\Delta_x)^s \eta, \end{cases}$$

we propose

$$\begin{cases} \frac{\varepsilon^{2s}}{\Delta t}(g^* - g^n) = \mathcal{L}^s(g^*) - \gamma g^* - I(\eta^n, \mathcal{M}) \\ \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^*) + \varepsilon v \partial_x g^{n+1} = \gamma g^{n+1} \\ \frac{1}{\Delta t}(\eta^{n+1} - \eta^n) = -(-\Delta_x)^s \eta^n \end{cases}$$

Compare to $\frac{\varepsilon^{2s}}{\Delta t}(g^{n+1}-g^n) + \varepsilon v \partial_x g^{n+1} = \mathcal{L}^s(g^{n+1}) - I(\eta^n, \mathcal{M})$

- alleviate ill-conditioning
- reduce computational cost

• asymptotic property

◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへの

To solve

$$\begin{cases} \varepsilon^{2s} \partial_t g + \varepsilon v \partial_x g = \mathcal{L}^s(g) - I(\eta, \mathcal{M}) + \eta \mathcal{L}^s(\mathcal{M}), \\ \partial_t \eta = -(-\Delta_x)^s \eta, \end{cases}$$

we propose

$$\begin{cases} \frac{\varepsilon^{2s}}{\Delta t}(g^* - g^n) = \mathcal{L}^s(g^*) - \gamma g^* - I(\eta^n, \mathcal{M}) \\ \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^*) + \varepsilon v \partial_x g^{n+1} = \gamma g^{n+1} \\ \frac{1}{\Delta t}(\eta^{n+1} - \eta^n) = -(-\Delta_x)^s \eta^n \end{cases}$$

Compare to $\frac{\varepsilon^{2s}}{\Delta t}(g^{n+1}-g^n) + \varepsilon v \partial_x g^{n+1} = \mathcal{L}^s(g^{n+1}) - I(\eta^n, \mathcal{M})$

- alleviate ill-conditioning
- reduce computational cost
- asymptotic property

イロト イヨト イヨト イヨト ヨー のへの

AP property

Proposition

The numerical solution $\rho^n = \langle f^n \rangle = \langle \eta^n \mathcal{M} + g^n \rangle$ satisfies

$$\frac{\rho^{n+1}-\rho^n}{\Delta t} = (-\Delta_x)^s \rho^n, \qquad as \quad \varepsilon \to 0.$$

• From the reconstruction formula

$$\frac{e^{n+1} - \rho^n}{\Delta t} = \left\langle \frac{\eta^{n+1} - \eta^n}{\Delta t} \mathcal{M} \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= -\left\langle (-\Delta_x)^s \eta^n \mathcal{M} \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= (-\Delta_x)^s \left\langle f^n - g^n \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= (-\Delta_x)^s \rho^n - (-\Delta_x)^s \left\langle g^n \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$

(日) (四) (王) (王) (王) (王)

AP property

Proposition

The numerical solution $\rho^n = \langle f^n \rangle = \langle \eta^n \mathcal{M} + g^n \rangle$ satisfies

$$\frac{\rho^{n+1}-\rho^n}{\Delta t} = (-\Delta_x)^s \rho^n, \qquad as \quad \varepsilon \to 0.$$

• From the reconstruction formula

$$\frac{\rho^{n+1} - \rho^n}{\Delta t} = \left\langle \frac{\eta^{n+1} - \eta^n}{\Delta t} \mathcal{M} \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= -\left\langle (-\Delta_x)^s \eta^n \mathcal{M} \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= (-\Delta_x)^s \left\langle f^n - g^n \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= (-\Delta_x)^s \rho^n - (-\Delta_x)^s \left\langle g^n \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$

(日) (四) (王) (王) (王) (王)

AP property

Proposition

The numerical solution $\rho^n = \langle f^n \rangle = \langle \eta^n \mathcal{M} + g^n \rangle$ satisfies

$$\frac{\rho^{n+1}-\rho^n}{\Delta t} = (-\Delta_x)^s \rho^n, \qquad as \quad \varepsilon \to 0.$$

• From the reconstruction formula

$$\frac{\rho^{n+1} - \rho^n}{\Delta t} = \left\langle \frac{\eta^{n+1} - \eta^n}{\Delta t} \mathcal{M} \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$
$$= \left(-\Delta_x \right)^s \rho^n - \left(-\Delta_x \right)^s \left\langle g^n \right\rangle + \left\langle \frac{g^{n+1} - g^n}{\Delta t} \right\rangle$$

$$\begin{aligned} \bullet & \left(\frac{\varepsilon^{2s}}{\Delta t} + \gamma + \mathcal{L}^{s}\right)g^{*} = \frac{\varepsilon^{2s}}{\Delta t}g^{n} - I(\eta^{n}, \mathcal{M}) \\ & \|g^{*}(x, \cdot)\|_{L_{v}^{\infty}} \lesssim \|I(\eta^{n}, \mathcal{M})\|_{L_{v}^{\infty}} + \frac{\varepsilon^{2s}}{\Delta t}\|g^{n}\|_{L_{v}^{\infty}} \end{aligned} \\ \bullet & \frac{\varepsilon^{2s}}{\Delta t}(g^{n+1} - g^{*}) + \varepsilon v \partial_{x}g^{n+1} = \gamma g^{n+1} \\ & |g^{n+1}(x, v)| \le \frac{2\varepsilon^{2s}}{\Delta t}\|g^{*}(\cdot, v)\|_{L_{x}^{\infty}} \end{aligned}$$

Outline



2 Numerical methods

3 Numerical examples



æ

・ロト ・回ト ・ヨト ・ヨト

Long time behavior

$$\partial_t f = \partial_v (vf) - (-\Delta_v)^s f, \qquad f(0,v) = e^{-v^2}$$

 $s = 0.5, \quad \mathcal{M}(v) = \pi^{-0.5} (1 + v^2)^{-1}:$



Figure: Here $\Delta t = 0.01$ and $N_v = 128$

æ

イロン イヨン イヨン イヨン

Long time behavior



Figure: Here use $N_v = 128$, $\Delta t = 0.01$. top: s = 0.6. bottom: s = 0.8.

Uniform accuracy in time

$$\begin{cases} \partial_t f + v \cdot \nabla_x f = \nabla_v \cdot (vf) - (-\Delta_v)^s f := \mathcal{L}^s(f), \quad s \in (0,1) \\ f(0,x,v) = \pi^{-0.5} e^{-15x^2} e^{-v^2}, \quad x \in [-5,5] \end{cases}$$



Figure: Convergence test for s = 0.4 (left) and s = 0.8 (right). $L_x = 5$, $N_x = 200$, $L_v = 3$ and $N_v = 128$, $\Delta t = 0.025$, 0.0125, 0.00625, 0.003125, 0.0015625.

・ロト ・ 日 ・ ・ ヨ ・ ・ ヨ ・ ・

Stability





Figure: $L_x = 5$, $L_v = 3$, $N_x = 100$, $N_v = 128$, $\Delta t = 0.01$.

28 / 32

《曰》 《圖》 《臣》 《臣》 三臣

Kinetic regime

Reference solution:

$$\begin{cases} \frac{f^* - f^n}{\Delta t} + v\partial_x f^n = 0, \\ \frac{f^{n+1} - f^*}{\Delta t} = \partial_v (vf^{n+1}) - (-\Delta_v)^s f^{n+1}. \end{cases}$$



Figure: For our AP scheme, $N_x = 200$, $N_v = 128$, $\Delta t = 0.01$. For reference solution, $N_x = 800$, $N_v = 256$, $\Delta t = 1e - 4$.

<ロ> (四) (四) (三) (三) (三) 三

Diffusive regime



Figure: $T = 0.1, N_x = 100, \Delta t = 0.01, N_v = 128.$

Li Wang (Minnesota)

Outline



2 Numerical methods

3 Numerical examples



æ

< ロ > (四) (四) (三) (三) (

Conclusion

We have developed an *asymptotic preserving* scheme for Lévy Fokker Planck equation with *fractional* diffusion limit.

Key ideas:

- spectral method with nonlocal basis
- a new macro-micro decomposition
- an operator splitting



《曰》 《圖》 《臣》 《臣》 三臣

Conclusion

We have developed an *asymptotic preserving* scheme for Lévy Fokker Planck equation with *fractional* diffusion limit.

Key ideas:

- spectral method with nonlocal basis
- a new macro-micro decomposition
- an operator splitting

Thank you!

< ロ > (四) (四) (三) (三) (