

Derivative estimates for screened Vlasov-Poisson system around Penrose-stable equilibria

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Abstract

In this paper, we establish derivative estimates for the Vlasov-Poisson system with screening interactions around Penrose-stable equilibria on the phase space $\mathbb{R}_x^d \times \mathbb{R}_v^d$, with dimension $d \geq 3$. In particular, we establish the optimal decay estimates for higher derivatives of the density of the perturbed system, precisely like the free transport, up to a log correction in time. This extends the recent work [13] by Han-Kwan, Nguyen and Rousset to higher derivatives of the density. The proof makes use of several key observations from [13] on the structure of the forcing term in the linear problem, with induction arguments to classify all the terms appearing in the derivative estimates.

1 Introduction

1.1 The system

In this paper, we consider the screened Vlasov-Poisson system on the phase space $(x, v) \in \mathbb{R}^d \times \mathbb{R}^d$, with the dimension $d \geq 3$:

$$\partial_t f + v \cdot \nabla_x f + E \cdot \nabla_v f = 0 \quad (1.1)$$

Here $f = f(t, x, v) \geq 0$ is the probability distribution of charged particles in plasma,

$$\rho(t, x) = \int_{\mathbb{R}^d} f(t, x, v) dv$$

is the electric charge density, and

$$E(t, x) = -\nabla_x (1 - \Delta_x)^{-1} (\rho - 1)$$

is electric field. This model has been investigated in physical literatures [6, 7, 2] and also recent mathematical works [5, 13]. We refer the readers to [19, 20, 15, 9, 17] for global existence and regularity results. The system (1.1) has

$$(f, \rho, E) = (\mu(v), 1, 0)$$

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as a steady solution for any smooth and decaying function $\mu(v) \geq 0$ with the normalized condition

$$\int_{\mathbb{R}^d} \mu(v) dv = 1.$$

We assume that $\mu(v)$ satisfies the Penrose stability condition ($\widehat{\nabla_v \mu}$ denotes the Fourier transform):

$$\inf_{\Im \tau \leq 0} \inf_{\xi \in \mathbb{R}^d} \left| 1 - \int_0^\infty e^{-i\tau s} \frac{1}{1 + |\xi|^2} i\xi \cdot \widehat{\nabla_v \mu}(s\xi) ds \right| \geq \kappa \quad \text{for } \kappa > 0, \quad (1.2)$$

which is classical in the study of Landau damping [18, 4], where the authors justify the asymptotic stability for homogeneous equilibria that satisfies (1.2) on the phase space $\mathbb{T}^d \times \mathbb{R}^d$, under analytic and Gevrey perturbations for the Vlasov-Poisson system. The stability condition (1.2) is also natural in studying the quasineutral limit of the Vlasov-Poisson equations [14, 11, 10] and the long time estimates for the Vlasov-Maxwell equations [12].

1.2 Previous works

In [5], Bedrossian, Masmoudi and Mouhot justify the asymptotic stability of the equilibria when $\mu(v)$ satisfies the condition (1.2). The proof is inspired by [4] for Landau damping on the confined phase space $\mathbb{T}^d \times \mathbb{R}^d$. Using the dispersive mechanism in Fourier space to control the plasma echo resonance, the authors in [5] prove that the Fourier mode of the density $\hat{\rho}(t, \xi)$ decays like $\frac{1}{(t|\xi|)^{N-\delta}}$ if the initial perturbation is in Sobolev space of high regularity order N and some $\delta \in (0, N)$. The decay is far from being optimal, as the dispersion in the physical space was not taken into account.

In the recent work [13], Han-Kwan, Nguyen, and Rousset revisit the asymptotic stability of equilibria that satisfy the condition (1.2). They obtain the decay estimates for the perturbed electric charge density $\rho(t)$ as follows:

$$\|\rho(t)\|_{L^1} + \langle t \rangle \|\nabla_x \rho(t)\|_{L^1} + \langle t \rangle^d \|\rho(t)\|_{L^\infty} + \langle t \rangle^{d+1} \|\nabla_x \rho(t)\|_{L^\infty} \lesssim \varepsilon_0 \log(1+t), \quad \langle t \rangle = \sqrt{t^2 + 1}.$$

Unlike [5], which relies on the nonlinear energy estimates, the authors in [13] use direct dispersive mechanism in the physical space, which is like the free transport up to $\log(t)$. This is achieved by a pointwise dispersive estimate, directly on the resolvent kernel for the linearized system around non-zero stable equilibria $\mu(v)$. This is followed by solving the equations by characteristic methods, inspired from the classical work of Bardos and Degond [3]. At the same time, the authors in [13] are able to propagate C^1 norm for the initial data and thus allow more general perturbations. We note that the dispersive mechanism of the free transport operator $\partial_t + v \cdot \nabla_x$ on $\mathbb{R}^d \times \mathbb{R}^d$ is also one of the key ingredients in the classical results of Bardos and Degond [3] in 1985, where they study the asymptotic stability of Vlasov-Poisson around vacuum ($\mu(v) = 0$).

Regarding the stability of vacuum (when $\mu(v) = 0$) for the unscreened Vlasov-Poisson systems, we refer the readers to the work [16] for the extension of Bardos-Degond results for optimal decays of higher derivatives. In [21], Smulevici obtains the spatial decay by the vector field methods. In [22], Wang justifies the stability of vacuum for Vlasov-Poisson by Fourier methods. In [8], Choi, Ha and Lee justify the same result for 2D screened Vlasov-Poisson.

2 Main results

2.1 Main theorem

In this paper, we will give decay estimates for higher derivatives of $\rho(t)$, under the assumption that the initial perturbation f_0 is small in suitable Sobolev space and for $\mu(v)$ satisfying decaying assumption: Given any $m \in \mathbb{N}$ and $M > 0$, there exists $C_{m,M} > 0$ such that

$$|\nabla_v^m \mu(v)| \leq C_{m,M} \langle v \rangle^{-M} \quad \text{for all } v \in \mathbb{R}^d. \quad (2.1)$$

The equation for the perturbed quantities around the equilibria of (1.1) reads

$$\begin{cases} \partial_t f + v \cdot \nabla_x f + E \cdot \nabla_v \mu = -E \cdot \nabla_v f, \\ E = -\nabla_x (1 - \Delta_x)^{-1} \rho, \\ \rho = \int_{\mathbb{R}^d} f dv. \end{cases} \quad (2.2)$$

Our main theorem is as follows:

Theorem 2.1. *Let $N > 1$ be an integer. Let $\mu(v) \geq 0$ be sufficiently smooth and satisfy the Penrose stability condition (1.2), the decaying bound (2.1) and the normalized condition $\int_{\mathbb{R}^d} \mu(v) dv = 1$. There exists $\varepsilon_0 > 0$ such that for all $f_0(x, v)$ satisfying the smallness assumption*

$$\max_{0 \leq k \leq N} \left(\|\nabla_{x,v}^k f_0\|_{L_{x,v}^1} + \|\nabla_{x,v}^k f_0\|_{L_x^1 L_v^\infty} \right) \leq \varepsilon_0,$$

the solution $f(t, x, v)$ to the equations (2.2) with initial data $f|_{t=0} = f_0(x, v)$ satisfies

$$\max_{0 \leq k \leq N} \left(\langle t \rangle^k \|\nabla_x^k \rho(t)\|_{L^1} + \langle t \rangle^{k+d} \|\nabla_x^k \rho(t)\|_{L^\infty} \right) \lesssim \varepsilon_0 \log(1+t).$$

where $\rho(t, x) = \int_{\mathbb{R}^d} f(t, x, v) dv$. Here the norm $\|\cdot\|_{L_x^p L_v^q}$ is defined by

$$\|g\|_{L_x^p L_v^q} = \left(\int_{\mathbb{R}^d} \|g(x, v)\|_{L_v^q}^p dx \right)^{1/p}.$$

2.2 Motivation

The improved decays for higher derivatives of $\rho(t)$ can be seen from the free transport equation

$$\partial_t f_{\text{free}} + v \cdot \nabla_x f_{\text{free}} = 0, \quad f|_{t=0} = f_0.$$

The solution is given by

$$f_{\text{free}} = f_0(x - tv, v), \quad \rho_{\text{free}}(t, x) = \int_{\mathbb{R}^d} f_0(x - tv, v) dv.$$

Making the change of variables $w = x - tv$, we obtain

$$\rho_{\text{free}}(t, x) = \int_{\mathbb{R}^d} f_0 \left(w, \frac{x-w}{t} \right) t^{-d} dw. \quad (2.3)$$

Hence

$$\|\rho_{\text{free}}(t)\|_{L^\infty} \leq t^{-d} \|f_0\|_{L_x^1 L_v^\infty} \quad \text{and} \quad \|\rho_{\text{free}}(t)\|_{L^1} \leq \|f_0\|_{L_x^1 L_v^1}.$$

Thus for the free transport equations, $\rho_{\text{free}}(t)$ decays like t^{-d} in L^∞ . This dispersive mechanism for the free transport (which can be seen as linearized Vlasov-Poisson around zero) is one of the key ingredients in the classical work [3] by Bardos and Degond.

Now we discuss the decay for one derivative of $\rho_{\text{free}}(t)$. Taking ∇_x on both sides of (2.3), we get

$$\nabla_x \rho_{\text{free}}(t) = t^{-d} t^{-1} \int_{\mathbb{R}^d} \nabla_v f_0 \left(w, \frac{x-w}{t} \right) dw,$$

and hence

$$\|\nabla_x \rho_{\text{free}}(t)\|_{L^\infty} \leq t^{-(d+1)} \|\nabla_v f_0\|_{L_x^1 L_v^\infty} \quad \text{and} \quad \|\nabla_x \rho_{\text{free}}(t)\|_{L^1} \leq t^{-1} \|\nabla_v f_0\|_{L_x^1 L_v^1}.$$

This implies that $\nabla_x \rho_{\text{free}}(t)$ decays like t^{-d-1} in L^∞ and t^{-1} in L^1 . This decaying mechanism is in fact extended to the Vlasov-Poisson system around vacuum by Hwang, Rendall and Velazquez [16]. In particular, the authors in [16] establish the improved decay estimates

$$\|\nabla_x^k \rho(t)\|_{L^\infty} \lesssim (1+t)^{-d-k} \quad \text{for any } k \geq 0 \quad (2.4)$$

for small initial data near vacuum.

The natural question is whether the estimate (2.4) still holds for solution to the screened Vlasov-Poisson under small perturbation around nonzero homogeneous equilibria $\mu(v)$ that satisfies Penrose condition (1.2). In this paper, we prove that this is essentially the case, namely $\nabla_x^k \rho(t)$ decays like $\nabla_x^k \rho_{\text{free}}(t)$, up to a log in time correction.

2.3 Outline of the proof

The set up: Using the characteristics

$$\begin{cases} \frac{d}{ds} X_{s,t}(x, v) &= V_{s,t}(x, v), & X_{t,t}(x, v) &= x, \\ \frac{d}{ds} V_{s,t}(x, v) &= E(s, X_{s,t}(x, v)), & V_{t,t}(x, v) &= v, \end{cases} \quad (2.5)$$

the solution $f(t, x, v)$ to the perturbation equation (2.2) can be written as

$$f(t, x, v) = f_0(X_{0,t}(x, v), V_{0,t}(x, v)) - \int_0^t E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) ds.$$

Integrating both sides in $v \in \mathbb{R}^d$ and using the fact that $E = -\nabla_x(1 - \Delta_x)^{-1} \rho$, we get

$$\rho(t, x) = \int_0^t \int_{\mathbb{R}^d} \nabla_x(1 - \Delta_x)^{-1} \rho(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds + S(t, x) \quad (2.6)$$

where $S(t, x)$ is given by

$$S(t, x) = \int_{\mathbb{R}^d} f_0(X_{0,t}(x, v), V_{0,t}(x, v)) dv + \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds - \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds. \quad (2.7)$$

Taking spacetime Fourier transform, one can express $\tilde{\rho}(\tau, \xi)$ as

$$\tilde{\rho}(\tau, \xi) = \frac{1}{1 - \tilde{K}(\tau, \xi)} \tilde{S}(\tau, \xi), \quad \text{where} \quad \tilde{K}(\tau, \xi) = \int_0^\infty e^{-i\tau t} \frac{i\xi}{1 + |\xi|^2} \cdot \widehat{\nabla_v \mu}(t\xi) dt. \quad (2.8)$$

The Penrose stability condition (1.2) is equivalent to

$$\inf_{\Im(\tau) \leq 0} \inf_{\xi \in \mathbb{R}^d} |1 - \tilde{K}(\tau, \xi)| \geq \kappa \quad \text{for some} \quad \kappa > 0,$$

which is to avoid the singularity in (2.8). Following [13], we can write in the original (t, x) variables as follows:

$$\rho(t) = S(t) + \int_0^t G(t-s) \star_x S(s) ds \quad (2.9)$$

where

$$G(t, x) = \mathcal{F}_{(\tau, \xi) \rightarrow (t, x)}^{-1} \left(\frac{\tilde{K}(\tau, \xi)}{1 - \tilde{K}(\tau, \xi)} \right). \quad (2.10)$$

One of the main results in [13] was to derive pointwise estimates for the resolvent kernel G .

Sketch of the proof of the main theorem:

Our first step is to derive higher derivative estimates for the Green kernel $G(t)$, by using Paley-Littlewood decomposition and localized frequency bounds of G . Making use of the decay bounds for $\nabla_x^k G(t)$, we are able to propagate the decay of $\nabla_x^k \rho(t)$ by the forcing term $S(t)$:

$$\|\nabla_x^k \rho(t)\|_{L^1} \lesssim \langle t \rangle^{-k} \log(1+t) \|S\|_{Y_t^N} \quad \text{and} \quad \|\nabla_x^k \rho(t)\|_{L^\infty} \lesssim \langle t \rangle^{-d-k} \log(1+t) \|S\|_{Y_t^N}. \quad (2.11)$$

where

$$\|S\|_{Y_t^N} = \max_{0 \leq k \leq N} \sup_{0 \leq s \leq t} \left(\langle s \rangle^k \|\nabla_x^k S(s)\|_{L^1} + \langle s \rangle^{d+k} \|\nabla_x^k S(s)\|_{L^\infty} \right).$$

Thus, it suffices to bound the derivatives of the forcing term $S(t, x)$, defined in (2.7). Inspired by [13, 3], we show that the trajectories $(X_{s,t}(x, v), V_{s,t}(x, v))$ are closed to the characteristics of the free transport $(x - (t-s)v, v)$. To bound $\nabla_x^k S(t, x)$, we also need to use induction argument to decompose $\nabla_x^k S(t, x)$ into quantities involving $\nabla_x^k E$, $\nabla_x^k \rho$ and lower order derivative terms involving characteristic trajectories.

2.4 Organization of the paper

The paper is organized as follows: In Section 3, we justify the estimate (2.11) for the density $\rho(t)$. In Section 4, we prove the decay estimates for the characteristics. In Section 5, we bound the forcing term $S(t, x)$, by determining the forms of its derivatives (Lemma 5.2 and 5.7), and then justify the decay estimates for each of the terms (Proposition 5.3 and Proposition 5.9).

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2.6 Notations

For any complex numbers A, B , we write $A \lesssim B$ or $A = O(B)$, to mean that there exists a universal constant $C_0 > 0$ such that $|A| \leq C_0|B|$.

For a vector field $F(x) = (F_1(x), \dots, F_q(x)) \in \mathbb{R}^q$ with $x \in \mathbb{R}^m$, we denote $\nabla_x^k F$ to be the set of derivatives

$$\{\partial^\alpha F_j : 1 \leq j \leq q, \quad |\alpha| = k\}.$$

Moreover, for two vector functions F, G defined on $x \in \mathbb{R}^m, y \in \mathbb{R}^n$, we denote $(\nabla_x^u F)(\nabla_y^v G)$ to be the set of all products XY , where $X \in \nabla_x^u F$ and $Y \in \nabla_y^v G$.

For any two vectors $a = (a_i)_{1 \leq i \leq p}, b = (b_i)_{1 \leq i \leq p}$, we denote $a \leq b$ if $a_i \leq b_i$ for all $1 \leq i \leq p$.

We also denote $\langle t \rangle = (1 + t^2)^{1/2}$. It is obvious that $\langle t \rangle \lesssim 1 + t \lesssim \langle t \rangle$ for all $t \geq 0$.

For a function $f(x)$ with $x \in \mathbb{R}^d$, we denote $\hat{f}(\xi)$ to be the standard Fourier transform of f , given by the formula

$$\hat{f}(\xi) = \int_{\mathbb{R}^d} f(x) e^{-ix \cdot \xi} dx.$$

For a function $h(t, x)$ with $x \in \mathbb{R}^d$ and $t \geq 0$, the space-time Fourier transform of h is denoted by $\tilde{h}(\tau, \xi)$, and is given by

$$\tilde{h}(\tau, \xi) = \int_0^\infty \int_{\mathbb{R}^d} h(t, x) e^{-i\tau t} e^{-ix \cdot \xi} dx dt.$$

We also use the Paley-Littlewood decomposition on $\mathbb{R}^d$. Let $\chi \in [0, 1]$ be a smooth compactly supported function on the annulus $\frac{1}{4} \leq |\xi| \leq 4$ and equal to one on the annulus $\frac{1}{2} \leq |\xi| \leq 2$. Define

$$\chi_q(\xi) = \chi\left(\frac{\xi}{2^q}\right) \quad \text{for } q \in \mathbb{Z}$$

For $u \in \mathcal{S}'(\mathbb{R}^d)$, we have the Paley-Littlewood decomposition

$$u = \sum_{q \in \mathbb{Z}} u_q, \quad \text{where } \hat{u}_q(\xi) = \hat{u}(\xi) \chi_q(\xi).$$

3 Linear estimates

3.1 Dispersive estimates for the Green kernel

Now let G be the kernel defined as in (2.10). Using Paley-Littlewood decomposition, we can decompose G as

$$G = \sum_{q \in \mathbb{Z}} G_q.$$

Now, we recall the following decaying bounds on the localized-in-frequency Green kernel G_q in [13].

Lemma 3.1. *Let $\mu(v)$ be smooth, satisfy the bound (2.1) and the Penrose stability condition (1.2). For any $K > 0$, there exists $A \geq 1$ such that for every $\delta \in (0, 1]$, and for every $q \in \mathbb{Z}$ with $2^q \geq A$:*

$$\|G_q(t)\|_{L^1} \lesssim \frac{2^{q(1+\delta)}}{1+2^{2q}} \cdot \frac{1}{(1+2^qt)^K} \quad \text{and} \quad \|G_q(t)\|_{L^\infty} \lesssim \frac{2^{q(d+1+\delta)}}{1+2^{2q}} \cdot \frac{1}{(1+2^qt)^K}.$$

Moreover, for $2^q \leq A$, one has

$$\|G_q(t)\|_{L^1} \lesssim \frac{2^q}{(1+2^qt)^K} \quad \text{and} \quad \|G_q(t)\|_{L^\infty} \lesssim \frac{2^{q(d+1)}}{(1+2^qt)^K}.$$

Making use of the decay bounds on the Green kernels $\{G_q\}_{q \in \mathbb{Z}}$, we establish the following bound on the derivatives of $G(t)$:

Theorem 3.2. *For any integer $k \geq 0$, there holds*

$$\|\nabla_x^k G(t)\|_{L^1} \lesssim t^{-k-1} \quad \text{and} \quad \|\nabla_x^k G(t)\|_{L^\infty} \lesssim t^{-d-1-k}$$

for $t > 0$, where G is the kernel defined in (2.10).

Proof. We take K so that $K > k + d + 1$. We bound $\|\nabla_x^k G(t)\|_{L^1}$ as follows:

$$\begin{aligned} \|\nabla_x^k G(t)\|_{L^1} &\lesssim \sum_{2^q \leq A} 2^{kq} \|G_q(t)\|_{L^1} + \sum_{2^q \geq A} 2^{kq} \|G_q(t)\|_{L^1} \\ &\lesssim \sum_{2^q \leq A} \frac{2^{q(k+1)}}{(1+2^qt)^K} + \sum_{2^q \geq A} \frac{2^{q(k+d+1+\delta)}}{(1+2^{2q})(1+2^qt)^K}. \end{aligned}$$

For the first term, we see that

$$\begin{aligned} \sum_{2^q \leq A} \frac{2^{q(k+1)}}{(1+2^qt)^K} &\lesssim \left(\sum_{2^q \leq t^{-1}} + \sum_{t^{-1} \leq 2^q \leq 1} + \sum_{1 \leq 2^q \leq A} \right) \frac{2^{q(k+1)}}{(1+2^qt)^K} \\ &\lesssim \sum_{2^q \leq t^{-1}} 2^{q(k+1)} + t^{-K} \sum_{2^q \leq 1} 2^{q(k+1-K)} + t^{-K} \sum_{1 \leq 2^q \leq A} 2^{q(k+1-K)} \\ &\lesssim t^{-(k+1)} + t^{-K} \lesssim t^{-(k+1)}. \end{aligned}$$

The second term is treated as follows:

$$\sum_{2^q \geq A} \frac{2^{q(k+d+1+\delta)}}{(1+2^{2q})(1+2^qt)^K} \lesssim t^{-K} \sum_{2^q \geq A} 2^{q(k+d+1+\delta-2-K)} \lesssim t^{-K} \lesssim t^{-k-1} \quad \text{since } K > k+1+d.$$

The decay bound for $\|\nabla_x^k G(t)\|_{L^1}$ is complete. Now we bound $\|\nabla_x^k G(t)\|_{L^\infty}$ as follows:

$$\begin{aligned} \|\nabla_x^k G(t)\|_{L^\infty} &\lesssim \sum_{2^q \leq A} 2^{kq} \|G_q(t)\|_{L^\infty} + \sum_{2^q \geq A} 2^{kq} \|G_q(t)\|_{L^\infty} \\ &\lesssim \sum_{2^q \leq A} \frac{2^{q(d+1+k)}}{(1+2^qt)^K} + \sum_{2^q \geq A} \frac{2^{q(d+1+\delta+k)}}{1+2^{2q}} \cdot \frac{1}{(1+2^qt)^K}. \end{aligned}$$

For the first term, we see that

$$\begin{aligned} \sum_{2^q \leq A} \frac{2^{q(d+1+k)}}{(1+2^qt)^K} &\lesssim \left(\sum_{2^q \leq t^{-1}} + \sum_{t^{-1} \leq 2^q \leq 1} + \sum_{1 \leq 2^q \leq A} \right) \frac{2^{q(d+1+k)}}{(1+2^qt)^K} \\ &\lesssim \sum_{2^q \leq t^{-1}} 2^{q(d+1+k)} + t^{-K} \sum_{t^{-1} \leq 2^q \leq 1} 2^{q(d+1+k-K)} + t^{-K} \sum_{1 \leq 2^q \leq A} 2^{q(d+1+k-K)} \\ &\lesssim t^{-(d+1+k)} + t^{-K} \lesssim t^{-(d+1+k)}. \end{aligned}$$

For the second term, we see that

$$\sum_{2^q \geq A} \frac{2^{q(d+1+\delta+k)}}{1+2^{2q}} \cdot \frac{1}{(1+2^qt)^K} \lesssim t^{-K} \sum_{2^q \geq A} 2^{q(d+1+\delta+k-K-2)} \lesssim t^{-K} \lesssim t^{-(d+1+k)}.$$

The proof is complete. $\square$

3.2 Decay estimates for the density and electric field of the linearized problem

In this section, we drive decay estimates for higher derivatives of the density $\rho(t)$, defined in (2.9) and the electric field $E = -\nabla_x(1 - \nabla_x)^{-1}\rho$. For $N > 0$, we define

$$\|S\|_{Y_t^N} = \max_{0 \leq k \leq N} \sup_{0 \leq s \leq t} \left(\langle s \rangle^k \|\nabla_x^k S(s)\|_{L^1} + \langle s \rangle^{d+k} \|\nabla_x^k S(s)\|_{L^\infty} \right). \quad (3.1)$$

By definition, we see that

$$\|\nabla_x^k S(s)\|_{L^1} \leq \langle s \rangle^{-k} \|S\|_{Y_t^N} \quad \text{and} \quad \|\nabla_x^k S(s)\|_{L^\infty} \leq \langle s \rangle^{-(k+d)} \|S\|_{Y_t^N}$$

for $0 \leq k \leq N$ and $0 \leq s \leq t$.

Theorem 3.3. *Let $N \geq 1$ be an integer. The density $\rho(t)$ defined in (2.9) satisfies the bound*

$$\max_{0 \leq k \leq N} \left\{ \langle t \rangle^k \|\nabla_x^k \rho(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \rho(t)\|_{L^\infty} \right\} \lesssim \log(1+t) \|S\|_{Y_t^N} \quad \text{for } t > 0. \quad (3.2)$$

Proof. Fix k so that $0 \leq k \leq N$. First we estimate $\nabla_x^k \rho$ in L^1 . Applying ∇_x^k to both sides of (2.9), we get

$$\nabla_x^k \rho(t) = \nabla_x^k S(t) + \int_0^{t/2} \nabla_x^k G(t-s) \star_x S(s) ds + \int_{t/2}^t G(t-s) \star_x \nabla_x^k S(s) ds.$$

This implies

$$\begin{aligned} \|\nabla_x^k \rho(t)\|_{L^1} &\leq \|\nabla_x^k S(t)\|_{L^1} + \int_0^{t/2} \|\nabla_x^k G(t-s)\|_{L^1} \|S(s)\|_{L^1} ds + \int_{t/2}^t \|G(t-s)\|_{L^1} \|\nabla_x^k S(s)\|_{L^1} ds \\ &\lesssim \langle t \rangle^{-k} \|S\|_{Y_t^N} + \|S\|_{Y_t^N} \left(\int_0^{t/2} \frac{1}{(t-s)^{k+1}} ds + \int_{t/2}^t \frac{1}{(1+t-s)} \langle s \rangle^{-k} ds \right) \\ &\lesssim \|S\|_{Y_t^N} \left(\langle t \rangle^{-k} + t^{-k} + \frac{\log(1+t)}{(1+t)^k} \right) \end{aligned}$$

Hence

$$\langle t \rangle^k \|\nabla_x^k \rho(t)\|_{L^1} \lesssim \log(1+t) \|S\|_{Y_t^N}.$$

Now we estimate the L^∞ norm of $\nabla_x^k \rho$. We have

$$\begin{aligned} \|\nabla_x^k \rho(t)\|_{L^\infty} &\leq \|\nabla_x^k S(t)\|_{L^\infty} + \int_0^{t/2} \|\nabla_x^k G(t-s)\|_{L^\infty} \|S(s)\|_{L^1} ds + \int_{t/2}^t \|G(t-s)\|_{L^1} \|\nabla_x^k S(s)\|_{L^\infty} ds \\ &\lesssim \|S\|_{Y_t^N} \langle t \rangle^{-k-d} + \|S\|_{Y_t^N} \left(\int_0^{t/2} \frac{1}{(t-s)^{k+d+1}} ds + \int_{t/2}^t \frac{1}{(1+t-s)} \langle s \rangle^{-k-d} ds \right) \\ &\lesssim \frac{\log(1+t)}{(1+t)^{d+k}} \|S\|_{Y_t^N}, \end{aligned}$$

and hence

$$\langle t \rangle^{d+k} \|\nabla_x^k \rho(t)\|_{L^\infty} \lesssim \log(1+t) \|S\|_{Y_t^N}.$$

The proof is complete. $\square$

Theorem 3.4. Assume that

$$\max_{0 \leq k \leq N} \left\{ \langle t \rangle^{d+k} \|\nabla_x^k \rho(t)\|_{L^\infty} + \langle t \rangle^k \|\nabla_x^k \rho(t)\|_{L^1} \right\} \lesssim \varepsilon \log(1+t),$$

there holds

$$\max_{0 \leq k \leq N} \left\{ \langle t \rangle^{d+k} \|\nabla_x^k E(t)\|_{L^\infty} + \langle t \rangle^k \|\nabla_x^k E(t)\|_{L^1} \right\} \lesssim \varepsilon \log(1+t).$$

Proof. Since $E = -\nabla_x(1 - \Delta_x)^{-1} \rho$, we have

$$\nabla_x^k E = -\nabla_x(1 - \Delta_x)^{-1} (\nabla_x^k \rho)$$

By the standard elliptic estimate, we get

$$\begin{cases} \|\nabla_x^k E(t)\|_{L^\infty} &\lesssim \|\nabla_x^k \rho(t)\|_{L^\infty} \lesssim \varepsilon \langle t \rangle^{-d-k} \log(1+t), \\ \|\nabla_x^k E(t)\|_{L^1} &\lesssim \|\nabla_x^k \rho(t)\|_{L^1} \lesssim \varepsilon \langle t \rangle^{-k} \log(1+t). \end{cases}$$

The proof is complete. $\square$

4 Decay estimates for the characteristics

4.1 Main theorem

From the equations (2.5), the characteristics $X_{s,t}(x, v)$ and $V_{s,t}(x, v)$ can be written as follows:

$$\begin{cases} X_{s,t}(x, v) &= x - (t - s)v + \int_s^t (\tau - s)E(\tau, X_{\tau,t}(x, v))d\tau, \\ V_{s,t}(x, v) &= v - \int_s^t E(\tau, X_{\tau,t}(x, v))d\tau. \end{cases}$$

Following [13], we define $(Y_{s,t}(x, v), W_{s,t}(x, v))$ so that

$$\begin{cases} X_{s,t}(x, v) = x - (t - s)v + Y_{s,t}(x - tv, v), \\ V_{s,t}(x, v) = v + W_{s,t}(x - tv, v). \end{cases} \quad (4.1)$$

Hence, we get

$$\begin{cases} Y_{s,t}(x, v) &= \int_s^t (\tau - s)E(\tau, x + \tau v + Y_{\tau,t}(x, v))d\tau, \\ W_{s,t}(x, v) &= - \int_s^t E(\tau, x + \tau v + Y_{\tau,t}(x, v))d\tau. \end{cases} \quad (4.2)$$

Our main theorem in this section is as follows:

Theorem 4.1. *Assume that*

$$\max_{0 \leq k \leq N} \left(\langle t \rangle^{d+k} \|\nabla_x^k E(t)\|_{L^\infty} \right) \lesssim \varepsilon \log(1 + t) \quad \text{for all } t > 0, \quad (4.3)$$

there holds, for all $s \in [0, t]$, the following inequalities:

$$\max_{0 \leq k \leq N} \|\nabla_v^k Y_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-2}} \quad (4.4)$$

and

$$\max_{0 \leq k \leq N} \|\nabla_v^k W_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-1}}. \quad (4.5)$$

Before giving the proof, we recall the Faa di Bruno's formula in [1], which allows us to compute higher order derivatives of $Y_{s,t}$ and $W_{s,t}$ by the generalized chain rule:

Lemma 4.2. *Let $u : \mathbb{R}^d \rightarrow \mathbb{R}^m$ and $F : \mathbb{R}^m \rightarrow \mathbb{R}$ be smooth functions. For each multi-index $\alpha \in \mathbb{N}^d$, we have*

$$\partial^\alpha (F \circ u) = \sum_{\mu, \nu} C_{\mu, \nu} \partial^\mu F \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq m} (\partial^\beta u^j)^{\nu_{\beta j}}$$

where $C_{\mu, \nu}$ are non negative integers, and the sum is taken over μ, ν such that $1 \leq |\mu| \leq |\alpha|$, $\nu_{\beta j} \in \mathbb{N}^$,*

$$\sum_{1 \leq |\beta| \leq |\alpha|} \nu_{\beta j} = \mu_j \quad \text{for } 1 \leq j \leq m, \quad \text{and} \quad \sum_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq m} \beta \nu_{\beta j} = \alpha.$$

Proof of Theorem 4.1: Fix $k \in \{0, 1, \dots, N\}$. We shall prove that

$$\langle s \rangle^{d-2} \|\nabla_v^k Y_{s,t}\|_{L^\infty} + \langle s \rangle^{d-1} \|\nabla_v^k W_{s,t}\|_{L^\infty} \lesssim \varepsilon \log(1+s).$$

In [13], the authors prove the above statement when $k \in \{0, 1\}$, thus we shall only consider $k \geq 2$. Let us first justify the bound for the case $k = 2$. Applying $\partial_{v_i v_j}^2$ to both sides of (4.2), we have

$$\begin{aligned} \partial_{v_i v_j}^2 Y_{s,t}(x, v) &= \int_s^t (\tau - s) \partial_{x_i x_j}^2 E(\tau, x + \tau v + Y_{\tau,t}(x, v)) (\tau + \partial_{v_j} Y_{\tau,t}(x, v)) (\tau + \partial_{v_i} Y_{\tau,t}(x, v)) d\tau \\ &\quad + \int_s^t (\tau - s) \partial_{x_i} E(\tau, x + \tau v + Y_{\tau,t}(x, v)) \partial_{v_i v_j}^2 Y_{\tau,t}(x, v) d\tau. \end{aligned}$$

This implies

$$\begin{aligned} |\partial_{v_i v_j}^2 Y_{s,t}(x, v)| &\leq \int_s^t (\tau - s) \|\nabla_x^2 E(\tau)\|_{L^\infty} (\tau + \|\nabla_v Y_{\tau,t}\|_{L^\infty})^2 d\tau + \int_s^t (\tau - s) \|\nabla_x E(\tau)\|_{L^\infty} \sup_{0 \leq s \leq t} \|\nabla_v^2 Y_{s,t}\|_{L^\infty} d\tau \\ &\lesssim \int_s^t (\tau - s) \cdot \varepsilon \log(1+\tau) \langle \tau \rangle^{-d-2} (\tau + \varepsilon)^2 d\tau + \sup_{0 \leq s \leq t} \|\nabla_v^2 Y_{s,t}\|_{L^\infty} \int_s^t (\tau - s) \cdot \varepsilon \log(1+\tau) \langle \tau \rangle^{-d-1} d\tau. \end{aligned}$$

Hence we get

$$\|\nabla_v^2 Y_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-2}} + \left(\varepsilon \frac{\log(1+s)}{(1+s)^{d-1}} \right) \sup_{0 \leq s \leq t} \|\nabla_v^2 Y_{s,t}\|_{L^\infty}.$$

Thus, as long as $\varepsilon \frac{\log(1+s)}{(1+s)^{d-1}}$ is small, we have

$$\|\nabla_v^2 Y_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-2}}.$$

Now we estimate $\nabla_v^2 W_{s,t}$. Applying $\partial_{v_i v_j}^2$ to both sides of (4.2), we get

$$\begin{aligned} \partial_{v_i v_j}^2 W_{s,t}(x, v) &= - \int_s^t \left(\partial_{x_i x_j}^2 E(\tau, x + \tau v + Y_{\tau,t}(x, v)) (\tau + \partial_{v_j} Y_{\tau,t}(x, v)) \right) (\tau + \partial_{v_i} Y_{\tau,t}(x, v)) d\tau \\ &\quad - \int_s^t \partial_{x_i} E(\tau, x + \tau v + Y_{\tau,t}(x, v)) \left(\partial_{v_i v_j}^2 Y_{\tau,t}(x, v) \right) d\tau. \end{aligned}$$

This implies

$$|\partial_{v_i v_j}^2 W_{s,t}(x, v)| \leq \int_s^t \|\nabla_x^2 E(\tau)\|_{L^\infty} (\tau + \|\nabla_v Y_{\tau,t}\|_{L^\infty})^2 d\tau + \int_s^t \|\nabla_x E(\tau)\|_{L^\infty} \|\nabla_v^2 Y_{\tau,t}\|_{L^\infty} d\tau$$

Using the fact that $\langle t \rangle^d \|E(t)\|_{L^\infty} + \langle t \rangle^{d+1} \|\nabla_x E(t)\|_{L^\infty} \lesssim \varepsilon \log(1+t)$ and $\|\nabla_v^2 Y_{\tau,t}\|_{L^\infty} \lesssim \varepsilon$, we get

$$\begin{aligned} |\partial_{v_i v_j}^2 W_{s,t}(x, v)| &\lesssim \varepsilon \int_s^t \frac{\log(1+\tau)}{(1+\tau)^{d+2}} (\tau + \varepsilon)^2 d\tau + \varepsilon^2 \int_s^t \frac{\log(1+\tau)}{(1+\tau)^{d+1}} d\tau \\ &\lesssim \varepsilon \int_s^t \frac{\log(1+\tau)}{(1+\tau)^d} d\tau + \varepsilon^2 \frac{\log(1+s)}{(1+s)^d} \lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-1}}. \end{aligned}$$

Now for a general multi-index α with $k = |\alpha| \geq 3$, we proceed by induction on k , by assuming that the decay estimates (4.4) and (4.5) are true for all index with length less than k . Applying the chain rule (4.2) for $Y_{s,t}$, we get

$$\begin{aligned}
\partial_v^\alpha Y_{s,t} &= \int_s^t (\tau - s) \sum_{(\mu, \nu) \in I} C_{\mu, \nu} \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \left(\partial_v^\beta (x_j + \tau v_j + Y_{\tau, t}^j(x, v)) \right)^{\nu_{\beta_j}} d\tau \\
&= \int_s^t (\tau - s) \sum_{(\mu, \nu) \in I} C_{\mu, \nu} \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \left(\partial_v^\beta (\tau v_j + Y_{\tau, t}^j(x, v)) \right)^{\nu_{\beta_j}} d\tau \\
&= \int_s^t (\tau - s) \sum_{(\mu, \nu) \in I} C_{\mu, \nu} \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \tau^{\nu_{\beta_j}} \left(\partial_v^\beta v_j \right)^{\nu_{\beta_j}} d\tau \\
&\quad + \int_s^t (\tau - s) \sum_{(\mu, \nu) \in I} C_{\mu, \nu} \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \left(\partial_v^\beta Y_{\tau, t}^j(x, v) \right)^{\nu_{\beta_j}} d\tau.
\end{aligned} \tag{4.6}$$

where

$$I = \left\{ (\mu, \nu) \in \mathbb{N}^{2d} : \sum_{1 \leq |\beta| \leq |\alpha|} \nu_{\beta_j} = \mu_j \quad \text{for } 1 \leq j \leq d, \quad \sum_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \beta \nu_{\beta_j} = \alpha \right\}.$$

Now we fix μ, ν and estimate each term appearing in (4.6). The first term can be estimated as follows:

$$\begin{aligned}
\int_s^t (\tau - s) \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \tau^{\nu_{\beta_j}} (\partial_v^\beta v_j)^{\nu_{\beta_j}} d\tau &\lesssim \int_s^t (\tau - s) \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+|\mu|}} \tau^{\sum_{j=1}^d \nu_{\beta_j}} d\tau \\
&\lesssim \int_s^t \tau^{\mu_j+1} \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+|\mu|}} d\tau \\
&\leq \int_s^t \varepsilon \frac{\log(1 + \tau)}{(1 + \tau)^{d-1}} d\tau \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-2}}.
\end{aligned}$$

Now we estimate the second term in (4.6), which is

$$\int_s^t (\tau - s) \partial_x^\mu E \prod_{1 \leq |\beta| \leq |\alpha|, 1 \leq j \leq d} \left(\partial_v^\beta Y_{\tau, t}^j(x, v) \right)^{\nu_{\beta_j}} d\tau. \tag{4.7}$$

We consider two cases:

Case 1: $\beta_j < |\alpha|$ for all $1 \leq j \leq d$.

In this case, we use the induction hypothesis on $\partial_v^\beta Y_{\tau, t}$, which is $|\partial_v^\beta Y_{\tau, t}^j(x, v)| \lesssim \varepsilon$. Hence (4.7) can be bounded by

$$\int_s^t (\tau - s) \|\partial_x^\mu E(\tau)\|_{L^\infty} \varepsilon^{\sum_{j=1}^d \nu_{\beta_j}} d\tau \lesssim \int_s^t \tau \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+|\mu|}} \varepsilon^{|\mu|} d\tau \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-2}}.$$

Case 2: There exists $j_0 \in \{1, 2, \dots, d\}$ such that $\beta_{j_0} = |\alpha|$.

In this case, since $|\beta| \leq |\alpha|$, we have $\beta_j = 0$ for all $j \neq j_0$. Hence the term is reduced to

$$\int_s^t (\tau - s) \partial_x^\mu E \cdot (\partial_{v_{j_0}}^{|\alpha|} Y_{\tau,t}^j)^{\nu_{|\alpha|}} O(\varepsilon) d\tau,$$

where we use the fact that $\|Y_{s,t}\|_{L^\infty} \lesssim \varepsilon$. Moreover, since $\sum_{j=1}^d \beta_j \nu_{\beta_j} = \alpha$, we have $\nu_{|\alpha|} \leq 1$. Hence

$$\begin{aligned} \int_s^t (\tau - s) \partial_x^\mu E \cdot (\partial_{v_{j_0}}^{|\alpha|} Y_{\tau,t}^j)^{\nu_{|\alpha|}} O(\varepsilon) &\lesssim \varepsilon \int_s^t \tau \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+|\mu|}} \sup_{0 \leq s \leq t} \|\partial_v^{|\alpha|} Y_{s,t}\|_{L^\infty} d\tau \\ &= \sup_{0 \leq s \leq t} \|\partial_v^{|\alpha|} Y_{s,t}\|_{L^\infty} \int_s^t \varepsilon^2 \frac{\log(1 + \tau)}{(1 + \tau)^{d+|\mu|-1}} d\tau \lesssim \varepsilon^2 \sup_{0 \leq s \leq t} \|\nabla_v^{|\alpha|} Y_{s,t}\|_{L^\infty}. \end{aligned}$$

Thus, we get the inequality of the form

$$\|\nabla_v^\alpha Y_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-2}} + \varepsilon^2 \sup_{0 \leq s \leq t} \|\nabla_v^{|\alpha|} Y_{s,t}\|_{L^\infty}.$$

Hence for ε small, we get

$$\|\nabla_v^\alpha Y_{s,t}\|_{L^\infty} \lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-2}} \quad \text{for all } 0 \leq s \leq t.$$

The proof is complete.

4.2 Straightening the characteristics

Next, we recall the following lemma about straightening the characteristics from [13].

Lemma 4.3. *Let $Y_{s,t}$ be defined as in (4.1) and (4.2). Assume that E satisfies the decay estimates (4.3), there holds*

$$\|\nabla_x Y_{s,t}\|_{L^\infty} \lesssim \frac{\varepsilon \log(1 + s)}{(1 + s)^{d-1}}.$$

Proof. By the definition of $Y_{s,t}$, we have

$$Y_{s,t}(x, v) = \int_s^t (\tau - s) E(\tau, x + \tau v + Y_{\tau,t}(x, v)) d\tau$$

Hence we get

$$\begin{aligned} |\nabla_x Y_{s,t}(x, v)| &\leq \int_s^t (\tau - s) \|\nabla_x E(\tau)\|_{L^\infty} (1 + \|\nabla_x Y_{\tau,t}\|_{L^\infty}) d\tau \\ &\lesssim \int_s^t (\tau - s) \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+1}} d\tau + \sup_{0 \leq \tau \leq t} \|\nabla_x Y_{\tau,t}\|_{L^\infty} \int_s^t (\tau - s) \frac{\varepsilon \log(1 + \tau)}{(1 + \tau)^{d+1}} d\tau \\ &\lesssim \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-1}} + \left\{ \varepsilon \frac{\log(1 + s)}{(1 + s)^{d-1}} \right\} \sup_{0 \leq s \leq t} \|Y_{s,t}\|_{L^\infty}. \end{aligned}$$

Hence as long as ε is small enough, we have

$$|\nabla_x Y_{s,t}(x, v)| \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^{d-1}}$$

The proof is complete. $\square$

Lemma 4.4. *For $0 \leq s \leq t$, there exists a C^1 map $(x, v) \rightarrow \Psi_{s,t}(x, v)$ such that*

$$X_{s,t}(x, \Psi_{s,t}(x, v)) = x - (t-s)v$$

for all $x, v \in \mathbb{R}^d$. Moreover, if E satisfies the estimates (4.3), there holds

$$\langle s \rangle^d |\Psi_{s,t}(x, v) - v| + \langle s \rangle^{d-1} |\nabla_v(\Psi_{s,t}(x, v) - v)| \lesssim \varepsilon \log(1+s)$$

for all $x, v \in \mathbb{R}^d$ and $0 \leq s \leq t$.

Proof. We write

$$X_{s,t}(x, v) = x - (t-s)(v + \Phi_{s,t}(x, v))$$

and will show that $(x, v) \rightarrow (x, v + \Phi_{s,t}(x, v))$ is a C^1 diffeomorphism. To this end, we prove that

$$\langle s \rangle^d \|\Phi_{s,t}\|_{L^\infty} + \langle s \rangle^d \|\nabla_x \Phi_{s,t}\|_{L^\infty} + \langle s \rangle^{d-1} \|\nabla_v \Phi_{s,t}\|_{L^\infty} \lesssim \varepsilon \log(1+s)$$

We have

$$\begin{aligned} \Phi_{s,t}(x, v) &= -\frac{1}{t-s}(X_{s,t}(x, v) - (t-s)v) \\ &= -\frac{1}{t-s} \int_s^t (\tau-s) E(\tau, x - (t-\tau)v + Y_{\tau,t}(x - vt, v)) d\tau \\ &\lesssim \frac{1}{t-s} \int_s^t (\tau-s) \|E(\tau)\|_{L^\infty} d\tau \lesssim \int_s^t \|E(\tau)\|_{L^\infty} d\tau. \end{aligned} \tag{4.8}$$

Since $E = -\nabla_x(1 - \Delta_x)^{-1}\rho$, we have

$$\|E(\tau)\|_{L^\infty} \lesssim \|\nabla_x \rho(\tau)\|_{L^\infty} \lesssim \frac{\varepsilon \log(1+\tau)}{(1+\tau)^{d+1}}.$$

Thus we have

$$|\Phi_{s,t}(x, v)| \lesssim \int_s^t \frac{\varepsilon \log(1+\tau)}{(1+\tau)^{d+1}} d\tau \lesssim \varepsilon \frac{\log(1+s)}{(1+s)^d}.$$

Thus $\langle s \rangle^d \|\Phi_{s,t}\|_{L^\infty} \lesssim \varepsilon \log(1+s)$. Next, applying ∇_x to both sides of (4.8), we have

$$\|\nabla_x \Phi_{s,t}\|_{L^\infty} \lesssim \frac{1}{t-s} \int_s^t (\tau-s) \|\nabla_x E(\tau)\|_{L^\infty} (1 + \|\nabla_x Y_{\tau,t}\|_{L^\infty}) d\tau.$$

Now using lemma 4.3, we get

$$\|\nabla_x \Phi_{s,t}\|_{L^\infty} \lesssim \int_s^t \frac{\varepsilon \log(1+\tau)}{(1+\tau)^{d+1}} d\tau \lesssim \varepsilon \langle s \rangle^{-d} \log(1+s).$$

Now we bound $\|\nabla_v \Phi_{s,t}\|_{L^\infty}$. Applying ∇_v to both sides of (4.8), we get

$$\|\nabla_v \Phi_{s,t}\|_{L^\infty} \lesssim \frac{1}{t-s} \int_s^t (\tau-s) \|\nabla_x E(\tau)\|_{L^\infty} \{(t-\tau) + t\|\nabla_x Y_{\tau,t}\|_{L^\infty}\} d\tau.$$

Again, using Lemma 4.3, we have

$$\begin{aligned} \|\nabla_v \Phi_{s,t}\|_{L^\infty} &\lesssim \frac{1}{t-s} \int_s^t (\tau-s) \frac{\varepsilon \log(1+\tau)}{(1+\tau)^{d+1}} \left((t-\tau) + \frac{\varepsilon t \log(1+\tau)}{(1+\tau)^{d-1}} \right) d\tau \\ &\lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-1}} + \varepsilon \int_s^t \frac{t(\tau-s)}{t-s} \frac{\log(1+\tau)^2}{(1+\tau)^{d-1}} d\tau \\ &\lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-1}} + \varepsilon \int_s^t (\tau-s) \frac{\log(1+\tau)^2}{(1+\tau)^{d-1}} d\tau + \varepsilon \int_s^t (\tau-s)^2 \frac{\log(1+\tau)^2}{(1+\tau)^{d-1}} d\tau \\ &\lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-1}}. \end{aligned}$$

Hence, the map $(x, v) \rightarrow (x, v + \Phi_{s,t}(x, v))$ is a C^1 diffeomorphism. Thus there exists a C^1 diffeomorphism $v \rightarrow \Psi_{s,t}(x, v)$ such that

$$X_{s,t}(x, \Psi_{s,t}(x, v)) = x - (t-s)v.$$

Combining this with $X_{s,t}(x, v) = x - (t-s)(v + \Phi_{s,t}(x, v))$, we have

$$\begin{cases} |\Psi_{s,t}(x, v) - v| &\lesssim \|\Phi_{s,t}\|_{L^\infty} \lesssim \varepsilon \langle s \rangle^{-d} \log(1+s), \\ |\nabla_v \Psi_{s,t}(x, v) - v| &\lesssim \|\nabla_v \Phi_{s,t}\|_{L^\infty} \lesssim \varepsilon \langle s \rangle^{-(d-1)} \log(1+s). \end{cases}$$

The proof is complete. $\square$

5 Decay estimates for the forcing term

In this section, we derive decay estimates for the derivatives of the forcing term, appearing in the equation (2.6). The forcing term $S(t, x)$ is given by

$$\begin{aligned} S(t, x) &= \int_{\mathbb{R}^d} f_0(X_{0,t}(x, v), V_{0,t}(x, v)) dv + \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds \\ &\quad - \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds \\ &= \mathcal{I}(t, x) + \mathcal{R}_L(t, x) - \mathcal{R}_{NL}(t, x), \end{aligned} \tag{5.1}$$

where

$$\begin{cases} \mathcal{I}(t, x) &= \int_{\mathbb{R}^d} f_0(X_{0,t}(x, v), V_{0,t}(x, v)) dv, \\ \mathcal{R}_L(t, x) &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds, \\ \mathcal{R}_{NL}(t, x) &= \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds. \end{cases}$$

Fix $N \geq 1$, we will give decay estimates for $\|S\|_{Y_t^N}$, under the decaying assumptions on derivatives of E , $Y_{s,t}$ and $W_{s,t}$. We also recall the smallness assumption for the initial perturbation $f_0(x, v)$ as follows:

$$\max_{0 \leq k \leq N} \|\nabla_{x,v}^k f_0\|_{L_x^1 L_v^\infty} \leq \varepsilon_0. \quad (5.2)$$

Our main theorem is as follows:

Theorem 5.1. *Let $N > 1$ be an integer. Assume that $E, W_{s,t}, Y_{s,t}$ satisfy the decay estimates (4.3), (4.5) and (4.4) and f_0 satisfies the smallness assumption (5.2), there holds*

$$\sup_{0 \leq k \leq N} \left(\langle t \rangle^k \|\nabla_x^k \mathcal{I}(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \mathcal{I}(t)\|_{L^\infty} \right) \lesssim \varepsilon_0, \quad (5.3)$$

and

$$\sup_{0 \leq k \leq N} \left(\langle t \rangle^k \|\nabla_x^k \mathcal{R}(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \mathcal{R}(t)\|_{L^\infty} \right) \lesssim \varepsilon^2.$$

5.1 Decay estimates for the initial data term

First, we estimate $\mathcal{I}(t, x)$, under suitable smallness assumption (5.2) on the initial data f_0 . In [13], the authors prove that

$$\|\mathcal{I}(t)\|_{L^1} + \langle t \rangle^d \|\mathcal{I}(t)\|_{L^\infty} + \langle t \rangle \|\nabla_x \mathcal{I}(t)\|_{L^1} + \langle t \rangle^{d+1} \|\nabla_x \mathcal{I}(t)\|_{L^\infty} \lesssim \varepsilon_0.$$

Hence, we shall establish the bound (5.3) for $k \geq 2$. First, we establish the following lemma

Lemma 5.2. *Let $k \geq 2$ and*

$$\mathcal{I}(t, x) = \int_{\mathbb{R}^d} f_0(X_{0,t}(x, v), V_{0,t}(x, v)) dv.$$

The term $\nabla_x^k \mathcal{I}(t, x)$ can be written as a sum of many terms, which are all in the form

$$\int_{\mathbb{R}^d} \nabla_{x,v}^\alpha f_0 \cdot (\nabla_v^{\beta_1} Y_{0,t})^{k_1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v^{\gamma_1} W_{0,t})^{s_1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k}} \quad (5.4)$$

where $\alpha, (\beta_1, k_1), \dots, (\beta_r, k_r)$, and $(\gamma_1, s_1), \dots, (\gamma_t, s_t)$ satisfy

$$1 \leq |\alpha| \leq k \quad \text{and} \quad (\beta_1 k_1 + \cdots + \beta_r k_r) + (\gamma_1 s_1 + \cdots + \gamma_t s_t) \leq k. \quad (5.5)$$

Proof. From (4.1), we get

$$\mathcal{I}(t, x) = \int_{\mathbb{R}^d} f_0(x - tv + Y_{0,t}(x - tv, v), v + W_{0,t}(x - tv, v)) dv.$$

Let $w = x - tv$, we get

$$\mathcal{I}(t, x) = \int_{\mathbb{R}^d} f_0 \left(w + Y_{0,t}(w, \frac{x-w}{t}), \frac{x-w}{t} + W_{0,t}(w, \frac{x-w}{t}) \right) \frac{dw}{t^d}.$$

By a direct calculation, we have

$$\begin{aligned}\partial_{x_i x_j}^2 \mathcal{I}(t, x) &= \int_{\mathbb{R}^3} (\partial_{x_i x_j} f_0 \partial_{v_j} Y_{0,t} + \partial_{x_i v_j} f_0 (1 + \partial_{v_j} W_{0,t})) (\partial_{v_i} Y_{0,t}) \frac{dw}{t^{d+2}} \\ &\quad + \int_{\mathbb{R}^3} (\partial_{x_j v_i} f_0 \partial_{v_j} Y_{0,t} + \partial_{v_i v_j} f_0 (1 + \partial_{v_j} W_{0,t})) (\partial_{v_i} W_{0,t}) \frac{dw}{t^{d+2}} \\ &\quad + \int_{\mathbb{R}^3} (\partial_{v_i} f_0) (\partial_{v_i v_j} W_{0,t}) \frac{dw}{t^{d+2}}.\end{aligned}\tag{5.6}$$

It is clear from the above that (5.6) that

$$\begin{aligned}\nabla_x^2 \mathcal{I}(t, x) &= \int_{\mathbb{R}^d} \{(\nabla_x^2 f_0)(\nabla_v Y_{0,t}) + (\nabla_{x,v} f_0)(\nabla_v Y_{0,t}) + (\nabla_x f_0)(\nabla_v^2 Y_{0,t}) + (\nabla_{x,v}^2 f_0)(\nabla_v Y_{0,t}) \\ &\quad + (\nabla_v^2 f_0)(\nabla_v W_{0,t}) + (\nabla_v^2 f_0)(\nabla_v W_{0,t})^2 + (\nabla_v f_0)(\nabla_v^2 W_{0,t})\} \frac{dw}{t^{d+2}}\end{aligned}$$

which satisfies the hypothesis for $k = 2$. Now by induction, we assume that this statement is true for k , and we shall prove it for $k + 1$. Applying ∂_{x_i} to the term (5.4) and using the product rules, we have three types of terms that appear, namely:

$$\begin{aligned}I_1 &= \int_{\mathbb{R}^d} \frac{d}{dx_i} (\nabla_{x,v}^\alpha f_0) (\nabla_v^{\beta_1} Y_{0,t})^{k_1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v^{\gamma_1} W_{0,t})^{s_1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k}} \\ I_2 &= \int_{\mathbb{R}^d} (\nabla_{x,v}^\alpha f_0) (\nabla_v Y_{0,t}) (\nabla_v^{\beta_1} Y_{0,t})^{k_1-1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v^{\gamma_1} W_{0,t})^{s_1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k+1}} \\ I_3 &= \int_{\mathbb{R}^d} (\nabla_{x,v}^\alpha f_0) (\nabla_v^{\beta_1} Y_{0,t})^{k_1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v W_{0,t}) (\nabla_v^{\gamma_1} W_{0,t})^{s_1-1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k+1}}\end{aligned}$$

Here, I_1, I_2, I_3 appear when ∂_{x_i} hits $\nabla_{x,v}^\alpha f_0, (\nabla_v^{\beta_1} Y_{0,t})^{k_1}$ and $(\nabla_v^{\gamma_1} W_{0,t})^{s_1}$ respectively. Note that we assume that the derivative hits the above terms on just the pairs (β_1, k_1) or (γ_1, s_1) , as this is up to a permutation of indices.

Treating I_1 :

By a direct calculation, we see that I_1 can be written as

$$\int_{\mathbb{R}^d} (\nabla_{x,v}^{\alpha+1} f_0 \cdot \nabla_v Y_{0,t} + \nabla_{x,v}^{\alpha+1} f_0 (1 + \nabla_v W_{0,t})) (\nabla_v^{\beta_1} Y_{0,t})^{k_1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v^{\gamma_1} W_{0,t})^{s_1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k+1}}$$

which satisfies the induction hypothesis for $|\alpha| + 1 = k + 1$.

Treating I_2 :

For I_2 , we check the condition (5.5) for the new indices and multi-indices, which is

$$|\alpha| \leq k + 1 \quad \text{and} \quad 1 + \beta_1(k_1 - 1) + (\beta_2 k_2 + \cdots + \beta_r k_r) + (\gamma_1 s_1 + \cdots + \gamma_t s_t) \leq k + 1$$

The statement $|\alpha| \leq k + 1$ is true, as $|\alpha| \leq k$. For the second condition, we note that, by the induction hypothesis:

$$(\beta_1 k_1 + \beta_2 k_2 + \cdots + \beta_r k_r) + (\gamma_1 s_1 + \cdots + \gamma_t s_t) \leq k.$$

Adding both sides of the above by $1 - \beta_1$, the new left hand side can be bounded by $k + 1 - \beta_1 \leq k + 1$, since $\beta_1 \geq 0$. The proof is complete. Finally, the term I_3 is treated exactly as I_2 , and we skip the details. $\square$

Theorem 5.3. Assume that

$$\max_{0 \leq k \leq N} \|\nabla_{x,v}^k f_0\|_{L_x^1 L_v^\infty} \leq \epsilon_0$$

and $E, Y_{s,t}, W_{s,t}$ satisfy the decay estimates (4.3), (4.4), and (4.5) for $0 \leq k \leq N$. Then

$$\mathcal{I}(t, x) = \int_{\mathbb{R}^d} f_0(X_{0,t}(x, v), V_{0,t}(x, v)) dv$$

satisfies the following decay estimate:

$$\sup_{0 \leq k \leq N} \left(\langle t \rangle^k \|\nabla_x^k \mathcal{I}(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \mathcal{I}(t)\|_{L^\infty} \right) \lesssim \epsilon_0.$$

Proof. By the above lemma, it suffices to prove that

$$\langle t \rangle^k \|\nabla_x^k \mathcal{J}(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \mathcal{J}(t)\|_{L^\infty} \lesssim \epsilon_0.$$

where

$$\left\{ \begin{array}{l} \mathcal{J} = \int_{\mathbb{R}^d} \nabla_{x,v}^\alpha f_0 \cdot (\nabla_v^{\beta_1} Y_{0,t})^{k_1} \cdots (\nabla_v^{\beta_r} Y_{0,t})^{k_r} \cdot (\nabla_v^{\gamma_1} W_{0,t})^{s_1} \cdots (\nabla_v^{\gamma_t} W_{0,t})^{s_t} \frac{dw}{t^{d+k}} \\ 1 \leq |\alpha| \leq k \quad \text{and} \quad (\beta_1 k_1 + \cdots + \beta_r k_r) + (\gamma_1 s_1 + \cdots + \gamma_t s_t) \leq k. \end{array} \right.$$

We have

$$\mathcal{J}(t, x) \lesssim t^{-k} \int_{\mathbb{R}^d} |\nabla_{x,v}^\alpha f_0|(X_{0,t}(x, v), V_{0,t}(x, v)) dv$$

Hence, we get

$$t^k \|\mathcal{J}(t)\|_{L^1} + t^{d+k} \|\mathcal{J}(t)\|_{L^\infty} \lesssim \epsilon_0.$$

The proof is complete. $\square$

5.2 Decay estimates for the reaction term

In this section, we estimate the derivatives of the reaction term

$$\mathcal{R} = \mathcal{R}_L - \mathcal{R}_{NL} = \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds - \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds$$

appearing as a forcing term in (5.1). For a general time-dependent vector field $E(s)$ and a smooth decaying function μ , we also denote $\mathcal{T}$ to be

$$\mathcal{T}(E, \mu) = \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds - \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds \quad (5.7)$$

Our main theorem is as follows:

Theorem 5.4. Let $N > 1$ be an integer. Assume that $E, W_{s,t}, Y_{s,t}$ satisfy the decay estimates (4.3), (4.5) and (4.4) for all $0 \leq k \leq N$, there holds

$$\max_{0 \leq k \leq N} \left(\langle t \rangle^k \|\nabla_x^k \mathcal{R}(t)\|_{L^1} + \langle t \rangle^{d+k} \|\nabla_x^k \mathcal{R}(t)\|_{L^\infty} \right) \lesssim \epsilon^2.$$

First, we recall the following proposition from [13]. We also give a detailed proof for the readers convenience.

Proposition 5.5. *Assuming that $E, Y_{s,t}, W_{s,t}$ satisfies the decaying estimates (4.3), (4.4) and (4.5) respectively, we have*

$$\|\mathcal{T}(E, \mu)\|_{L^1} + \langle t \rangle^d \|\mathcal{T}(E, \mu)\|_{L^\infty} \lesssim \varepsilon^2.$$

Proof. Making the change of variables $v \rightarrow \Psi_{s,t}(x, v)$ so that $X_{s,t}(x, \Psi_{s,t}(x, v)) = x - (t - s)v$ (see Lemma 4.4), we have

$$\begin{aligned} \mathcal{T}(E, \mu) &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \mu(v) dv ds - \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds \\ &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \mu(v) dv ds \\ &\quad - \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \mu(V_{s,t}(x, \Psi_{s,t}(x, v))) \det(\nabla_v \Psi_{s,t}(x, v)) dv ds \end{aligned}$$

Hence, one can rewrite $\mathcal{T}$ as $\mathcal{T}_1 + \mathcal{T}_2$, where

$$\begin{aligned} \mathcal{T}_1 &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \{\mu(v) - \mu(V_{s,t}(x, \Psi_{s,t}(x, v)))\} dv ds \\ \mathcal{T}_2 &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \mu(V_{s,t}(x, \Psi_{s,t}(x, v))) \{1 - \det(\nabla_v \Psi_{s,t}(x, v))\} dv ds \end{aligned}$$

Bounding $\mathcal{T}_1(t)$:

We have

$$\begin{aligned} \mathcal{T}_1(t, x) &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t - s)v) \cdot \nabla_v \{\mu(v) - \mu(V_{s,t}(x, \Psi_{s,t}(x, v)))\} dv ds \\ &\lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t - s)v)| \langle v \rangle^{-M} \cdot |v - V_{s,t}(x, \Psi_{s,t}(x, v))| dv ds \\ &\lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t - s)v)| \langle v \rangle^{-M} \{|v - V_{s,t}(x, v)| + |V_{s,t}(x, v) - V_{s,t}(x, \Psi_{s,t}(x, v))|\} dv ds \\ &\lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t - s)v)| \langle v \rangle^{-M} \{|v - V_{s,t}(x, v)|_{L^\infty} + \|\nabla_v V_{s,t}\|_{L^\infty} |v - \Psi_{s,t}(x, v)|\} dv ds \end{aligned}$$

Using the fact that

$$|v - V_{s,t}(x, v)| \lesssim \frac{\varepsilon \log(1 + s)}{(1 + s)^{d-1}}, \quad |v - \Psi_{s,t}(x, v)| \lesssim \frac{\varepsilon \log(1 + s)}{(1 + s)^d} \quad \text{and} \quad \|\nabla_v V_{s,t}\|_{L^\infty} \lesssim 1, \quad (5.8)$$

we get

$$\mathcal{T}_1(t, x) \lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t - s)v)| \langle v \rangle^{-M} \left(\frac{\varepsilon \log(1 + s)}{(1 + s)^{d-1}} + \frac{\varepsilon \log(1 + s)}{(1 + s)^d} \right) dv ds.$$

Hence

$$\begin{aligned}\|\mathcal{T}_1(t)\|_{L^1} &\lesssim \int_0^t \frac{\varepsilon \log(1+s)}{(1+s)^{d-1}} \int_{\mathbb{R}^d} \|E(s)\|_{L^1} \langle v \rangle^{-M} dv \\ &\lesssim \int_0^t \frac{\varepsilon^2 \log(1+s)^2}{(1+s)^{d-1}} ds \lesssim \varepsilon^2,\end{aligned}$$

and

$$\begin{aligned}\|\mathcal{T}_1(t)\|_{L^\infty} &\lesssim \int_0^t \int_{\mathbb{R}^d} \|E(s)\|_{L^\infty} \langle v \rangle^{-M} \left(\frac{\varepsilon \log(1+s)}{(1+s)^{d-1}} + \frac{\varepsilon \log(1+s)}{(1+s)^d} \right) dv ds \\ &\lesssim \int_0^t \frac{\varepsilon^2 \log(1+s)^2}{(1+s)^{2d-1}} ds \lesssim \varepsilon^2 \log(1+t) \int_0^t \frac{\log(1+s)}{(1+s)^{2d-1}} ds \\ &\lesssim \varepsilon^2 \frac{\log(1+t)^2}{(1+t)^{2d-2}} \lesssim \varepsilon^2 \langle t \rangle^{-d}.\end{aligned}$$

Thus

$$\|\mathcal{T}_1(t)\|_{L^1} + \langle t \rangle^d \|\mathcal{T}_1(t)\|_{L^\infty} \lesssim \varepsilon^2.$$

Bounding $\mathcal{T}_2(t)$:

Since $v \rightarrow \Psi_{s,t}(x, v)$ is a diffeomorphism, making the change of variables $\Psi_{s,t}^{-1} : \Psi_{s,t}(x, v) \rightarrow v$ gives

$$\begin{aligned}\mathcal{T}_2(t, x) &= \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(V_{s,t}(x, v)) (1 - \det(\nabla_v \Psi_{s,t}(x, v)) \det(\nabla_v(\Psi_{s,t}^{-1}(x, v)))) dv ds \\ &\lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t-s)v)| \cdot |\nabla_v \mu|(V_{s,t}(x, v)) \cdot |\det(\nabla_v \Psi_{s,t}(x, v))^{-1} - 1| dv ds\end{aligned}$$

Using the inequality $|\nabla_v(\Psi_{s,t}(x, v) - v)| \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^{d-1}}$, we have

$$\mathcal{T}_2(t, x) \lesssim \int_0^t \int_{\mathbb{R}^d} |E(s, x - (t-s)v)| \cdot |\nabla_v \mu|(V_{s,t}(x, v)) \cdot \frac{\varepsilon \log(1+s)}{(1+s)^{d-1}} ds dv$$

Hence we have

$$\begin{aligned}\|\mathcal{T}_2(t)\|_{L^1} &\lesssim \int_0^t \left\{ \varepsilon \frac{\log(1+s)}{(1+s)^{d-1}} \|E(s)\|_{L^1} \int_{\mathbb{R}^d} \sup_{x \in \mathbb{R}^d} |\nabla_v \mu|(V_{s,t}(x, v)) dv \right\} ds \\ &\lesssim \int_0^t \varepsilon^2 \frac{\log(1+s)^2}{(1+s)^d} ds \lesssim \varepsilon^2.\end{aligned}$$

and

$$\begin{aligned}\|\mathcal{T}_2(t)\|_{L^\infty} &\lesssim \int_0^t \|E(s)\|_{L^\infty} \cdot \frac{\varepsilon \log(1+s)}{(1+s)^{d-1}} \sup_{x \in \mathbb{R}^d} \int_{\mathbb{R}^d} |\nabla_v \mu|(V_{s,t}(x, v)) dv ds \\ &\lesssim \int_0^t \varepsilon^2 \frac{\log(1+s)^2}{(1+s)^{2d-1}} ds \lesssim \varepsilon^2 \log(1+t) \int_0^t \frac{\log(1+s)}{(1+s)^{2d-1}} ds \lesssim \varepsilon^2 \frac{\log(1+t)}{(1+t)^{2d-2}} \\ &\lesssim \varepsilon^2 \langle t \rangle^{-d}\end{aligned}$$

This implies that

$$\|\mathcal{T}_2(t)\|_{L^1} + \langle t \rangle^d \|\mathcal{T}_2(t)\|_{L^\infty} \lesssim \varepsilon^2.$$

The proof is complete. $\square$

Lemma 5.6. *There holds*

$$\begin{aligned} \partial_j \mathcal{T}(E, \mu) &= \frac{1}{t} (\mathcal{T}(s \partial_j E, \mu) + \mathcal{T}(E, \partial_j \mu)) \\ &+ \frac{1}{t} \sum_{k=1}^d \int_0^t \int_{\mathbb{R}^d} \partial_{v_j} Y_{s,t}(x - tv, v) \cdot \nabla_x E_k(s, X_{s,t}(x, v)) (\partial_k \mu)(V_{s,t}(x, v)) dv ds \\ &+ \frac{1}{t} \sum_{k=1}^d \int_0^t \int_{\mathbb{R}^d} E_k(s, X_{s,t}(x, v)) (\partial_{v_j} W_{s,t})(x - tv, v) \cdot \nabla_v (\partial_k \mu)(V_{s,t}(x, v)) dv ds. \end{aligned}$$

Proof. We recall that

$$\mathcal{R}_{NL} = \int_0^t \int_{\mathbb{R}^d} E(s, X_{s,t}(x, v)) \cdot \nabla_v \mu(V_{s,t}(x, v)) dv ds.$$

Using the identities (4.1), we have

$$\mathcal{R}_{NL} = \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v + Y_{s,t}(x - tv, v)) \cdot \nabla_v \mu(v + W_{s,t}(x - tv, v)) dv ds.$$

Making the change of variable $w = x - tv$, we obtain

$$\begin{aligned} \mathcal{R}_{NL} &= \int_0^t \int_{\mathbb{R}^d} E\left(s, w + \frac{s}{t}(x - w) + Y_{s,t}(w, \frac{x-w}{t})\right) \cdot \nabla_v \mu\left(\frac{x-w}{t} + W_{s,t}(w, \frac{x-w}{t})\right) t^{-d} dw ds \\ &= t^{-d} \sum_{k=1}^d \int_0^t \int_{\mathbb{R}^d} E_k\left(s, w + \frac{s}{t}(x - w) + Y_{s,t}(w, \frac{x-w}{t})\right) \partial_k \mu\left(\frac{x-w}{t} + W_{s,t}(w, \frac{x-w}{t})\right) dw ds \end{aligned}$$

Similarly, for $\mathcal{R}_L$ we have

$$\mathcal{R}_L = \int_0^t \int_{\mathbb{R}^d} E(s, x - (t-s)v) \cdot \nabla_v \mu(v) dv ds = t^{-d} \sum_{k=1}^d \int_0^t \int_{\mathbb{R}^d} E_k\left(s, w + \frac{s}{t}(x - w)\right) \partial_k \mu\left(\frac{x-w}{t}\right) dw ds$$

The lemma follows by a direct calculation. The proof is complete. $\square$

Now we establish the following lemma, by induction on the degree of derivatives.

Lemma 5.7. *Let $n \geq 2$, then $\nabla_x^n \mathcal{R}(t, x)$ can be written as a sum of terms, are all either of the form*

$$\frac{1}{t^n} \mathcal{T}(s^k \nabla_x^k E, f(\mu))$$

or the form

$$\begin{aligned} & \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \cdots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \cdot \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \cdots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \\ & \quad \times \left\{ (s^{t_1} \nabla_x^{u_1} E) \cdots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dv ds \end{aligned}$$

where $f(\mu)$ is some expression only depending on $\mu(v)$ or its derivatives. Moreover, the indices satisfy the following set of conditions:

- No loss of derivative condition: $k \leq n$, $\max\{\{m_1, k_1, \dots, m_a, k_a\} \cup \{n_1, l_1, \dots, n_b, l_b\} \cup \{t_1, u_1, \dots, t_c, u_c\}\} \leq n$.
- E-decay condition: $(t_1, t_2, \dots, t_c) \leq (u_1, u_2, \dots, u_c)$.
- Y-W show-up condition: $\min\{a, b\} \geq 1$.
- E-show up condition: $c \geq 1$.
- W-E decay condition: If $b = 0$ then $t_h + 1 \leq u_h$ for some $1 \leq h \leq c$.

Let us call the first form to be type-I and the second one is type-II.

Remark 5.8. The conditions on the indices are important for the decay estimates. The first condition means that we do not lose derivatives in the estimates. The second condition requires a good decay for the quantities $s^{t_i} \nabla_x^{u_i} E$. The third condition means that at least $Y_{s,t}$ or $W_{s,t}$ (or their derivatives) shows up in the expression. The forth condition means that E or its derivatives must show up in the expression. Finally, the last condition means that if $W_{s,t}$ (or its derivatives) does not show up, then we have more gradient of E to control the power of s . The reason for the last condition will be clear when we estimate (5.11) later in the paper.

Proof. The lemma is proved by induction on n . Assuming the lemma is true for n , we justify the above claim for $n + 1$.

Gradient of type-I terms:

Applying ∂_j to the term $\mathcal{T}(s^k \nabla_x^k E, f(\mu))$ and using Lemma 5.6, we have

$$\begin{aligned} & \frac{1}{t^n} \partial_j T(s^k \nabla_x^k E, f(\mu)) = \frac{1}{t^{n+1}} \left(T(s^{k+1} \nabla_x^k \partial_j E, f(\mu)) + T(s^k \nabla_x^k E, \partial_j f(\mu)) \right) \\ & + \frac{1}{t^{n+1}} \int_0^t \int_{\mathbb{R}^d} \left((\partial_{v_j} Y_{s,t}(x - tv, v)) \cdot \nabla_x (s^k \nabla_x^k E)(s, X_{s,t}(x, v)) (\nabla_v f(\mu))(V_{s,t}(x, v)) \right) dv ds \\ & + \frac{1}{t^{n+1}} \int_0^t \int_{\mathbb{R}^d} (s^k \nabla_x^k E)(s, X_{s,t}(x, v)) \{ \partial_{v_j} W_{s,t}(x - tv, v) \cdot \nabla_v^2 f(\mu)(V_{s,t}(x, v)) \} dv ds. \end{aligned}$$

We can see that all of the above terms are either type-I or type-II, and satisfy the induction hypothesis with order $n + 1$.

Gradient of type-II form:

Now we consider the type-II term:

$$\mathcal{D} = \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \cdot \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \\ \times \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dv ds.$$

Here $(\nabla_v^{m_i} Y_{s,t})^{k_i}, (\nabla_v^{n_j} W_{s,t})^{l_j}$ is evaluated at $(x - tv, v)$ and $\nabla_x^{u_r} E$ is evaluated at $(s, X_{s,t}(x, v))$. Making the change of variables $w = x - tv$, we can rewrite $\mathcal{D}$ as

$$t^{-d} \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \\ \times \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dw ds.$$

where $(\nabla_v^{m_i} Y_{s,t})^{k_i}, (\nabla_v^{n_j} W_{s,t})^{l_j}$ is evaluated at $(w, \frac{x-w}{t})$ and $\nabla_x^{u_r} E$ is evaluated at

$$\left(s, Y_{s,t}(w, \frac{x-w}{t}) + w + \frac{s}{t}(x-w) \right)$$

Applying ∇_x to $\mathcal{D}$ and using the product rules, we have

$$\begin{cases} \mathcal{D}_1 &= t^{-d} \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} k_1 \left((\nabla_v^{m_1} Y_{s,t})^{k_1-1} \left\{ \nabla_v^{m_1+1} Y_{s,t} \right\} \frac{1}{t} \right) \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \\ &\quad \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dw ds, \\ \mathcal{D}_2 &= t^{-d} \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \\ &\quad \left(l_1 \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1-1} (\nabla_v^{n_1+1} W_{s,t}) \right\} \frac{1}{t} \right) \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \\ &\quad \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dw ds, \\ \mathcal{D}_3 &= t^{-d} \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \\ &\quad \left(\frac{1}{t} \left\{ s^{t_1} (\nabla_x^{u_1+1} E) (\nabla_v Y_{s,t}) \right\} + \frac{1}{t} \left\{ s^{t_1+1} \nabla_x^{u_1+1} E \right\} \right) \dots (s^{t_c} \nabla_x^{u_c} E) f(\mu) dw ds. \end{cases}$$

Now making the change of variables $v = \frac{x-w}{t}$, we get

$$\begin{cases} \mathcal{D}_1 &= \frac{1}{t^{n+1}} \int_0^t \int_{\mathbb{R}^d} k_1 \left((\nabla_v^{m_1} Y_{s,t})^{k_1-1} \left\{ \nabla_v^{m_1+1} Y_{s,t} \right\} \right) \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \\ &\quad \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dv ds, \\ \mathcal{D}_2 &= \frac{1}{t^{n+1}} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \\ &\quad \left(l_1 \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1-1} (\nabla_v^{n_1+1} W_{s,t}) \right\} \right) \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \\ &\quad \left\{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \right\} f(\mu) dv ds, \\ \mathcal{D}_3 &= \frac{1}{t^{n+1}} \int_0^t \int_{\mathbb{R}^d} \left\{ (\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a} \right\} \left\{ (\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b} \right\} \\ &\quad \left\{ s^{t_1} (\nabla_x^{u_1+1} E) (\nabla_v Y_{s,t}) + s^{t_1+1} \nabla_x^{u_1+1} E \right\} \dots (s^{t_c} \nabla_x^{u_c} E) f(\mu) dv ds. \end{cases}$$

From the above expressions, it is straightforward that the new terms are of type-II, which satisfy the induction hypothesis with $n+1$. The proof is complete. $\square$

Proposition 5.9. *Let $\mathcal{R}$ be defined as in (5.7). There holds, for $n \geq 2$, the decaying estimates*

$$\langle t \rangle^n \|\nabla_x^n \mathcal{R}(t)\|_{L^1} + \langle t \rangle^{d+n} \|\nabla_x^k \mathcal{R}(t)\|_{L^\infty} \lesssim \varepsilon^2.$$

Proof. By Lemma 5.7, $\nabla_x^n \mathcal{R}(t, x)$ can be decomposed as a sum of many terms, all are either of the form $\mathcal{R}_1$ or $\mathcal{R}_2$, where

$$\begin{cases} \mathcal{R}_1 &= \frac{1}{t^n} \mathcal{T}(s^k \nabla_x^k E, f(\mu)) \\ \mathcal{R}_2 &= \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \{(\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a}\} \cdot \{(\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b}\} \\ &\quad \{(s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E)\} f(\mu) dv ds \end{cases}$$

where

$$k \leq n \quad \text{and} \quad \max\{\{m_1, k_1, \dots, m_a, k_a\} \cup \{n_1, l_1, \dots, n_b, l_b\} \cup \{t_1, u_1, \dots, t_c, u_c\}\} \leq n.$$

Bounding $\mathcal{R}_1(t)$:

We have

$$\mathcal{R}_1(t) = \frac{1}{t^n} \mathcal{T}(s^k \nabla_x^k E, f(\mu))$$

with $k \leq n$ and $f(\mu)$ is a decay function in v which only depends on μ or its derivatives. Applying Proposition 5.5, we only need to check the assumption

$$\|s^k \nabla_x^k E(s)\|_{L^\infty} \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^d},$$

which is true, since $\|\nabla_x^k E(s)\|_{L^\infty} \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^{d+k}}$. The proof for $\mathcal{R}_1$ is complete.

Bounding $\mathcal{R}_2(t)$:

Now we bound $\mathcal{R}_2(t)$, where

$$\begin{aligned} \mathcal{R}_2(t, x) &= \frac{1}{t^n} \int_0^t \int_{\mathbb{R}^d} \{(\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a}\} \cdot \{(\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b}\} \\ &\quad \{(s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E)\} f(\mu) dv ds \end{aligned}$$

Here $(\nabla_v^{m_i} Y_{s,t})^{k_i}, (\nabla_v^{n_j} W_{s,t})^{l_j}$ is evaluated at $(x - tv, v)$ and $\nabla_x^{u_r} E$ is evaluated at $(s, X_{s,t}(x, v))$.

We also recall the following set of conditions, proved in Lemma 5.7:

$$\left\{ \begin{array}{l} k \leq n \quad \text{and} \quad \max\{\{m_1, k_1, \dots, m_a, k_a\} \cup \{n_1, l_1, \dots, n_b, l_b\} \cup \{t_1, u_1, \dots, t_c, u_c\}\} \leq n, \\ (t_1, t_2, \dots, t_c) \leq (u_1, u_2, \dots, u_c), \\ \min\{a, b\} \geq 1, \\ c \geq 1 \\ \text{If } b = 0 \quad \text{then} \quad t_h + 1 \leq u_h \quad \text{for some } 1 \leq h \leq c. \end{array} \right. \quad (5.9)$$

First, we will show that $\|\mathcal{R}_2(t)\|_{L^1} \lesssim \varepsilon^2 \langle t \rangle^{-n}$.

Using the third condition in (5.9) and the fact that

$$\max_{0 \leq k \leq n} \left(\|\nabla_v^k Y_{s,t}\|_{L^\infty} + \|\nabla_v^k W_{s,t}\|_{L^\infty} \right) \lesssim \varepsilon \frac{\log(1+s)}{(1+s)^{d-2}},$$

we get

$$\mathcal{R}_2(t, x) \lesssim t^{-n} \int_0^t \int_{\mathbb{R}^d} \varepsilon \frac{\log(1+s)}{(1+s)^{d-2}} \prod_{i=1}^c (s^{t_i} \|\nabla_x^{u_i} E(s)\|_{L^\infty}) |f(\mu)| (V_{s,t}(x, v)) dv ds. \quad (5.10)$$

Now using the second and the forth condition in (5.9), and the decay of $\nabla_x^{u_i} E$, we get

$$s^{t_i} \|\nabla_x^{u_i} E(s)\|_{L^\infty} \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^{d+u_i-t_i}} \lesssim \frac{\varepsilon \log(1+s)}{(1+s)^d}.$$

Applying the above inequality to (5.10), we obtain

$$\begin{aligned} \mathcal{R}_2(t, x) &\lesssim t^{-n} \int_0^t \int_{\mathbb{R}^d} \varepsilon^2 \cdot \frac{\log(1+s)}{(1+s)^{d-2}} \cdot \frac{\log(1+s)}{(1+s)^d} |f(\mu)| (V_{s,t}(x, v)) dv ds \\ &\lesssim \varepsilon^2 t^{-n} \int_0^t \int_{\mathbb{R}^d} \frac{\log(1+s)^2}{(1+s)^{2d-2}} |f(\mu)| (V_{s,t}(x, v)) dv ds. \end{aligned}$$

Hence, using the decaying assumption of μ , we obtain

$$\|\mathcal{R}_2(t)\|_{L^1} \lesssim \varepsilon^2 t^{-n} \int_0^t \frac{\log(1+s)^2}{(1+s)^{2d-2}} \left(\int_{\mathbb{R}^d \times \mathbb{R}^d} |f(\mu)| (V_{s,t}(x, v)) dx dv \right) ds \lesssim \varepsilon^2 t^{-n}.$$

The decaying bound for $\|\mathcal{R}_2(t)\|_{L^1}$ is complete.

We split the integral in $\mathcal{R}_2$ into $\int_0^{t/2} + \int_{t/2}^t$, so that $\mathcal{R}_2 = \mathcal{R}_3 + \mathcal{R}_4$, where

$$\begin{cases} \mathcal{R}_3(t, x) &= \frac{1}{t^n} \int_{t/2}^t \int_{\mathbb{R}^d} \{(\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a}\} \cdot \{(\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b}\} \\ &\quad \{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \} f(\mu) dv ds \\ \mathcal{R}_4(t, x) &= \frac{1}{t^n} \int_0^{t/2} \int_{\mathbb{R}^d} \{(\nabla_v^{m_1} Y_{s,t})^{k_1} \dots (\nabla_v^{m_a} Y_{s,t})^{k_a}\} \cdot \{(\nabla_v^{n_1} W_{s,t})^{l_1} \dots (\nabla_v^{n_b} W_{s,t})^{l_b}\} \\ &\quad \{ (s^{t_1} \nabla_x^{u_1} E) \dots (s^{t_c} \nabla_x^{u_c} E) \} f(\mu) dv ds \end{cases}$$

For $\mathcal{R}_3(t, x)$, by the same argument for $\mathcal{R}_1(t, x)$, we have the pointwise bound

$$\begin{aligned} \mathcal{R}_3(t, x) &\lesssim \varepsilon^2 t^{-n} \int_{t/2}^t \frac{\log(1+s)^2}{(1+s)^{2d-2}} \int_{\mathbb{R}^d} |f(\mu)| (V_{s,t}(x, v)) dv ds \\ &\lesssim \varepsilon^2 t^{-n-d} \int_{t/2}^t \frac{\log(1+s)^2}{(1+s)^{d-2}} ds \lesssim \varepsilon^2 t^{-(n+d)} \end{aligned}$$

Thus $\|\mathcal{R}_3(t)\|_{L^\infty} \lesssim \varepsilon^2 t^{-(n+d)}$.

Now to bound $\|\mathcal{R}_4(t)\|_{L^\infty}$, we use the inequalities

$$\max_{0 \leq k \leq n} \left(\langle s \rangle^{d-2} \|\nabla_v^k Y_{s,t}\|_{L^\infty} + \langle s \rangle^{d-1} \|\nabla_v^k W_{s,t}\|_{L^\infty} \right) \lesssim \varepsilon \log(1+s)$$

to get

$$\begin{aligned}
\mathcal{R}_4(t, x) &\lesssim t^{-n} \int_0^{t/2} \int_{\mathbb{R}^d} \left(\frac{\varepsilon \log(1+s)}{(1+s)^{d-2}} \right)^a \left(\frac{\varepsilon \log(1+s)}{(1+s)^{d-1}} \right)^b \\
&\quad \times \prod_{i=1}^c (s^{t_i} |\nabla_x^{u_i} E|(s, X_{s,t}(x, v))) \cdot |f(\mu)|(V_{s,t}(x, v)) dv ds \\
&\lesssim t^{-n} \varepsilon^{a+b} \int_0^{t/2} \int_{\mathbb{R}^d} \frac{(\log(1+s))^{a+b}}{(1+s)^{a(d-2)+b(d-1)}} \\
&\quad \times \prod_{i=1}^c \{s^{t_i} |\nabla_x^{u_i} E|(s, X_{s,t}(x, v))\} \cdot |f(\mu)|(V_{s,t}(x, v)) dv ds
\end{aligned}$$

Using the change of variable $v \rightarrow \Psi_{s,t}(x, v)$ so that $X_{s,t}(x, \Psi_{s,t}(x, v)) = x - (t-s)v$, we have

$$\begin{aligned}
\mathcal{R}_4(t, x) &\lesssim t^{-n} \varepsilon^{a+b} \int_0^{t/2} \int_{\mathbb{R}^d} \frac{(\log(1+s))^{a+b}}{(1+s)^{a(d-2)+b(d-1)}} \prod_{i=1}^c \{s^{t_i} |\nabla_x^{u_i} E|(s, x - (t-s)v)\} \\
&\quad \cdot |f(\mu)|(V_{s,t}(x, \Psi_{s,t}(x, v))) |\det(\nabla_v \Psi_{s,t}(x, v))| dv ds
\end{aligned}$$

Now making the change of variables $w = x - (t-s)v$, we have

$$\begin{aligned}
\mathcal{R}_4(t, x) &\lesssim t^{-n} \varepsilon^{a+b} \int_0^{t/2} (t-s)^{-d} \int_{\mathbb{R}^d} \frac{(\log(1+s))^{a+b}}{(1+s)^{a(d-2)+b(d-1)}} \prod_{i=1}^c \{s^{t_i} |\nabla_x^{u_i} E|(s, w)\} \\
&\quad \cdot |f(\mu)| \left(V_{s,t}(x, \Psi_{s,t}(x, \frac{x-w}{t})) \right) dw ds \\
&\lesssim t^{-(n+d)} \varepsilon^{a+b} \int_0^{t/2} \frac{\log(1+s)^{a+b}}{(1+s)^{a(d-2)+b(d-1)}} \cdot \min_{1 \leq i \leq c} \{s^{t_i} \|\nabla_x^{u_i} E(s)\|_{L^1}\} ds \\
&\lesssim t^{-(n+d)} \varepsilon^{a+b} \int_0^{t/2} \frac{\log(1+s)^{a+b}}{(1+s)^{a(d-2)+b(d-1)}} \cdot \min_{1 \leq i \leq c} \left\{ s^{t_i} \frac{\varepsilon \log(1+s)}{(1+s)^{u_i}} \right\} ds \\
&\lesssim t^{-(n+d)} \varepsilon^{a+b+1} \int_0^{t/2} \frac{(\log(1+s))^{a+b+1}}{(1+s)^{\{a(d-2)+b(d-1)+\max_{1 \leq i \leq c}(u_i-t_i)\}}} ds.
\end{aligned}$$

Now using the third condition in (5.9), we get $\varepsilon^{a+b+1} \lesssim \varepsilon^2$ as long as ε is small. Hence we get

$$\|\mathcal{R}_4(t)\|_{L^\infty} \lesssim \varepsilon^2 t^{-(n+d)} \int_0^{t/2} \frac{(\log(1+s))^{a+b+1}}{(1+s)^{\{a(d-2)+b(d-1)+\max_{1 \leq i \leq c}(u_i-t_i)\}}} ds. \quad (5.11)$$

Thus, it suffices to prove that

$$\mathcal{C} = \int_0^{t/2} \frac{(\log(1+s))^{a+b+1}}{(1+s)^{\{a(d-2)+b(d-1)+\max_{1 \leq i \leq c}(u_i-t_i)\}}} ds \lesssim 1.$$

We consider two cases:

Case 1: $b = 0$.

In this case, we use the last condition listed in (5.9) to get $\max_{1 \leq i \leq c} (u_i - t_i) \geq 1$, and hence

$$\mathcal{C} \lesssim \int_0^{t/2} \frac{(\log(1+s))^{a+1}}{(1+s)^{a(d-2)+1}} ds \lesssim \int_0^{t/2} \frac{(\log(1+s))^{a+1}}{(1+s)^{a(d-2)+1}} ds \lesssim 1.$$

Case 2: $b \geq 1$.

In this case, we estimate $\mathcal{C}$ as follows:

$$\mathcal{C} \lesssim \int_0^{t/2} \frac{(\log(1+s))^a}{(1+s)^{a(d-2)}} \cdot \frac{(\log(1+s))^{b+1}}{(1+s)^{b(d-1)}} ds \lesssim \int_0^{t/2} \frac{(\log(1+s))^{b+1}}{(1+s)^{b(d-1)}} ds \lesssim 1.$$

The proof is complete. □

Finally, we give the proof for the main Theorem 2.1.

Proof of Theorem 2.1. Let

$$\mathcal{N}(t) = \sup_{0 \leq s \leq t} \max_{0 \leq k \leq N} \frac{(\langle s \rangle^k \|\nabla_x^k \rho(s)\|_{L^1} + \langle s \rangle^{k+d} \|\nabla_x^k \rho(s)\|_{L^\infty})}{\log(1+s)}.$$

Now we fix a constant $M_0 > 0$, which will be chosen later. We prove that $\mathcal{N}(t) \leq M_0 \varepsilon_0$ for all $t \geq 0$, as long as ε_0 is small enough. Let

$$T_\star = \sup\{T > 0 : \mathcal{N}(t) \leq M_0 \varepsilon_0 \text{ for all } 0 \leq t \leq T\} \quad (5.12)$$

We shall prove that $T_\star = \infty$ by contradiction argument. Assuming that $T_\star < \infty$, we have

$$\mathcal{N}(T_\star) = \varepsilon_0 \quad \text{and} \quad \sup_{0 \leq t \leq T} \mathcal{N}(t) < \varepsilon_0 \quad \text{for all } T < T_\star.$$

For $t \in (0, T_\star)$, by Theorem 3.3 and 5.3, there exists $C_0, C_1 > 0$ such that

$$\mathcal{N}(t) \leq C_0 \|S\|_{Y_t^N} \leq C_0 C_1 (\varepsilon_0 + \varepsilon_0^2)$$

for ε_0 small enough. Let $t \rightarrow T_\star$, we have

$$M_0 \varepsilon_0 \leq C_0 C_1 (\varepsilon_0 + \varepsilon_0^2)$$

which is false, as long as $M_0 > 2C_0 C_1$ and ε_0 small. The proof is complete.

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