The Landau equation as a Gradient Flow

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Abstract

We propose a gradient flow perspective to the spatially homogeneous Landau equation for soft potentials. We construct a tailored metric on the space of probability measures based on the entropy dissipation of the Landau equation. Under this metric, the Landau equation can be characterized as the gradient flow of the Boltzmann entropy. In particular, we characterize the dynamics of the PDE through a functional inequality which is usually referred as the Energy Dissipation Inequality (EDI). Furthermore, analogous to the optimal transportation setting, we show that this interpretation can be used in a minimizing movement scheme to construct solutions to a regularized Landau equation.

1 Introduction

The Landau equation is an important partial differential equation in kinetic theory. It gives a description of colliding particles in plasma physics [37], and it can be formally derived as a limit of the Boltzmann equation where grazing collisions are dominant [16, 44]. Similar to the Boltzmann equation (see [7] for a consistency result and related derivation issues), the rigorous derivation of the Landau equation from particle dynamics is still a huge challenge. For a spatially homogeneous density of particles $f = f_t(v)$ for $t \in (0, \infty), v \in \mathbb{R}^d$ the homogeneous Landau equation reads

$$\partial_t f(v) = \nabla_v \cdot \left(f(v) \int_{\mathbb{R}^d} |v - v_*|^{2+\gamma} \Pi[v - v_*] (\nabla_v \log f(v) - \nabla_{v_*} \log f(v_*)) f(v_*) dv_* \right).$$
(1)

For notational convenience, we sometimes abbreviate $f = f_t(v)$ and $f_* = f_t(v_*)$. We also denote the differentiations by $\nabla = \nabla_v$ and $\nabla_* = \nabla_{v_*}$. The physically relevant parameters are usually d = 2, 3 and $\gamma \ge -d - 1$ with $\Pi[z] = I - \frac{z \otimes z}{|z|^2}$ being the projection matrix onto $\{z\}^{\perp}$. In this paper, for simplicity we will focus in the case d = 3 and vary the weight

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parameter γ , although most of our results are valid in arbitrary dimension. The regime $0 < \gamma < 1$ corresponds to the so-called *hard potentials* while $\gamma < 0$ corresponds to the *soft potentials* with a further classification of $-2 \leq \gamma < 0$ as the moderately soft potentials and $-4 \leq \gamma < -2$ as the very soft potentials. The particular instances of $\gamma = 0$ and $\gamma = -d$ are known as the Maxwellian and Coulomb cases respectively.

The purpose of this work is to propose a new perspective inspired from gradient flows for weak solutions to (1), which is in analogy with the relationship of the heat equation and the 2-Wasserstein metric, see [36, 3]. One of the fundamental steps is to symmetrize the right hand of (1). More specifically, if we consider a test function $\phi \in C_c^{\infty}(\mathbb{R}^d)$ we can formally characterize the equation by

$$\frac{d}{dt} \int_{\mathbb{R}^d} \phi f dv = -\frac{1}{2} \iint_{\mathbb{R}^{2d}} f f_* |v - v_*|^{2+\gamma} (\nabla \phi - \nabla_* \phi_*) \cdot \Pi[v - v_*] (\nabla \log f - \nabla_* \log f_*) dv_* dv, \quad (2)$$

where the change of variables $v \leftrightarrow v_*$ has been exploited. Building our analogy with the heat equation and the 2-Wasserstein distance, we define an appropriate gradient

$$\tilde{\nabla}\phi := |v - v_*|^{1 + \frac{\gamma}{2}} \Pi[v - v_*] (\nabla \phi - \nabla_* \phi_*),$$

so that equation (2) now looks like

$$\frac{d}{dt} \int_{\mathbb{R}^d} \phi f dv = -\frac{1}{2} \iint_{\mathbb{R}^{2d}} f f_* \tilde{\nabla} \phi \cdot \tilde{\nabla} \log f dv_* dv,$$

noting that $\Pi^2 = \Pi$. To highlight the use of this interpretation, we notice that $\tilde{\nabla}\phi = 0$, when we choose as test functions $\phi = 1$, v_i , $|v|^2$ for $i = 1, \ldots, d$ which immediately shows that formally the equation conserves mass, momentum and energy. The action functional defining the Landau metric mimics the Benamou-Brenier formula [6] for the 2-Wasserstein distance, see [23, 24, 26] for other distances defined analogously for nonlinear and non-local mobilities. In fact, the Landau metric is built by considering a minimizing action principle over curves that are solutions to the appropriate continuity equation, that is

$$d_L(f,g) := \min_{\substack{\partial_t \mu + \frac{1}{2}\tilde{\nabla} \cdot (V\mu\mu_*) = 0\\ \mu_0 = f, \, \mu_1 = g}} \left\{ \frac{1}{2} \int_0^1 \iint_{\mathbb{R}^{2d}} |V|^2 \, d\mu(v) d\mu(v_*) dt \right\},\tag{3}$$

where the $\tilde{\nabla}$ is the appropriate divergence; the formal adjoint to the appropriate gradient (see Section 2.1).

Also, we notice that analogously to the heat equation, written as the continuity equation $\partial_t f = \nabla \cdot (f \nabla \log f)$, the Landau equation can be formally re-written as

$$\partial_t f = \frac{1}{2} \tilde{\nabla} \cdot (f f_* \tilde{\nabla} \log f),$$

equivalent to the continuity equation with non-local velocity field given by

$$\begin{cases} \partial_t f + \nabla \cdot (U(f)f) = 0 \\ U(f) := -\int_{\mathbb{R}^d} |v - v_*|^{2+\gamma} \Pi[v - v_*] \left(\nabla \log f - \nabla_* \log f_* \right) f_* dv_* \,. \end{cases}$$
(4)

Considering the evolution of Boltzmann entropy we formally obtain

$$\frac{d}{dt} \int_{\mathbb{R}^d} f \log f dv =: -D(f_t) = -\frac{1}{2} \iint_{\mathbb{R}^{2d}} |\tilde{\nabla} \log f|^2 f f_* dv_* dv \le 0.$$
(5)

In physical terms this is referred to as the *entropy dissipation* or *entropy production* for it formally shows that the entropy functional

$$\mathcal{H}[f] := \int_{\mathbb{R}^d} f \log f dv$$

is non-increasing along the dynamics of the Landau equation. Moreover, by integrating equation (5) in time one formally obtains

$$\mathcal{H}[f_t] + \int_0^t D(f_s) ds = \mathcal{H}[f_0].$$
(6)

In [44], Villani introduced the notion of H-solution, which captures this formal property. Motivated by the physical considerations of certain conserved quantities and entropy dissipation, H-solutions provided a step towards well-posedness of the Landau equation in the soft potential case. One advantage to this approach is that it avoids assuming that the solutions belongs to $L^p(\mathbb{R}^3)$ for p > 1. For moderately soft potentials, the propagation of L^p norms is proven and this is enough to make sense of classical weak solutions [47]. In the very soft potential case, there is no longer a guarantee of L^p propagation due to the singularity of the weight. We refer to [17, Section 1.2] for a heuristic description of this difficulty.

Similar to H-solutions our approach will also be based on the entropy dissipation (6). Following De Giorgi's minimizing movement ideas [2, 3], we characterize the Landau equation by its associated Energy Dissipation Inequality. More specifically, we show that weak solutions to (1) with initial data f_0 are completely determined by the following functional inequality:

$$\mathcal{H}[f_t] + \frac{1}{2} \int_0^t |\dot{f}|^2_{d_L}(s) \, ds + \frac{1}{2} \int_0^t D(f_s) \, ds \le \mathcal{H}[f_0] \qquad \text{for a.e. every } t > 0,$$

where $|\dot{f}|^2_{d_L}(s)$ stands for the metric derivative associated to the Landau metric defined above. Our analysis is also largely inspired by Erbar's approach in viewing the Boltzmann equation as a gradient flow [25] and recent numerical simulations of the homogeneous Landau equation in [15] based on a regularized version of (4). In contrast with the classical 2-Wasserstein metric, one of the main features of the Landau equation (1) and metric (3) is that they are non-local. Hence, the convergence analysis usually relying on convexity and lower-semi continuity needs to be adapted to deal with the non-locality of this equation. In particular, our characterization Theorem 11 is based in using (expected) a-priori estimates to deal with the non-locality through appropriate bounds.

On the other hand, the state of the art related to the uniqueness for the Landau equation depends on the range of values γ may take. In the cases of hard potentials or Maxwellian, the uniqueness theory is very well understood due to Villani and the third author [21, 22, 45]. In the soft potential case, one of the first major contributions to the general theory

of the spatially inhomogeneous Landau equation $(\gamma \geq -3)$ was the global existence and uniqueness result by Guo [35]. This result was achieved in a perturbative framework with high regularity assumptions on the initial data. Through probabilistic arguments, the next major improvement to uniqueness for $\gamma \in (-3, 0)$ came from Fournier and Guérin [27]. Their result established uniqueness in a class of solutions that shrinks as γ decreases towards -3, as more L^p and moments assumptions are needed. In their proof, uniqueness is shown by proving stability with respect to the 2-Wasserstein metric.

Still lots of open questions for the soft potential case remain. In particular, a fundamental question like uniqueness for the Coulomb case is unresolved. To tackle this and other problems an array of novel methods have been employed. Here is an incomplete sample of the contributions made in this direction which highlight the difficulties of the soft potential case [22, 21, 1, 12, 11, 47, 33, 31, 34, 32, 43, 30, 42, 29]. A brief glance at some of these references illustrates the breadth of techniques that have found partial success at answering the open questions; probability-based arguments, kinetic and parabolic theory, and many more.

The purpose of this paper is to bring in another set of techniques to help answer some of these fundamental questions. The gradient flow theory applied to PDEs has flourished in the last decades. In their seminal paper [36], Jordan, Kinderlehrer, and Otto proposed a variational approach (JKO scheme) extended later on to a wide class of PDEs using the optimal transportation distance of probability measures. These results and many more achievements from their contemporaries allowed for novel approaches to questions of existence, uniqueness, convergence to equilibrium, and other aspects of a large class of PDE; we mention [3, 40] for a coherent exposition of these techniques and the relevant literature, even as more advances have been made since then.

The advantage of our variational characterization of the Landau equation is that it unveils new possible routes of showing convergence results for this equation. First of all, it allows for natural regularizations of the Landau equation by taking the steepest descent of regularized entropy functionals instead of the Boltzmann entropy as in [14]. This idea was recently developed in [15] leading to structure preserving particle schemes with good accuracy. We can also consider the framework of convergence of gradient flows based on Γ -convergence introduced in [39, 41] to attack the convergence of these numerical methods [15]. Moreover, this approach is flexible enough to also study the rigorous convergence of the grazing collision limit of the Boltzmann equation to the Landau equation. In this case, we can take advantage of the similar developed framework of Erbar for the Boltzmann equation [25] to set this question in simple terms. Namely, the convergence of the associated metrics and the lower-semicontinuous limit of the dissipations. Being able to do this even at the regularized level would be already a breakthrough in understanding the connection between these equations. Finally, deriving uniqueness from the variational structure is classically done through convexity properties of the entropy functional with respect to the geodesics of the Landau metric. This is another important avenue of research that our work opens.

The plan of this paper is as follows. Section 2 introduces the prerequisites and contains the statements of the main results. We first construct and analyze in Section 3 the Landau metric based on (3). For a regularized problem, Section 4 shows the equivalence between weak solutions and gradient flows, while Section 5 shows the existence of gradient flow solutions via a Minimizing Movement scheme. Finally, we show in Section 6 that a gradient flow solution is equivalent to H-solutions of the Landau equation (1) under some integrability assumptions. Appendix A is devoted to some technical lemmata needed in the proof of the main theorems regarding the chain rule identity behind the definition of weak solutions for the regularized Landau equation.

2 Preliminaries and the main results

We start by introducing the necessary notation and definitions together with a quick overview of gradient flow concepts to make our main results fully self-contained.

2.1 Notations and definitions

We denote

$$a \lesssim b \iff \exists C(\dots) > 0 \text{ s.t. } a \le C(\dots)b.$$

We adopt the Japanese angle bracket notation for a smooth alternative to absolute value

$$\langle v \rangle^2 = 1 + |v|^2, \quad v \in \mathbb{R}^d.$$

For $\epsilon > 0$, we denote our regularization kernel to be an exponential distribution

$$G^{\epsilon}(v) = \epsilon^{-d} G(v/\epsilon), \quad G(v) = C_d \exp(-\langle v \rangle), \quad C_d = \left(\int_{\mathbb{R}^d} \exp(-\langle v \rangle) dv\right)^{-1}.$$

Our results work for some general tailed behaviour in the kernels given by

$$G^{s,\epsilon}(v) = \epsilon^{-d} G^s(v/\epsilon), \quad G^s(v) = C_{s,d} \exp(-\langle v \rangle^s), \quad C_{s,d} = \left(\int_{\mathbb{R}^d} \exp(-\langle v \rangle^s) dv \right)^{-1},$$

for s > 0; we point out some of the limitations and restrictions on s > 0 in the later estimates. We shall refer to $G^{2,\epsilon}$ as the Maxwellian regularization. We denote the space of probability measures over \mathbb{R}^d by $\mathscr{P}(\mathbb{R}^d)$, endowed with the weak topology against bounded continuous functions. We will mostly be dealing with the Lebesgue measure on \mathbb{R}^d as our reference measure which we denote by \mathcal{L} . The subset $\mathscr{P}^a(\mathbb{R}^d) \subset \mathscr{P}(\mathbb{R}^d)$ denotes the set of absolutely continuous probability measures with respect to Lebesgue measure. For p > 0, we also define the probability measures with finite *p*-moments $\mathscr{P}_p(\mathbb{R}^d)$ by

$$\mathscr{P}_p(\mathbb{R}^d) := \left\{ \mu \in \mathscr{P}(\mathbb{R}^d) \, \middle| \, m_p(\mu) := \int_{\mathbb{R}^d} \left\langle v \right\rangle^p d\mu(v) < \infty \right\}$$

Finally, for E > 0, we consider the subset $\mathscr{P}_{p,E}(\mathbb{R}^d) \subset \mathscr{P}_p(\mathbb{R}^d)$ of probability measures with p-moments uniformly bounded by E;

$$\mathscr{P}_{p,E}(\mathbb{R}^d) := \left\{ \mu \in \mathscr{P}_p(\mathbb{R}^d) \, \middle| \, m_p(\mu) \le E \right\}.$$

We denote by \mathcal{M} the space of signed Radon measures on $\mathbb{R}^d \times \mathbb{R}^d$ with the standard weak^{*} topology against the continuous and compactly supported functions of $\mathbb{R}^d \times \mathbb{R}^d$. The

space \mathcal{M}^d is the space of signed *d*-length Radon measures. For T > 0, we will add the time contribution of the measures by denoting \mathcal{M}_T to be the space of signed Radon measures on $\mathbb{R}^d \times \mathbb{R}^d \times [0, T]$ with the usual weak* topology. Similarly, \mathcal{M}_T^d will be the space of signed *d*-length Radon measures on $\mathbb{R}^d \times \mathbb{R}^d \times [0, T]$.

For $\mu \in \mathscr{P}(\mathbb{R}^d)$, we define a family of regularized entropies $\mathcal{H}_{\epsilon}[\mu]$ by

$$\mathcal{H}_{\epsilon}[\mu] := \int_{\mathbb{R}^d} [\mu * G^{\epsilon}](v) \log[\mu * G^{\epsilon}](v) dv$$

which we shall see is well-defined provided μ has a finite moment in Lemma 29. Formally, one can calculate the first variation of this functional in \mathscr{P}_2 as

$$\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu}(v) = G^{\epsilon} * \log[\mu * G^{\epsilon}](v)$$

For a functional $\mathcal{F} : \mathscr{P}^a(\mathbb{R}^d) \to \mathbb{R}$ with first variation $\frac{\delta \mathcal{F}}{\delta f}$, we refer to the \mathcal{F} Landau equation as

$$\partial_t f = \nabla \cdot \left(f \int_{\mathbb{R}^d} f_* |v - v_*|^{2+\gamma} \Pi[v - v_*] \left(\nabla \frac{\delta \mathcal{F}}{\delta f} - \nabla_* \frac{\delta \mathcal{F}_*}{\delta f_*} \right) dv_* \right).$$
(7)

To clarify the meaning of $\nabla \cdot$, for a given test function ϕ and vector-valued test function A, we have

$$\iint_{\mathbb{R}^{2d}} [\tilde{\nabla}\phi](v,v_*) \cdot A(v,v_*) dv_* dv = -\int_{\mathbb{R}^d} \phi(v) [\tilde{\nabla} \cdot A](v) dv.$$

In this way, the \mathcal{F} Landau equation (7) can be concisely written as

$$\partial_t f = \frac{1}{2} \tilde{\nabla} \cdot \left(f f_* \tilde{\nabla} \frac{\delta \mathcal{F}}{\delta f} \right).$$

Note, by formally testing (7) with $\phi = \frac{\delta \mathcal{F}}{\delta f}$, one obtains an analogy of Boltzmann's H-theorem with the functional \mathcal{F} ;

$$\frac{d}{dt}\mathcal{F}[f_t] = -D_{\mathcal{F}}(f_t) := -\frac{1}{2} \iint_{\mathbb{R}^{2d}} ff_* \left| \tilde{\nabla} \frac{\delta \mathcal{F}}{\delta f} \right|^2 dv dv_* \le 0.$$

We will refer to $D_{\mathcal{F}}$ as the \mathcal{F} dissipation. These notations induce our notion of weak solutions to the \mathcal{F} Landau equation (7).

Definition 1 (Weak \mathcal{F} solutions). For T > 0, we say that a curve $f \in C([0, T]; L^1(\mathbb{R}^d))$ is a weak solution to the \mathcal{F} Landau equation (7) if the following hold.

1. $f\mathcal{L}$ is a probability measure with uniformly bounded second moment so that

$$f_t \ge 0, \quad \int_{\mathbb{R}^d} f_t(v) dv = 1, \quad \forall t \in [0, T], \quad \sup_{t \in [0, T]} \int_{\mathbb{R}^d} \langle v \rangle^2 f_t(v) dv < \infty.$$

2. The \mathcal{F} dissipation is time integrable

$$\int_0^T D_{\mathcal{F}}(f_t) dt = \frac{1}{2} \int_0^T \iint_{\mathbb{R}^{2d}} ff_* \left| \tilde{\nabla} \frac{\delta \mathcal{F}}{\delta f} \right|^2 dv dv_* dt < \infty.$$

3. For every test function $\phi \in C_c^{\infty}((0,T) \times \mathbb{R}^d)$, equation (7) is satisfied in weak form

$$\int_0^T \int_{\mathbb{R}^d} \partial_t \phi f_t(v) dv dt = \frac{1}{2} \int_0^T \iint_{\mathbb{R}^{2d}} f f_* \tilde{\nabla} \phi \cdot \tilde{\nabla} \frac{\delta \mathcal{F}}{\delta f} dv dv_* dt.$$

For $\epsilon > 0$, we will refer to the weak \mathcal{H}_{ϵ} solutions as ϵ -solutions and, recalling \mathcal{H} is the Boltzmann entropy, we will refer to weak \mathcal{H} solutions as just weak solutions or H-solutions.

2.2 Quick review of gradient flow theory

We recall the basic definitions of gradient flow theory that can be found in more generality in [3, Chapter 1]. Throughout, (X, d) denotes a complete (pseudo)-metric space X with (pseudo)-metric d. Points $a < b \in \mathbb{R}$ will refer to endpoints of some interval. $F : X \to$ $(-\infty, \infty]$ will denote a proper function.

Definition 2 (Absolutely continuous curve). A function $\mu : t \in (a, b) \mapsto \mu_t \in X$ is said to be an *absolutely continuous curve* if there exists $m \in L^2(a, b)$ such that for every $s \leq t \in (a, b)$

$$d(\mu_t, \mu_s) \le \int_s^t m(r) dr.$$

Among all possible functions m in Definition 2, one can make the following minimal selection.

Definition 3 (Metric derivative). For an absolutely continuous curve $\mu : (a, b) \to X$, we define its *metric derivative* at every $t \in (a, b)$ by

$$|\dot{\mu}|(t) := \lim_{h \to 0} \frac{d(\mu_{t+h}, \mu_t)}{|h|}.$$

Further properties of the metric derivative can be found in [3, Theorem 1.1.2].

Definition 4 (Strong upper gradient). The function $g : X \to [0, \infty]$ is a strong upper gradient with respect to F if for every absolutely continuous curve $\mu : t \in (a, b) \mapsto \mu_t \in X$ we have that $g \circ \mu : (a, b) \to [0, \infty]$ is Borel and the following inequality holds

$$|F[\mu_t] - F[\mu_s]| \le \int_s^t g(\mu_r) |\dot{\mu}|(r) dr, \quad \forall a < s \le t < b.$$

Using Young's inequality and moving everything to one side, the inequality in Definition 4 implies

$$F[\mu_t] - F[\mu_s] + \frac{1}{2} \int_s^t g(\mu_r)^2 dr + \frac{1}{2} \int_s^t |\dot{\mu}|^2(r) dr \ge 0, \quad \forall a < s \le t < b.$$

If the reverse inequality also holds, one obtains the stronger Energy Dissipation *Equality*. This leads to our notion of gradient flows.

Definition 5 (Curve of maximal slope). An absolutely continuous curve $\mu : (a, b) \to X$ is said to be a *curve of maximal slope* for F with respect to its strong upper gradient $g: X \to [0, \infty]$ if $F \circ \mu : (a, b) \to [0, \infty]$ is non-increasing and the following inequality holds

$$F[\mu_t] - F[\mu_s] + \frac{1}{2} \int_s^t g(\mu_r)^2 dr + \frac{1}{2} \int_s^t |\dot{\mu}|^2(r) dr \le 0, \quad \forall a < s \le t < b.$$

F has the following natural candidates for upper gradient.

Definition 6 (Slopes). We define the *local slope of* F by

$$|\partial F|(\mu) := \limsup_{\nu \to \mu} \frac{(F(\nu) - F(\mu))^+}{d(\nu, \mu)}.$$

The superscript '+' refers to the positive part. The relaxed slope of F is given by

$$|\partial^{-}F|(\mu) := \inf\{\liminf_{n \to \infty} |\partial F|(\mu_n) : \mu_n \to \mu, \sup_{n \in \mathbb{N}} (d(\mu_n, \mu), F(\mu_n)) < +\infty\}.$$

2.3 Main results

In order to understand the Landau equation as a gradient flow, we need to clarify what type of object the corresponding metric is.

Theorem 7 (Distance on $\mathscr{P}_{2,E}(\mathbb{R}^d)$). The (pseudo)-metric d_L on $\mathscr{P}_{2,E}(\mathbb{R}^d)$, satisfies:

- *d_L*-convergent sequences are weakly convergent.
- *d_L*-bounded sets are weakly compact.
- The map $(\mu_0, \mu_1) \mapsto d_L(\mu_0, \mu_1)$ is weakly lower semicontinuous.
- For any $\tau \in \mathscr{P}_2(\mathbb{R}^d)$ the subset $\mathscr{P}_{\tau}(\mathbb{R}^d) := \left\{ \mu \in \mathscr{P}_{2,m_2(\tau)}(\mathbb{R}^d) \, | \, d_L(\mu,\tau) < \infty \right\}$ is a complete geodesic space.

The content of this theorem is essentially that our new proposed distance actually provides a meaningful topological structure on $\mathscr{P}_{2,E}(\mathbb{R}^d)$. Furthermore, the connection to ϵ solutions of Landau is established when considering the previous notions of slope and upper gradient with respect to d_L .

Theorem 8 (Epsilon equivalence). Fix any $\epsilon, E > 0, \gamma \in [-4, 0]$. Assume that a curve $\mu : [0,T] \to \mathscr{P}_{2,E}(\mathbb{R}^d)$ has a density $\mu_t = f_t \mathcal{L}$. Then μ is a curve of maximal slope for \mathcal{H}_{ϵ} with respect to its upper gradient $\sqrt{D_{\mathcal{H}_{\epsilon}}}$ if and only if its density f is an ϵ -solution to the Landau equation.

From the numerical perspective, we can also construct ϵ -solutions using the JKO scheme (see Section 5) which is the following

Theorem 9 (Existence of curves of maximal slope). For any $\epsilon, E > 0, \gamma \in [-4, 0]$, and initial data $\mu_0 \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, there exists a curve of maximal slope in $\mathscr{P}_{2,E}(\mathbb{R}^d)$ for \mathcal{H}_{ϵ} with respect to its upper gradient $\sqrt{D_{\mathcal{H}_{\epsilon}}}$. **Remark 10.** The choice of an exponential convolution kernel G^{ϵ} is perhaps unnatural compared to the Maxwellian regularization $G^{2,\epsilon}$ for the regularized entropy \mathcal{H}_{ϵ} . We discuss in more detail the estimates that fail using $G^{2,\epsilon}$ in Remark 32 as it pertains to Theorem 8. With respect to Theorem 9, the general construction of *some* curve can be done even with the Maxwellian regularization. However, due to the same lack of estimates, this curve might not be a curve of maximal slope with respect to $\sqrt{D_{\mathcal{H}_{\epsilon}}}$. This is discussed in Remark 36.

Motivated by recent numerical experiments [15], Theorems 8 and 9 provide the theoretical basis to this ϵ approximated Landau equation. In the limit $\epsilon \to 0$, more assumptions are required.

Theorem 11 (Full equivalence). We fix d = 3 and $\gamma \in (-3, 0]$. Suppose that for some T > 0, a curve $\mu : [0, T] \to \mathscr{P}(\mathbb{R}^3)$ has a density $\mu_t = f_t \mathcal{L}$ that satisfies the following set of assumptions

(A1) (Moments and L^p) Assume that for some $0 < \eta \leq \gamma + 3$, we have

$$\langle v \rangle^{2-\gamma} f_t(v) \in L^{\infty}_t(0,T; L^1_v \cap L^{\frac{3-\eta}{3+\gamma-\eta}}_v(\mathbb{R}^3)).$$

(A2) (Finite entropy) We assume that the the entropy is bounded in time

$$\mathcal{H}(f_t) = \int_{\mathbb{R}^3} f_t \log f_t \in L_t^{\infty}(0, T).$$

(A3) (Finite entropy-dissipation) We assume that the entropy-dissipation of f is integrable in time

$$D(f_t) = D_{\mathcal{H}}(f_t) = \frac{1}{2} \iint_{\mathbb{R}^6} ff_* \left| \tilde{\nabla} \frac{\delta \mathcal{H}}{\delta f} \right|^2 dv dv_* = \frac{1}{2} \iint_{\mathbb{R}^6} ff_* |v - v_*|^{\gamma+2} |\Pi[v - v_*] (\nabla \log f - \nabla_* \log f_*)|^2 dv dv_* \in L^1_t(0, T)$$

Then μ is a curve of maximal slope for \mathcal{H} with respect to its upper gradient \sqrt{D} if and only if its density f is a weak solution of the Landau equation.

Remark 12. When $\gamma \in [-2, 0]$, it is known that for suitable initial data (lying in weighted L^p spaces for p large enough and for a sufficient power-like weight), weak solutions of Landau equation satisfying (A1)–(A3) are known to exist (and to be strong and unique under extra conditions). We refer to [47], and Appendix B of [18] when $\gamma > -2$, for details.

When $\gamma \in (-3, -2)$, Assumption (A1) is not known to hold for global weak solutions with large initial data. Solutions satisfying (A1)–(A3) are nevertheless known to exist for initial data close to equilibrium (cf. [35], in a much larger spatially inhomogeneous context), or in the Coulomb case $\gamma = -3$ (in that case $\frac{3-\eta}{3+\gamma-\eta}$ being replaced by ∞) for large initial data, but on specific intervals of times only ([20, 4]). It is an open problem to find the range of values γ under which we can show the existence of curves of maximal slope for the original Landau equation (1), or equivalently, contructing solutions of the original Landau equation passing $\epsilon \to 0$ in Theorem 9. Some of the difficulties to achieve this result are the propagation of moments for the regularized Landau equation uniformly in ϵ and the compactness of sequences with bounded in ϵ regularized entropy dissipation $D_{\mathcal{H}_{\epsilon}}$. The rest of this work is devoted to show the main four theorems in the next four sections.

3 The Landau metric d_L

Our approach to defining the distance d_L mentioned in Theorem 7 closely follows the dynamic formulation of transport distances originally due to Benamou and Brenier [6] and further extended by Dolbeault, Nazaret, and Savaré [23]. We also refer the reader to Erbar [25] for a similar approach.

3.1 Grazing continuity equation

We consider for $\gamma \in [-4, 0]$ the grazing continuity equation:

$$\partial_t \mu_t + \frac{1}{2} \tilde{\nabla} \cdot M_t = 0, \quad \text{in } (0,T) \times \mathbb{R}^d,$$
(8)

which is interpreted in the sense of distributions. For every $\phi \in C_c^{\infty}((0,T) \times \mathbb{R}^d)$, we have

$$\int_0^T \int_{\mathbb{R}^d} \partial_t \phi(t, v) d\mu_t(v) dt + \frac{1}{2} \int_0^T \iint_{\mathbb{R}^{2d}} [\tilde{\nabla}\phi](t, v, v_*) dM_t(v, v_*) dt = 0.$$

Equivalently, for $\zeta \in C_c^{\infty}(\mathbb{R}^d)$,

$$\frac{d}{dt} \int_{\mathbb{R}^d} \zeta(v) d\mu_t(v) = \frac{1}{2} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \zeta(v, v_*) dM_t(v, v_*).$$
(9)

The curves $(\mu_t)_{t \in [0,T]}, (M_t)_{t \in [0,T]}$ are Borel families of measures belonging to \mathcal{M}_+ and \mathcal{M}^d respectively. We will refer to μ from the pair as a *curve* and M as a *grazing rate*. For some regularity properties, we will also need to assume the following moment condition

$$\int_{0}^{T} \iint_{\mathbb{R}^{2d}} (1+|v|+|v_{*}|) d|M_{t}|(v,v_{*}) dt < \infty.$$
(10)

We first establish some a-priori properties of solutions to the grazing continuity equation.

Lemma 13 (Continuous representative). For families (μ_t) , (M_t) satisfying the grazing continuity equation and the finite moment condition (10), there exists a unique weakly^{*} continuous representative curve $(\tilde{\mu}_t)_{t \in [0,T]}$ such that $\tilde{\mu}_t = \mu_t$ a.e. $t \in [0,T]$. Furthermore, for any $\phi \in C_c^{\infty}((0,T) \times \mathbb{R}^d)$ and any $t_0, t_1 \in [0,T]$, we have the following formula

$$\int_{\mathbb{R}^d} \phi_{t_1} d\tilde{\mu}_{t_1} - \int_{\mathbb{R}^d} \phi_{t_0} d\tilde{\mu}_{t_0} = \int_{t_0}^{t_1} \int_{\mathbb{R}^d} \partial_t \phi d\mu_t dt + \frac{1}{2} \int_{t_0}^{t_1} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \phi dM_t dt.$$

Proof. This proof is nearly identical to [3, Lemma 8.1.2]. There, it was crucial to estimate the distributional time derivative of $t \mapsto \mu_t$. We perform the analogous estimate here to highlight the difference in our context. Fix $\zeta \in C_c^{\infty}(\mathbb{R}^d)$ and consider the map

$$t \in (0,T) \mapsto \mu_t(\zeta) = \int_{\mathbb{R}^d} \zeta(v) d\mu_t(v) \in \mathbb{R}.$$

According to (9), the distributional time derivative is

$$\dot{\mu}_t(\zeta) = \frac{1}{2} \iint_{\mathbb{R}^{2d}} \tilde{\nabla}\zeta dM_t(v, v_*) = \frac{1}{2} \iint_{\mathbb{R}^{2d}} |v - v_*|^{1 + \frac{\gamma}{2}} \Pi[v - v_*] (\nabla\zeta - \nabla_*\zeta_*) dM_t(v, v_*).$$

Using the moment condition (10) and a mean-value estimate for $\gamma \in [-4, -2)$, we have the following estimates depending on $\gamma \in [-4, 0]$,

$$|\dot{\mu}_t(\zeta)| \lesssim \begin{cases} \sup_{w \in \mathbb{R}^d} |\nabla\zeta(w)| \iint_{\mathbb{R}^{2d}} (1+|v|+|v_*|) d|M_t|(v,v_*) & \gamma \in [-2,0] \\ \frac{1}{2} \sup_{w \in \mathbb{R}^d} |D^2\zeta(w)| \iint_{\mathbb{R}^{2d}} (1+|v|+|v_*|) d|M_t|(v,v_*) & \gamma \in [-4,-2) \end{cases}$$

The rest of the proof proceeds as in [3, Lemma 8.1.2] using the C^2 -norm of ζ for the soft potentials $\gamma \in [-4, -2)$ as opposed to their C^1 control of ζ .

Define

$$m(\mu_t) := \int_{\mathbb{R}^d} v d\mu_t(v), \quad E(\mu_t) := \int_{\mathbb{R}^d} |v|^2 d\mu_t(v)$$

Lemma 14 (Conservation lemma). Fix $\gamma \in [-4, 0]$ and let $(\mu_t)_{t \in [0,T]}, (M_t)_{t \in [0,T]}$ be Borel families of measures in \mathcal{M}_+ , \mathcal{M}^d respectively satisfying (8) and the moment condition (10). Assume further that $(\mu_t)_{t \in [0,T]}$ is weakly* continuous with respect to t. We have that mass and momentum are conserved;

$$\mu_t(\mathbb{R}^d) = \mu_0(\mathbb{R}^d), \quad m(\mu_t) = m(\mu_0), \quad \forall t \in [0, T].$$

In the case $\gamma \in [-4, -2]$ we have that the energy is conserved;

$$E(\mu_t) = E(\mu_0), \quad \forall t \in [0, T].$$

Proof. We show the proof of the conservation of energy for $\gamma \in [-4, -2]$. We consider a fixed $\varphi \in C_c^{\infty}(B_2)$ which satisfies

$$0 \le \varphi \le 1$$
 and $\varphi(v) = 1$ in B_1 .

We denote

$$\varphi_R(v) = \varphi(v/R).$$

Using the grazing continuity equation, we have, recalling $w(|v - v_*|) = |v - v_*|^{1+\frac{\gamma}{2}}$, that

$$\int_{\mathbb{R}^d} |v|^2 \varphi_R(v) \, d\mu_t(v) - \int_{\mathbb{R}^d} |v|^2 \varphi_R(v) \, d\mu_0(v)$$

=
$$\int_0^t \iint_{\mathbb{R}^{2d}} w \Pi \left(v \varphi_R(v) + |v|^2 \frac{\nabla \varphi(v/R)}{R} - v_* \varphi_R(v_*) - |v_*|^2 \frac{\nabla \varphi(v_*/R)}{R} \right) \, dM_s(v, v_*) ds$$
(11)

Estimating, using that $\Phi_R(v)$ we use the cancelation from the projection Π to obtain

$$\begin{aligned} \left| \int_0^t \iint_{\mathbb{R}^{2d}} w \Pi \left(v \varphi_R(v) - v_* \varphi_R(v_*) \right) \, dM_s \right| &\leq \int_0^t \iint_{(B_R \times B_R)^c} w \left| v \varphi_R(v) - v_* \varphi_R(v_*) \right| \, d|M_s| \\ &\lesssim \int_0^t \iint_{(B_R \times B_R)^c} 1 + |v| + |v_*| \, d|M_s|, \end{aligned}$$

where we have used $\gamma \in [-4, -2]$ to bound

$$w |v\varphi_R(v) - v_*\varphi_R(v_*)| \lesssim \begin{cases} 1 & |v - v_*| \le 1 \\ |v| + |v_*| & |v - v_*| \ge 1 \end{cases}$$

Similarly, using that $\nabla \phi_R$ is supported in $B_{2R} \setminus B_R$ and that $\left| D\left\{ |v|^2 \frac{\nabla \varphi(v/R)}{R} \right\} \right| \lesssim 1$, we obtain that

$$\left| \int_0^t \iint_{\mathbb{R}^{2d}} w \Pi \left(|v|^2 \frac{\nabla \varphi(v/R)}{R} - |v_*|^2 \frac{\nabla \varphi(v_*/R)}{R} \right) dM_s \right| \lesssim \iint_{(B_R \times B_R)^c} 1 + |v| + |v_*| d|M_s|$$

where we have controlled the difference with a mean-value type estimate. From the previous bounds, we can use hypothesis (10) to take $R \to \infty$ in (11) and obtain the conservation of energy

$$\int_{\mathbb{R}^d} |v|^2 \varphi_R(v) \ d\mu_t(v) = \int_{\mathbb{R}^d} |v|^2 \varphi_R(v) \ d\mu_0(v).$$

The proofs for conservation of mass and momentum involve testing the grazing continuity equation against ϕ_R and $v_i\phi_R$ respectively where v_i is the *i*-th component of v. For these statements, the case $\gamma \in [-4, -2]$ follows the same as just presented. For $\gamma \in [-2, 0]$, the estimates can be more blunt since the weight is no longer singular.

Remark 15. Note that as γ increases into the range (-2, 0], the weight function w starts adding growth so the mean-value type argument in Lemma 14 no longer helps unless more moments of M are assumed than (10). Due to the conservation of mass, the unique weakly* continuous representative $(\tilde{\mu}_t)$ of Lemma 13 has the additional property of being weakly continuous in the context of $\mathscr{P}(\mathbb{R}^d)$.

Based on the previous results, we propose the following definition.

Definition 16 (Grazing continuity equation). For some terminal time T > 0, we define \mathcal{GCE}_T to be the set of pairs of measures $(\mu_t, M_t)_{t \in [0,T]}$ satisfying the following:

- 1. $\mu_t \in \mathscr{P}(\mathbb{R}^d)$ is weakly continuous with respect to $t \in [0, T]$. $(M_t)_{t \in [0, T]}$ is a family of Borel measures belonging to \mathcal{M}^d .
- 2. We have the moment bound

$$\int_0^T \iint_{\mathbb{R}^{2d}} (1+|v|+|v_*|) d|M_t|(v,v_*) dt < \infty.$$

3. The grazing continuity equation (8) is satisfied in the distributional sense. That is, for every $\phi \in C_c^{\infty}((0,T) \times \mathbb{R}^d)$,

$$\int_0^T \int_{\mathbb{R}^d} \partial_t \phi d\mu_t dt + \frac{1}{2} \int_0^T \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \phi dM_t dt = 0,$$

or equivalently for every $\zeta \in C_c^{\infty}(\mathbb{R}^d)$,

$$\frac{d}{dt} \int_{\mathbb{R}^d} \zeta(v) d\mu_t(v) = \frac{1}{2} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \zeta(v, v_*) dM_t(v, v_*).$$

For fixed probability measures λ, ν , we may also specify the subset $\mathcal{GCE}(\lambda, \nu)$ as those pairs $(\mu, M) \in \mathcal{GCE}_T$ such that $\mu_0 = \lambda$, $\mu_T = \nu$. For E > 0, we will speak of curves $(\mu, M) \in \mathcal{GCE}_T^{2,E}$ such that

$$m_2(\mu_t) = \int_{\mathbb{R}^d} \langle v \rangle^2 d\mu_t(v) \le E, \quad \forall t \in [0, T].$$

3.2 Action of a curve

In this section, we construct the action of a curve under the grazing continuity equation. We introduce the following function $\alpha : \mathbb{R}^d \times \mathbb{R}_{\geq 0} \to [0, \infty]$ by

$$\alpha(u,s) := \begin{cases} \frac{|u|^2}{2s}, & s \neq 0\\ 0, & s = 0, u = 0\\ \infty, & s = 0, u \neq 0 \end{cases}$$

Lemma 17. α is lower semi-continuous (lsc), convex, and positively 1-homogeneous.

For fixed $\mu \in \mathscr{P}(\mathbb{R}^d), M \in \mathcal{M}^d$, we define $\mu^1 \in \mathscr{P}(\mathbb{R}^d \times \mathbb{R}^d)$ by

$$\mu^{1}(dv, dv_{*}) := \mu(dv)\mu(dv_{*}).$$

Consider $\tau \in \mathcal{M}$ given by $\tau = \mu^1 + |\mathcal{M}|$ and the decompositions $\mu^1 = f^1 \tau$ and $\mathcal{M} = N \tau$. We define the action functional as

$$\mathcal{A}(\mu, M) := \iint_{\mathbb{R}^{2d}} \alpha(N, f^1) d\tau.$$
(12)

This is well-defined by the 1-homogeneity of α . The following lemma establishes a more concrete expression for the action functional.

Lemma 18. Let $\mu \in \mathscr{P}(\mathbb{R}^d)$ be absolutely continuous with respect to \mathcal{L} and $\mu = f\mathcal{L}$. Let $M \in \mathcal{M}^d$ be given such that $\mathcal{A}(\mu, M) < \infty$. Then, M is absolutely continuous with respect to ff_*dvdv_* given by density $U : \mathbb{R}^d \times \mathbb{R}^d \to \mathbb{R}^d$ such that $M = ff_*Udvdv_* = mdvdv_*$ and

$$\mathcal{A}(\mu, M) = \frac{1}{2} \iint_{\mathbb{R}^{2d}} ff_* |U|^2 dv dv_* = \frac{1}{2} \iint_{\mathbb{R}^{2d}} \frac{|m|^2}{ff_*} dv dv_*.$$

Proof. The proof is identical to [25, Lemma 3.6] up to appropriate modifications.

Lemma 19 (Lower semi-continuity of action functional). The action functional \mathcal{A} as defined in (12) is lower semi-continuous in both arguments. Specifically, if $\mu_n \rightharpoonup \mu$ weakly in $\mathscr{P}(\mathbb{R}^d)$ and $M_n \stackrel{*}{\rightharpoonup} M$ weakly* in \mathcal{M}^d , we have

$$\mathcal{A}(\mu, M) \le \liminf_{n \to \infty} \mathcal{A}(\mu_n, M_n).$$

Proof. This result is an application of the general lsc result in [8, Theorem 3.4.3] since α satisfies the required convexity, lsc, and homogeneity assumptions by Lemma 17.

Another useful property of the action functional is the compactness provided by bounded action. We first state

Lemma 20. Let $F : \mathbb{R}^{2d} \to [0,\infty]$ be measurable and fix any $\mu \in \mathscr{P}(\mathbb{R}^d)$, $M \in \mathcal{M}^d$. We have the following bound:

$$\iint_{\mathbb{R}^{2d}} F(v, v_*) d|M|(v, v_*) \le \sqrt{2} \mathcal{A}(\mu, M)^{\frac{1}{2}} \left(\iint_{\mathbb{R}^{2d}} F(v, v_*)^2 d\mu(v) d\mu(v_*) \right)^{\frac{1}{2}}$$
(13)

Proof. This proof follows [25, Lemma 3.8]. We provide the simple argument by Cauchy-Schwarz for completeness. By considering $\tau = \mu \otimes \mu + |M|$, we estimate

$$\begin{split} \iint_{\mathbb{R}^{2d}} Fd|M|(v,v_*) &\leq \iint_{\mathbb{R}^{2d}} F\left|\frac{dM}{d\tau}\right| d\tau(v,v_*) = \iint_{\mathbb{R}^{2d}} F\left(\left|\frac{dM}{d\tau}\right| \middle/ \sqrt{2\frac{d\mu \otimes \mu}{d\tau}}\right) \sqrt{2\frac{d\mu \otimes \mu}{d\tau}} d\tau \\ &\leq \left(\iint_{\mathbb{R}^{2d}} \alpha \left(\frac{dM}{d\tau}, \frac{d\mu \otimes \mu}{d\tau}\right) d\tau\right)^{\frac{1}{2}} \left(\iint_{\mathbb{R}^{2d}} 2F^2 d\mu \otimes \mu\right)^{\frac{1}{2}} \\ &= \sqrt{2}\mathcal{A}(\mu,M)^{\frac{1}{2}} \left(\iint_{\mathbb{R}^{2d}} F(v,v_*)^2 d\mu(v) d\mu(v_*)\right)^{\frac{1}{2}}. \end{split}$$

Remark 21. Suppose we have $\mu_t \in \mathscr{P}(\mathbb{R}^d)$ such that

$$m_2(\mu_t) = \int_0^T \int_{\mathbb{R}^d} |v|^2 \ d\mu_t(v) dt < \infty,$$

then for $M \in \mathcal{M}_T^d$ the previous estimate (13) yields

$$\int_{0}^{T} \iint_{\mathbb{R}^{2d}} 1 + |v| + |v_{*}|d|M_{t}|(v,v_{*})dt \lesssim \int_{0}^{T} \mathcal{A}(\mu_{t},M_{t})^{\frac{1}{2}} \left(1 + 2\int_{\mathbb{R}^{d}} |v|^{2} d\mu_{t}\right)^{\frac{1}{2}} dt.$$
(14)

Therefore, if the integral in time of the second moment of μ is bounded, then M satisfies the moments conditions (10) and the energy is conserved (14). In the sequel, we will be considering curves that have bounded second moment which guarantee (14). **Proposition 22.** Let $(\mu_t^n, M_t^n)_n$ be a sequence in \mathcal{GCE}_T such that $(\mu_0^n)_n$ is tight and we have the following uniform bounds

$$\sup_{n\in\mathbb{N}}\int_0^T\int_{\mathbb{R}^d}|v|^2\ d\mu_t^n dt<\infty\qquad and\qquad \sup_{n\in\mathbb{N}}\int_0^T\mathcal{A}(\mu_t^n,M_t^n)\ dt<\infty.$$
(15)

Then, there exists $(\mu_t, M_t) \in \mathcal{GCE}_T$ such that, possibly after extracting a subsequence, we have the following convergences

$$\mu_t^n \rightharpoonup \mu_t \quad weakly \ in \ \mathscr{P}(\mathbb{R}^d), \quad \forall t \in [0,T]$$
$$M_t^n dt \stackrel{*}{\rightharpoonup} M_t dt \quad weakly^* \ in \ \mathcal{M}_T^d$$

Furthermore, along this subsequence we have the following lower semi-continuity

$$\int_0^T \mathcal{A}(\mu_t, M_t) \ dt \le \liminf_{n \to \infty} \int_0^T \mathcal{A}(\mu_t^n, M_t^n) \ dt.$$

Sketch proof. This result follows from a similar proof to [23, Lemma 4.5] and [25, Proposition 3.11] which we sketch. The second moment bound for μ^n in (15) produces a limit μ . The bounded action in (15) and the estimate (14) produce a limit $M_t dt$ for a subsequence of $M_t^n dt$. The lower semi-continuity follows from Fatou's lemma and Lemma 19.

3.3 Properties of the Landau metric

We define the distance, d_L induced by the action functional on $\mathscr{P}_{2,E}(\mathbb{R}^d)$. Throughout, we will be working in the grazing continuity equation space defined earlier by $\mathcal{GCE}_T^{2,E}$ for T > 0 some terminal time and E > 0 any second moment bound.

Definition 23. For $\lambda, \nu \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ we define the (square of the) Landau distance by

$$d_L^2(\lambda,\nu) := \inf\left\{ T \int_0^T \mathcal{A}(\mu_t, M_t) dt \, \middle| \, (\mu, M) \in \mathcal{GCE}_T^{2,E}(\lambda,\nu) \right\}.$$
(16)

We have an equivalent characterization of d_L which can be seen in other PDE contexts such as [25, 23].

Lemma 24. Given $\lambda, \nu \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, we have

$$d_L(\lambda,\nu) = \inf\left\{\int_0^T \sqrt{\mathcal{A}(\mu_t, M_t)} dt \,\middle|\, (\mu, M) \in \mathcal{GCE}_T^{2,E}(\lambda,\nu)\right\}.$$
(17)

Proof. This proof uses the same reparameterisation technique in [23, Theorem 5.4]. \Box

Proposition 25 (Minimizing curve). Suppose that $\mu_0, \mu_1 \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ are probability measures such that $d_L(\mu_0, \mu_1) < \infty$. Then there exists a curve $(\mu, M) \in \mathcal{GCE}_1^{2,E}(\mu_0, \mu_1)$ attaining the infimum of (16) (equivalently, also (17)) and $\mathcal{A}(\mu_t, M_t) = d_L^2(\mu_0, \mu_1)$ for almost every $t \in [0, 1]$.

Proof. This result follows from the direct method of calculus of variations where the lower semicontinuity comes from Proposition 22.

Proof of Theorem 7. We prove the statements in exactly the order they are presented in the theorem, starting with the properties of the proposed Landau distance as a metric. The positivity of d_L follows from the positivity of α . We now check that d_L satisfies the properties of a metric.

d_L distinguishes points

Fix $\mu_0, \mu_1 \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, we check that $d_L(\mu_0, \mu_1) = 0 \iff \mu_0 = \mu_1$. Suppose that $d_L(\mu_0,\mu_1) = 0$. By Proposition 25 we can find $(\mu, M) \in \mathcal{GCE}_1^{2,E}(\mu_0,\mu_1)$ which is a minimizing curve and moreover $0 = d_L(\mu_0, \mu_1) = \mathcal{A}(\mu_t, M_t)$ implies M = 0. The grazing continuity equation reduces to $\partial_t \mu_t = 0$ which implies μ_t is constant in time.

The converse statement follows similarly by pairing the constant curve $\mu: t \mapsto \mu_0 = \mu_1$ with the zero measure so that $(\mu, 0) \in \mathcal{GCE}_{1}^{2,E}(\mu_{0}, \mu_{1})$.

Symmetry

Symmetry follows because time can be reversed for every curve. For instance, if $(\mu, M) \in$ $\mathcal{GCE}_T^{2,E}(\mu_0,\mu_1)$, then one can check that the pair

$$\mu^r : t \mapsto \mu(T-t), \quad M^r : t \mapsto -M(T-t)$$

belong to $\mathcal{GCE}_T^{2,E}(\mu_1,\mu_0)$ with the same action. Triangle inequality

We sketch the argument using a glueing lemma as in [23, Lemma 4.4]. Let $\mu^0, \mu^1, \mu^2 \in$ $\mathscr{P}_{2,E}(\mathbb{R}^d)$ be such that $d_L(\mu^0, \mu^1) < \infty$ and $d_L(\mu^1, \mu^2) < \infty$. If not, $d_L(\mu^0, \mu^2) \le d_L(\mu^0, \mu^1) +$ $d_L(\mu^1, \mu^2)$ holds trivially. By Proposition 25, we can find minimizing curves connecting these probability measures

$$\left\{\begin{array}{ll} (\mu^{0\to1},M^{0\to1}) &\in \mathcal{GCE}_1^{2,E}(\mu^0,\mu^1)\\ (\mu^{1\to2},M^{1\to2}) &\in \mathcal{GCE}_1^{2,E}(\mu^1,\mu^2) \end{array}\right\}.$$

Their concatenation from time 0 to 1 is given by

$$\mu_t := \begin{cases} \mu_{2t}^{0 \to 1}, & 0 \le t \le 1/2 \\ \mu_{2(t-1/2)}^{1 \to 2}, & 1/2 \le t \le 1 \end{cases}, \quad M_t := \begin{cases} 2M_{2t}^{0 \to 1}, & 0 \le t \le 1/2 \\ 2M_{2t-1/2}^{1 \to 2}, & 1/2 < t \le 1 \end{cases}$$

One can check that $(\mu, M) \in \mathcal{GCE}_1^{2,E}(\mu^0, \mu^2)$, so it is an admissible competitor in the computation of $d_L(\mu^0, \mu^2)$. By looking at the action on the different time pieces, we obtain

$$d_L(\mu^0, \mu^2) \le \int_0^1 \mathcal{A}(\mu_t, M_t) dt = d_L(\mu^0, \mu^1) + d_L(\mu^1, \mu^2).$$

 $\frac{d_L \text{-}convergence/boundedness implies weak convergence/compactness}{\text{Fix } \mu^n, \mu^\infty \in \mathscr{P}_{2,E} \text{ for } n \in \mathbb{N} \text{ be such that } d_L(\mu^\infty, \mu^n) \to 0 \text{ as } n \to \infty. \text{ By Proposition 25,}$ take minimizing curves $(\nu^n, M^n) \in \mathcal{GCE}_1^{2,E}(\mu^{\infty}, \mu^n)$ such that

$$d_L(\mu^{\infty}, \mu^n) = \mathcal{A}(\nu_t^n, M_t^n), \quad \text{a.e. } t \in [0, 1].$$

By compactness in Proposition 22, there are limits $(\nu, M) \in \mathcal{GCE}_1^{2,E}$ such that $\nu^n \rightharpoonup \nu$ and $M^n \stackrel{*}{\rightharpoonup} M$ up to a subsequence. Moreover, the lower semicontinuity in Proposition 22 gives

$$\mathcal{A}(\nu_t, M_t) \leq \liminf_{n \to \infty} \mathcal{A}(\nu_t^n, M_t^n) = 0,$$

hence M = 0 so that ν is a constant in time. Since $\nu(0) = \mu^{\infty}$, this implies $\mu^{\infty} = \nu(1) = \lim_{n \to \infty} \mu^n$ which establishes the weak convergence.

 $(\mathscr{P}_{\tau}, d_L)$ is a complete geodesic space

We start with the geodesic property from completely analogous arguments to Erbar [25], the remaining statement that \mathscr{P}_{τ} equipped with d_L is a complete geodesic space follows. Fix $\tau \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ with $\mu_0, \mu_1 \in \mathscr{P}_{\tau}$, the triangle inequality ensures $d_L(\mu_0, \mu_1) < \infty$ so Proposition 25 guarantees the existence of a minimizing curve $(\mu, M) \in \mathcal{GCE}_1^{2,E}(\mu_0, \mu_1)$. One easily sees that this also induces a minimizing curve for intermediate times. More precisely, for every $0 \leq r \leq s \leq 1$, we have that $(t \mapsto \mu_{t+r}, t \mapsto M_{t+r}) \in \mathcal{GCE}_{s-r}^{2,E}(\mu_r, \mu_s)$ also minimizes $d_L(\mu_r, \mu_s)$.

To show completeness, let $(\mu^n)_{n\in\mathbb{N}}$ be a Cauchy sequence in \mathscr{P}_{τ} . The sequence is certainly d_L -bounded so by Proposition 22, we can find, up to extraction of a weakly convergent subsequence, $\mu^{\infty} \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ such that $\mu^n \rightharpoonup \mu^{\infty}$ in $\mathscr{P}_{2,E}(\mathbb{R}^d)$. Lower semi-continuity of d_L and the Cauchy property of the subsequence gives

$$d_L(\mu^n, \mu^\infty) \le \liminf_{m \to \infty} d_L(\mu^n, \mu^m) \to 0, \text{ as } n \to \infty.$$

For any $n \in \mathbb{N}$ the triangle inequality gives

$$d_L(\mu^{\infty},\tau) \le d_L(\mu^{\infty},\mu^n) + d_L(\mu^n,\tau) < \infty,$$

So $\mu^{\infty} \in \mathscr{P}_{\tau}$.

Proposition 26 (Metric derivative). A curve $(\mu_t)_{t\in[0,T]} \subset \mathscr{P}_{2,E}(\mathbb{R}^d)$ is absolutely continuous with respect to d_L if and only if there exists a Borel family $(M_t)_{t\in[0,T]}$ belonging to \mathcal{M}_T^d such that $(\mu, M) \in \mathcal{GCE}_T^{2,E}$ with the property that

$$\int_0^T \sqrt{\mathcal{A}(\mu_t, M_t)} dt < \infty.$$

In this equivalence, we have a bound on the metric derivative

$$\lim_{h \downarrow 0} \frac{d_L^2(\mu_{t+h}, \mu_t)}{h^2} =: |\dot{\mu}|^2(t) \le \mathcal{A}(\mu_t, M_t), \quad a.e. \ t \in (0, T).$$

Furthermore, there exists a unique Borel family $(\tilde{M}_t)_{t \in [0,T]}$ belonging to \mathcal{M}^d which is characterized by

$$M_t = U\mu_t \otimes \mu_t \qquad and \qquad U \in T_\mu := \overline{\{\tilde{\nabla}\phi \mid \phi \in C_c^\infty(\mathbb{R}^d)\}}^{L^2(\mu_t \otimes \mu_t)}$$

such that $(\mu, \tilde{M}) \in \mathcal{GCE}_T^E(\mu_0, \mu_T)$ where we have equality:

$$|\dot{\mu}|^2(t) = \mathcal{A}(\mu_t, M_t), \quad a.e. \ t \in (0, T).$$

Proof. The argument follows exactly as in [23, Theorem 5.17].

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4 Energy dissipation equality

The goal in this section is to prove Theorem 8 which states that the notions of gradient flow solutions coincide with ϵ -solutions to the Landau equation. To fix ideas, we recall the regularized entropy functionals acting on probability measures

$$\mathcal{H}_{\epsilon}[\mu] = \int_{\mathbb{R}^d} (\mu * G^{\epsilon})(v) \log(\mu * G^{\epsilon})(v) dv,$$

with $G^{\epsilon}(v)$ given by

$$G^{\epsilon}(v) = \epsilon^{-d} C_d \exp\left\{-\left\langle \frac{v}{\epsilon} \right\rangle\right\}.$$

The crucial ingredient to prove Theorem 8 is the following

Proposition 27 (Chain Rule ϵ). Suppose $(\mu, M) \in \mathcal{GCE}_T^{2,E}$ and

$$\int_0^T \mathcal{A}(\mu_t, M_t) dt < \infty.$$

Then, $\sup_{t \in [0,T]} \mathcal{H}_{\epsilon}[\mu_t] < \infty$ and the 'chain rule' holds

$$\mathcal{H}_{\epsilon}[\mu_{r}] - \mathcal{H}_{\epsilon}[\mu_{s}] = \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \cdot dM_{t} dt, \quad \forall 0 \le s \le r \le T.$$
(18)

Remark 28. Recall the expression for the dissipation

$$D_{\epsilon}[\mu] = \frac{1}{2} \iint_{\mathbb{R}^{2d}} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^2 d\mu(v) d\mu(v_*).$$

Using a time integrated version of Lemma 20, we have the estimate

$$\frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right| \cdot d|M_{t}|(v, v_{*}) dt \leq \int_{s}^{r} \mathcal{A}(\mu_{t}, M_{t})^{\frac{1}{2}} D_{\epsilon}[\mu_{t}]^{\frac{1}{2}} dt.$$

Therefore, under the hypothesis of Proposition 27, we have that

$$|\mathcal{H}_{\epsilon}(\mu_{r}) - \mathcal{H}_{\epsilon}(\mu_{r})| \leq \int_{s}^{r} |\dot{\mu}|(t) D_{\epsilon}[\mu_{t}]^{\frac{1}{2}} dt,$$

which implies that $D_{\epsilon}[\mu_t]^{\frac{1}{2}}$ is a strong upper gradient of \mathcal{H}_{ϵ} , see Definition 4.

Taking Proposition 27 for granted, we can prove Theorem 8.

Proof of Theorem 8. Throughout, $\mu = f\mathcal{L}$ is a curve of probability measures with uniformly bounded second moment.

Weak ϵ -solution \implies Curve of maximal slope

Consider f an ϵ -solution to the Landau equation. Define $m = -ff_*\tilde{\nabla}\frac{\delta \mathcal{H}_{\epsilon}}{\delta f}$ so that the pair of measures $(\mu = f\mathcal{L}, M = m\mathcal{L} \otimes \mathcal{L})$ therefore belong to \mathcal{GCE}_T^E . Indeed, the distributional grazing continuity equation from Definition 16 is precisely the weak ϵ Landau equation. Based on the definition of M and the finite \mathcal{H}_{ϵ} dissipation, we have the bound

$$\int_0^T \mathcal{A}(\mu_t, M_t) dt = \int_0^T D_{\epsilon}(f_t) dt < \infty,$$

which implies the weak continuity of μ . By Proposition 26, we have

$$|\dot{\mu}|^2(t) = \mathcal{A}(\mu_t, M_t) = D_{\epsilon}(f_t) < \infty, \quad \text{a.e. } t \in [0, T].$$

Using Proposition 27, we have for any $0 \le s \le r \le T$

$$\mathcal{H}_{\epsilon}[\mu_r] - \mathcal{H}_{\epsilon}[\mu_s] + \frac{1}{2} \int_s^r D_{\epsilon}(\mu_t) dt + \frac{1}{2} \int_s^r |\dot{\mu}|^2(t) dt \le 0.$$

According to Definition 5, this is the curve of maximal slope property.

Curve of maximal slope \implies weak ϵ -solution

Assume that $\mu = f\mathcal{L}$ is a curve of maximal slope for \mathcal{H}_{ϵ} with respect to the upper gradient $\sqrt{D_{\epsilon}}$. Since μ is absolutely continuous with respect to d_L , Proposition 26 guarantees existence of a unique curve $M : t \in [0,T] \mapsto M_t \in \mathcal{M}^d$ such that $\int_0^T \sqrt{\mathcal{A}(\mu_t, M_t)} dt < \infty$ and $|\dot{\mu}|^2(t) = \mathcal{A}(\mu_t, M_t)$ a.e. $t \in [0,T]$. Furthermore, the pair $(\mu, M) \in \mathcal{GCE}_T^E$. According to Lemma 18, let $M = m\mathcal{L} \otimes \mathcal{L}$ for some measurable function m. We apply the chain rule (18) with Cauchy-Schwarz and Young's inequalities with minus signs in the follow computations.

$$\begin{aligned} \mathcal{H}_{\epsilon}[f_{T}] - \mathcal{H}_{\epsilon}[f_{0}] &= \frac{1}{2} \int_{0}^{T} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta f} \cdot m dv dv_{*} dt \\ &\geq -\frac{1}{2} \int_{0}^{T} \left(\iint_{\mathbb{R}^{2d}} f f_{*} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta f} \right|^{2} dv dv_{*} \right)^{\frac{1}{2}} \left(\iint_{\mathbb{R}^{2d}} \frac{|m|^{2}}{f f_{*}} dv dv_{*} \right)^{\frac{1}{2}} dt \\ &\geq -\frac{1}{2} \int_{0}^{T} \left(\frac{1}{2} \iint_{\mathbb{R}^{2d}} f f_{*} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta f} \right|^{2} dv dv_{*} \right) dt - \frac{1}{2} \int_{0}^{T} \left(\frac{1}{2} \iint_{\mathbb{R}^{2d}} \frac{|m|^{2}}{f f_{*}} dv dv_{*} \right) dt \\ &= -\frac{1}{2} \int_{0}^{T} D_{\epsilon}(f_{t}) dt - \frac{1}{2} \int_{0}^{T} |\dot{f}|^{2}(t) dt. \end{aligned}$$

All the inequalities in the calculations above are actually equalities owing to the fact that μ is a curve of maximal slope. In particular, since we have the equality in the Young's inequality, this implies that $\frac{m}{\sqrt{ff_*}} = -\sqrt{ff_*}\tilde{\nabla}\frac{\delta \mathcal{H}_{\epsilon}}{\delta f}$. As in the previous direction, the weak ϵ Landau equation coincides with the grazing continuity equation when m is equal to $-ff_*\tilde{\nabla}\frac{\delta \mathcal{H}_{\epsilon}}{\delta f}$. \Box

The rest of this section is devoted to proving Proposition 27. We need some lemmata to establish crucial estimates. The following result is a variation of [10, Lemma 2.6].

Lemma 29 (Carlen-Carvalho [10]). Let μ be a probability measure on \mathbb{R}^d with finite second moment/energy, $m_2(\mu) \leq E$ for E > 0. Then, for every $\epsilon > 0$, there exists a constant $C = C(\epsilon, E) > 0$ such that

$$|\log(\mu * G^{\epsilon})(v)| \le C\left\langle \frac{v}{\epsilon} \right\rangle.$$

Proof. Starting with an upper bound, we easily see

$$\mu * G^{\epsilon}(v) = \int_{\mathbb{R}^d} G^{\epsilon}(v - v') d\mu(v') \lesssim_{\epsilon} 1.$$

Turning to the lower bound, we cut off the integration domain to $|v'| \leq R$, for some R > 0 to be chosen later. We estimate, for $\epsilon > 0$ small enough

$$\left\langle \frac{v - v'}{\epsilon} \right\rangle = \sqrt{1 + \left| \frac{v - v'}{\epsilon} \right|^2} \le \sqrt{1 + 2\left| \frac{v}{\epsilon} \right|^2 + 2\left(\frac{R}{\epsilon}\right)^2} \le \sqrt{2}\left(\left\langle \frac{v}{\epsilon} \right\rangle + \left\langle \frac{R}{\epsilon} \right\rangle \right).$$

This is substituted into $G^{\epsilon}(v-v')$ to obtain

$$\mu * G^{\epsilon}(v) \ge \int_{|v'| \le R} G^{\epsilon}(v - v') d\mu(v') \gtrsim_{\epsilon} \exp\left\{-\sqrt{2}\left(\left\langle \frac{v}{\epsilon} \right\rangle + \left\langle \frac{R}{\epsilon} \right\rangle\right)\right\} \int_{|v'| \le R} d\mu(v') d\mu(v') \le \epsilon$$

At this point, we appeal to Chebyshev's inequality to see

$$\int_{|v'| \le R} d\mu(v') = 1 - \int_{|v'| \ge R} d\mu(v') \ge 1 - \frac{1}{R^2} \int_{|v'| \ge R} |v'|^2 d\mu(v').$$

We can now choose, for example, large R such that $1 - \frac{E}{R^2} \ge \frac{1}{2}$ to uniformly lower bound the integral $\int_{|v'| \le R} d\mu(v')$ away from 0 and then conclude the result after applying logarithms. \Box

Lemma 30 (log-derivative estimates). For fixed $\epsilon > 0$ we have the formula

$$\nabla G^{\epsilon}(v) = -\frac{1}{\epsilon} \left\langle \frac{v}{\epsilon} \right\rangle^{-1} G^{\epsilon}(v) \frac{v}{\epsilon}.$$
(19)

For $\mu \in \mathscr{P}(\mathbb{R}^d)$, denoting $\partial^i = \frac{\partial}{\partial v^i}$ and $\partial^{ij} = \frac{\partial^2}{\partial v^i \partial v^j}$, we obtain

$$\left|\nabla \log(\mu * G^{\epsilon})(v)\right| \le \frac{1}{\epsilon}, \quad \left|\partial^{ij} \log(\mu * G^{\epsilon})(v)\right| \le \frac{4}{\epsilon^2}.$$
 (20)

Proof. Equation (19) is a direct computation after noticing

$$\frac{\nabla G^{\epsilon}}{G^{\epsilon}} = \nabla \log G^{\epsilon} = \nabla \left(-\left\langle \frac{v}{\epsilon} \right\rangle + const. \right) = -\frac{1}{\epsilon} \left\langle \frac{v}{\epsilon} \right\rangle^{-1} \frac{v}{\epsilon}.$$

The first order log-derivative estimate of (20) is calculated using formula (19) to obtain

$$\begin{aligned} |\nabla(\mu * G^{\epsilon})(v)| &= |\mu * \nabla G^{\epsilon}(v)| \leq \frac{1}{\epsilon} \int_{\mathbb{R}^d} \left\langle \frac{v - v'}{\epsilon} \right\rangle^{-1} \left| \frac{v - v'}{\epsilon} \right| G^{\epsilon}(v - v') d\mu(v') \\ &\leq \frac{1}{\epsilon} \int_{\mathbb{R}^d} G^{\epsilon}(v - v') d\mu(v') = \frac{1}{\epsilon} (\mu * G^{\epsilon})(v). \end{aligned}$$

For the second order, we first look at $\partial^{ij}\mu * G^{\epsilon}$ which can be computed with the help of (19)

$$\begin{split} |\partial^{ij}\mu * G^{\epsilon}(v)| &= \left| \partial^{i} \left(-\frac{1}{\epsilon} \int_{\mathbb{R}^{d}} \left\langle \frac{v - v'}{\epsilon} \right\rangle^{-1} \frac{v^{j} - v'^{j}}{\epsilon} G^{\epsilon}(v - v') d\mu(v') \right) \right| = \\ \left| \frac{1}{\epsilon^{2}} \int_{\mathbb{R}^{d}} \left(\left\langle \frac{v - v'}{\epsilon} \right\rangle^{-3} \frac{v^{i} - v'^{i}}{\epsilon} \frac{v^{j} - v'^{j}}{\epsilon} + \delta^{ij} \left\langle \frac{v - v'}{\epsilon} \right\rangle^{-1} \\ &- \left\langle \frac{v - v'}{\epsilon} \right\rangle^{-2} \frac{v^{i} - v'^{i}}{\epsilon} \frac{v^{j} - v'^{j}}{\epsilon} \right) G^{\epsilon}(v - v') d\mu(v') \right| \\ &\leq \frac{3}{\epsilon^{2}} \mu * G^{\epsilon}(v). \end{split}$$

Combining this estimate with the previous first order one, we have

$$\left|\partial^{ij}\log(\mu * G^{\epsilon})(v)\right| = \left|\frac{\partial^{ij}\mu * G^{\epsilon}}{\mu * G^{\epsilon}} - \frac{(\partial^{i}\mu * G^{\epsilon})(\partial^{j}\mu * G^{\epsilon})}{(\mu * G^{\epsilon})^{2}}\right| \le \frac{4}{\epsilon^{2}}.$$

Lemma 31. Fix $\epsilon > 0$ and $\gamma \in [-4, 0]$ with $\mu \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ for some E > 0. We have

- 1. <u>Moderately soft case $\gamma \in [-2,0]$:</u> $\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right| = \left|\tilde{\nabla}[G^{\epsilon} * \log(\mu * G^{\epsilon})](v,v_{*})\right| \lesssim_{\epsilon} |v|^{1+\frac{\gamma}{2}} + |v_{*}|^{1+\frac{\gamma}{2}}.$
- 2. Very soft case $\gamma \in [-4, -2]$:

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right| \lesssim_{\epsilon} 1.$$

In particular, it holds

$$\iint_{\mathbb{R}^{2d}} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right|^2 d\mu(v) d\mu(v_*) \le E.$$

Proof. We develop the expression for $\tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu}$ in integral form to be used throughout this proof.

$$\tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} = \tilde{\nabla} G^{\epsilon} * \log(\mu * G^{\epsilon})(v, v_{*})$$

$$= |v - v_{*}|^{1 + \frac{\gamma}{2}} \Pi[v - v_{*}] (\nabla_{v} G^{\epsilon} * \log(\mu * G^{\epsilon})(v) - \nabla_{v_{*}} G^{\epsilon} * \log(\mu * G^{\epsilon})(v_{*})) \qquad (21)$$

$$= |v - v_{*}|^{1 + \frac{\gamma}{2}} \Pi[v - v_{*}] \int_{\mathbb{R}^{d}} G^{\epsilon}(v') \left(\frac{\nabla \mu * G^{\epsilon}}{\mu * G^{\epsilon}} (v - v') - \frac{\nabla \mu * G^{\epsilon}}{\mu * G^{\epsilon}} (v_{*} - v') \right) dv'.$$

1. Moderately soft case $\gamma \in [-2, 0]$: We use (a concave version of) the triangle inequality (valid since $1 + \frac{\gamma}{2} \ge 0$) and the first estimate of (20) to bound the last line of (21)

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right| \leq 2^{1+\frac{\gamma}{2}} (|v|^{1+\frac{\gamma}{2}} + |v_*|^{1+\frac{\gamma}{2}}) \frac{2}{\epsilon} \int_{\mathbb{R}^d} G^{\epsilon}(v') dv' \lesssim_{\epsilon} |v|^{1+\frac{\gamma}{2}} + |v_*|^{1+\frac{\gamma}{2}} dv' \leq_{\epsilon} |v|^{1+\frac{\gamma}{2}} dv' <_{\epsilon} |v|^{1+\frac{\gamma}{2}} dv' <_{\epsilon$$

2. <u>Very soft case $\gamma \in [-4, -2]$:</u> We perform estimates in two cases, the far field $|v - v_*| \ge 1$ and near field $|v - v_*| \le 1$. $|v - v_*| \ge 1$:

In the far field, we have $|v-v_*|^{1+\frac{\gamma}{2}} \leq 1$ hence we can brutally estimate (21) using again the first estimate of (20) to obtain, similar to the moderately soft case, the estimate

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right| \leq \frac{2}{\epsilon}$$

 $\frac{|v - v_*| \le 1}{We \text{ can remove the singularity from the weight with a mean-value estimate and the weight with a mean-value estimate and the weight with$ second estimate of (20)

$$\left|\frac{\nabla\mu * G^{\epsilon}}{\mu * G^{\epsilon}}(v - v') - \frac{\nabla\mu * G^{\epsilon}}{\mu * G^{\epsilon}}(v_* - v')\right| \le \sup_{i,j=1,\dots,d} \left|\left|\partial^i \left(\frac{\partial^j \mu * G^{\epsilon}}{\mu * G^{\epsilon}}\right)\right|\right|_{L^{\infty}} |v - v_*| \le \frac{4}{\epsilon^2} |v - v_*|.$$

Inserting this into (21), we have

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right| \leq \frac{4}{\epsilon^{2}}|v-v_{*}|^{2+\frac{\gamma}{2}}\int_{\mathbb{R}^{d}}G^{\epsilon}(v')dv' \leq \frac{4}{\epsilon^{2}}.$$

Remark 32. Originally, we considered the general family of convolution kernels $G^{s,\epsilon}$ described in Section 2.1. Besides the context of the Landau equation, Lemma 30 (excluding the second order log-derivative estimate) can be generalized to this family of s-order tailed exponential distributions with additional moment assumptions on μ . In particular, equations (19) and (20) (for $s \ge 1$) become

$$\frac{\nabla G^{s,\epsilon}}{G^{s,\epsilon}}(v) = -\frac{s}{\epsilon} \left\langle \frac{v}{\epsilon} \right\rangle^{s-2} \frac{v}{\epsilon}, \quad \frac{|\nabla(\mu * G^{s,\epsilon})|}{\mu * G^{s,\epsilon}}(v) \lesssim \frac{1}{\epsilon^s} \left\langle v \right\rangle^{s-1}.$$

Since Maxwellians are known to be stationary solutions for the Landau equation, we wanted to perform the regularization with s = 2. However, the analogous estimates of Lemma 30 for s = 2 are not sufficient for Lemma 31 in the \mathscr{P}_2 framework. For example, in the moderately soft potential case, the estimate reads

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{2,\epsilon}}{\delta\mu}\right| \lesssim_{\epsilon} \langle v \rangle^{2+\frac{\gamma}{2}} + \langle v_* \rangle^{2+\frac{\gamma}{2}} \notin L^2(\mu \otimes \mu).$$

However, there is one value of $\gamma = -2$ for which the estimates hold when using a Maxwellian regularization kernel $G^{2,\epsilon}$. A restriction to \mathscr{P}_4 resolves the issue mentioned above for the moderately soft potential case, but then a fourth moment propagation is needed which we did not pursue. A similar issue is present in the very soft potential case.

Proof of Proposition 27. To prove equation (18), our strategy is to regularize the pair (μ, M) in time with parameter $\delta > 0$ and differentiate the regularization. Then we obtain uniform bounds in δ needed to take the limit $\delta \to 0$.

Finite regularized entropy

We have the following chain of inequalities

$$\mathcal{H}_{\epsilon}[\mu_{t}] = \int_{\mathbb{R}^{d}} (\mu_{t} * G^{\epsilon})(v) \log(\mu_{t} * G^{\epsilon})(v) dv \lesssim_{\epsilon, E} \int_{\mathbb{R}^{d}} (\mu_{t} * G^{\epsilon})(v) \langle v \rangle dv \lesssim_{\epsilon} 1 + E.$$

The first inequality comes from Lemma 29 because $\log(\mu_t * G^{\epsilon})$ has linear growth (uniform in time) while in the second inequality, one realises that $\mu_t * G^{\epsilon}$ has as many moments as μ_t with computable constants.

Time regularization with $\delta > 0$

Without loss of generality, let μ be the weakly time continuous representative (Lemma 13) and M be the optimal grazing rate (Proposition 26) achieving the finite distance d_L . We first regularize the pair (μ, M) in time for a fixed parameter $\delta > 0$ as follows. Take $\eta \in C_0^{\infty}(\mathbb{R})$ with the following properties

$$supp \eta \subset (-1, 1), \quad \eta \ge 0, \quad \eta(t) = \eta(-t), \quad \int_{-1}^{1} \eta(t) dt = 1.$$

We define the following measures for $t \in [0, T]$, by taking convex combinations

$$\mu_t^{\delta} := \int_{-1}^1 \eta(t') \mu_{t-\delta t'} dt', \quad M_t^{\delta} := \int_{-1}^1 \eta(t') M_{t-\delta t'} dt'.$$

Here, we constantly extend the measures in time. That is, if $t - \delta t' \in [-\delta, 0]$, we treat $\mu_{t-\delta t'} = \mu_0, M_{t-\delta t'} = 0$. For the other end point, if $t - \delta t' \in [T, T + \delta]$, we set $\mu_{t-\delta t'} = \mu_T, M_{t-\delta t'} = 0$. This transformation is stable so that $(\mu^{\delta}, M^{\delta}) \in \mathcal{GCE}_T$ and in particular, the distributional grazing continuity equation holds

$$\partial_t \mu_t^\delta + \frac{1}{2} \tilde{\nabla} \cdot M_t^\delta = 0.$$

We derive equation (18) using this regularized grazing continuity equation. Consider

$$\mathcal{H}_{\epsilon}[\mu_t^{\delta}] = \int_{\mathbb{R}^d} (\mu_t^{\delta} * G^{\epsilon})(v) \log(\mu_t^{\delta} * G^{\epsilon})(v) dv,$$

which we differentiate with respect to t by appealing to the Dominated Convergence Theorem. Firstly, due to the time regularization, we have

$$\partial_t \left\{ (\mu_t^{\delta} * G^{\epsilon}) \log(\mu_t^{\delta} * G^{\epsilon}) \right\} = \left[(\partial_t \mu_t^{\delta}) * G^{\epsilon} \right] (\log(\mu_t^{\delta} * G^{\epsilon}) + 1).$$

The L_v^1 bound is obtained on the following difference quotient for a fixed time step h > 0

$$\begin{aligned} &\left|\frac{1}{h}[(\mu_{t+h}^{\delta}*G^{\epsilon})\log(\mu_{t+h}^{\delta}*G^{\epsilon})-(\mu_{t}^{\delta}*G^{\epsilon})\log(\mu_{t}^{\delta}*G^{\epsilon})]\right|\\ &\leq \frac{1}{h}\left|(\mu_{t+h}^{\delta}*G^{\epsilon})-(\mu_{t}^{\delta}*G^{\epsilon})\right|\sup_{s\in[t,t+h]}\left|\log(\mu_{s}^{\delta}*G^{\epsilon})+1\right|.\end{aligned}$$

where we have used the Mean Value theorem with the chain rule. Applying Lemma 29, we obtain

$$\left|\frac{1}{h}\left[\left(\mu_{t+h}^{\delta} \ast G^{\epsilon}\right)\log\left(\mu_{t+h}^{\delta} \ast G^{\epsilon}\right) - \left(\mu_{t}^{\delta} \ast G^{\epsilon}\right)\log\left(\mu_{t}^{\delta} \ast G^{\epsilon}\right)\right]\right| \lesssim_{\epsilon,E} \frac{1}{h}\left|\left(\mu_{t+h}^{\delta} \ast G^{\epsilon}\right) - \left(\mu_{t}^{\delta} \ast G^{\epsilon}\right)\right|\left\langle v\right\rangle$$

We apply the Mean Value Theorem on the difference quotient again to get

$$\left|\frac{1}{h}\left[\left(\mu_{t+h}^{\delta} \ast G^{\epsilon}\right)\log\left(\mu_{t+h}^{\delta} \ast G^{\epsilon}\right) - \left(\mu_{t}^{\delta} \ast G^{\epsilon}\right)\log\left(\mu_{t}^{\delta} \ast G^{\epsilon}\right)\right]\right| \lesssim_{\delta,\epsilon} ||\eta'||_{L^{\infty}} \left(\mu_{0} \ast G^{\epsilon} + \int_{0}^{T} \mu_{t} \ast G^{\epsilon} dt\right) \langle v \rangle.$$

Since μ has finite second order moments, this last expression belongs to L_v^1 . By the Dominated Convergence Theorem,

$$\frac{d}{dt}\mathcal{H}_{\epsilon}[\mu_{t}^{\delta}] = \int_{\mathbb{R}^{d}} \left[(\partial_{t}\mu_{t}^{\delta}) * G^{\epsilon} \right] \left(\log(\mu_{t}^{\delta} * G^{\epsilon}) + 1 \right) dv = \int_{\mathbb{R}^{d}} (\partial_{t}\mu_{t}^{\delta}) \cdot \left[G^{\epsilon} * \log(\mu_{t}^{\delta} * G^{\epsilon}) \right] dv$$

The last line is achieved by the self-adjointness of convolution with G^{ϵ} and eliminating the constant term due to the conserved mass of μ^{δ} . Integrating in t, we obtain

$$\mathcal{H}_{\epsilon}[\mu_{r}^{\delta}] - \mathcal{H}_{\epsilon}[\mu_{s}^{\delta}] = \int_{s}^{r} \int_{\mathbb{R}^{d}} (\partial_{t}\mu_{t}^{\delta}) \cdot [G^{\epsilon} * \log(\mu_{t}^{\delta} * G^{\epsilon})] dv dt$$
$$= \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} [\tilde{\nabla}G^{\epsilon} * \log(\mu_{t}^{\delta} * G^{\epsilon})] \cdot dM_{t}^{\delta} dt$$
$$= \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu_{t}^{\delta}} \cdot dM_{t}^{\delta} dt.$$
(22)

We now turn to establishing estimates independent of $\delta > 0$ to pass to the limit. Estimates on the right-hand side of (22):

According to Lemma 31, we have the estimate

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu^{\delta}}\right| \lesssim_{\epsilon,E} |v|^p + |v_*|^p,$$

where $p \leq 1$. By the first moment assumption of M_t , we have

$$\int_0^T \iint_{\mathbb{R}^{2d}} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_t^{\delta}} \right| d|M_t|(v, v_*) dt \lesssim_{\epsilon, E} \int_0^T \iint_{\mathbb{R}^{2d}} |v| + |v_*| d|M_t|(v, v_*) dt < \infty.$$

This estimate also extends to M_t^δ

$$\int_0^T \iint_{\mathbb{R}^{2d}} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_t^{\delta}} \right| d|M_t^{\delta}|(v, v_*) dt < \infty.$$

Note that these estimates are independent of $\delta > 0$. Convergence $\delta \to 0$:

Using the weak in time continuity of μ , we can consider

$$|\mu_t^{\delta} * G^{\epsilon}(v') - \mu_t * G^{\epsilon}(v')| \le \int_{-1}^1 \eta(t') |\langle \mu_{t-\delta t'}, G^{\epsilon}(v'-\cdot)\rangle - \langle \mu_t, G^{\epsilon}(v'-\cdot)\rangle |dt'.$$

The \cdot stands for the convoluted variable. Since t belongs to a compact set, the function $t \mapsto \langle \mu_t, G^{\epsilon}(v' - \cdot) \rangle$ is uniformly continuous from the weak continuity of μ . In particular, using the continuity in v' and the lower bound from Lemma 29 we conclude that for any R > 0

$$|\log(\mu_t^{\delta} * G^{\epsilon}) - \log(\mu_t * G^{\epsilon})| \to 0 \quad \text{uniformly on } B_R.$$
(23)

Therefore by Lemma 29,

$$\begin{split} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} - \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \right| &= \left| \tilde{\nabla} G^{\epsilon} * \log(\mu_{t}^{\delta} * G^{\epsilon})(v, v_{*}) - \tilde{\nabla} G^{\epsilon} * \log(\mu_{t} * G^{\epsilon})(v, v_{*}) \right| \\ &\leq \int_{\mathbb{R}^{d}} w |\nabla G^{\epsilon}(v - v') - \nabla G^{\epsilon}(v_{*} - v')| |\log(\mu_{t}^{\delta} * G^{\epsilon}(v')) - \log(\mu_{t} * G^{\epsilon}(v'))| \, dv' \\ &\leq \int_{B_{R_{0}}^{c}} w |\nabla G^{\epsilon}(v - v') - \nabla G^{\epsilon}(v_{*} - v')| C_{\epsilon} \left\langle v' \right\rangle \, dv' \\ &\quad + \sup_{B_{R_{0}}} |\log(\mu_{t}^{\delta} * G^{\epsilon}) - \log(\mu_{t} * G^{\epsilon})| \int_{B_{R_{0}}} w |\nabla G^{\epsilon}(v - v') - \nabla G^{\epsilon}(v_{*} - v')| \, dv'. \end{split}$$

For a fixed (v, v_*) , we obtain the convergence to zero by taking $\delta \to 0$ and $R_0 \to \infty$ in the previous estimate. Using continuity, we obtain that for any R > 0

$$\left|\tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu_{t}^{\delta}}(v,v_{*}) - \tilde{\nabla}\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu_{t}}(v,v_{*})\right| \to 0 \quad \text{uniformly on } [0,T] \times B_{R} \times B_{R}.$$
(24)

We turn to the limit estimate for the right hand side of (22). For any R > 0, we have

$$\begin{aligned} \left| \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} \cdot dM_{t}^{\delta} dt - \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \cdot dM_{t} dt \right| \\ &\leq \left| \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \left(\tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} - \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \right) \cdot dM_{t}^{\delta} dt \right| + \left| \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \cdot dM_{t}^{\delta} dt - \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \cdot dM_{t} dt \\ &\leq \int_{s}^{r} \iint_{B_{R} \times B_{R}} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} - \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \right| d|M_{t}^{\delta}| dt + \int_{s}^{r} \iint_{(B_{R} \times B_{R})^{C}} \left| \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} - \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \right| d|M_{t}^{\delta}| dt + o(1). \end{aligned}$$

The last term is o(1) as $\delta \to 0$ due to similar estimates from the previous step. By sending $\delta \to 0$ (the first term vanishes due to (24)) and then sending $R \to \infty$ (the second term vanishes again due to the estimate from the previous step), we obtain the convergence

$$\lim_{\delta \to 0} \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}^{\delta}} \cdot dM_{t}^{\delta} dt = \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \cdot dM_{t}^{\delta} dt.$$
(25)

Convergence of the left-hand side of (22)

By (23), Lemma 29 and the uniform bound on the second moment, we have that

$$\begin{aligned} |\mathcal{H}_{\epsilon}[\mu_{t}^{\delta}] - \mathcal{H}_{\epsilon}[\mu_{t}]| &\leq \int_{\mathbb{R}^{d}} |(\mu_{t}^{\delta} * G^{\epsilon}) \log(\mu_{t}^{\delta} * G^{\epsilon})(v) - (\mu_{t} * G^{\epsilon}) \log(\mu_{t} * G^{\epsilon})(v)| dv \\ &\to 0, \quad \text{as } \delta \to 0. \end{aligned}$$

Therefore, by the previous equation and (25) we can take $\delta \to 0$ in (22) to obtain

$$\mathcal{H}_{\epsilon}[\mu_{r}] - \mathcal{H}_{\epsilon}[\mu_{s}] = \frac{1}{2} \int_{s}^{r} \iint_{\mathbb{R}^{2d}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu_{t}} \cdot dM_{t}(v, v_{*}) dt,$$

which is the desired result.

5 JKO scheme for ϵ -Landau equation

This section is devoted to the proof of Theorem 9 after a series of preliminary lemmata. Our construction of curves of maximal slope in Theorem 9 uses the basic minimizing movement/variational approximation scheme of Jordan et al. [36]. Fix a small time step $\tau > 0$ and initial datum $\mu_0 \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ and consider the recursive minimization procedure for $n \in \mathbb{N}$

$$\nu_0^{\tau} := \mu_0, \qquad \nu_n^{\tau} \in \operatorname{argmin}_{\lambda \in \mathscr{P}_{2,E}} \left[\mathcal{H}_{\epsilon}(\lambda) + \frac{1}{2\tau} d_L^2(\nu_{n-1}^{\tau}, \lambda) \right].$$
(26)

Then, we concatenate these minimizers into a curve by setting

$$\mu_0^{\tau} := \mu_0, \qquad \mu_t^{\tau} := \nu_n^{\tau}, \text{ for } t \in ((n-1)\tau, n\tau].$$
 (27)

The scheme given by (26) and (27) satisfies the abstract formulation in [3] giving

Proposition 33 (Landau JKO scheme). For any $\tau > 0$ and $\mu_0 \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, there exists $\nu_n^{\tau} \in \mathscr{P}_{2,E}(\mathbb{R}^d)$ for every $n \in \mathbb{N}$ as described in (26). Furthermore, up to a subsequence of μ_t^{τ} described in (27) as $\tau \to 0$, there exists a locally absolutely continuous curve $(\mu_t)_{t\geq 0}$ such that

$$\mu_t^{\tau} \rightharpoonup \mu_t, \qquad \forall t \in [0, \infty).$$

Proof. Our metric setting is $(\mathscr{P}_{\mu_0}, d_L)$ (see Theorem 7) with the weak topology σ . This space is essentially $\mathscr{P}_{2,E}(\mathbb{R}^d)$ except we need to make sure that d_L is a proper metric, hence we remove the probability measures with infinite Landau distance. We follow the proof of Erbar [25] which consists in verifying [3, Assumptions 2.1 a,b,c]. These assumptions are listed and verified now.

1. \mathcal{H}_{ϵ} is sequentially σ -lsc on d_L -bounded sets: Suppose $\mu_n \in \mathscr{P}_{2,E}(\mathbb{R}^d) \rightharpoonup \mu \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, this implies $\mu_n * G^{\epsilon} \rightharpoonup \mu * G^{\epsilon}$ in $\mathscr{P}_2(\mathbb{R}^d)$. It is known that

$$\mathcal{H}(\mu) = \begin{cases} \int_{\mathbb{R}^d} f(v) \log f(v) dv, & \mu = f\mathcal{L} \\ +\infty, & \text{else} \end{cases}$$

is σ -lsc and since $\mathcal{H}_{\epsilon}(\mu) = \mathcal{H}(\mu * G^{\epsilon})$, we achieve the first property.

2. \mathcal{H}_{ϵ} is lower bounded: By Carlen-Carvalho Lemma 29 for fixed $\epsilon > 0$, $\log(\mu * G^{\epsilon})$ is uniformly lower bounded by a linearly growing term. For fixed $\mu \in \mathscr{P}_{2,E}(\mathbb{R}^d)$, we have, with Cauchy-Schwarz

$$\mathcal{H}_{\epsilon}(\mu) \gtrsim_{\epsilon} - \int_{\mathbb{R}^d} \langle v \rangle \mu * G^{\epsilon}(v) dv \ge - \left(\int_{\mathbb{R}^d} \langle v \rangle^2 \mu * G^{\epsilon}(v) dv \right)^{\frac{1}{2}} \ge -(\mathcal{O}(\epsilon) + E)^{\frac{1}{2}} > -\infty.$$

3. d_L -bounded sets are relatively sequentially σ -compact: This is one of the consequences from Theorem 7.

The existence of minimizers, ν_n^{τ} , to (26) and limits, μ_t , to (27) is guaranteed from [3, Corollary 2.2.2] and [3, Proposition 2.2.3], respectively.

At the abstract level, the limit curve constructed in Proposition 33 has no relation to $\sqrt{D_{\epsilon}}$. The following lemmata bridge this gap.

Lemma 34. For any $\mu_0 \in \mathscr{P}_2(\mathbb{R}^d)$, we have

$$\sqrt{D_{\epsilon}(\mu_0)} \le |\partial^- \mathcal{H}_{\epsilon}|(\mu_0)$$

Proof. For fixed $\epsilon, R_1, R_2 > 0$ and $\gamma \in \mathbb{R}$, take T > 0 from Theorem 47 in Appendix A and the unique weak solution $\mu \in C([0,T]; \mathscr{P}_2(\mathbb{R}^d))$ to

$$\begin{cases} \partial_t \mu = \nabla \cdot \{ \mu \phi_{R_1} \int_{\mathbb{R}^d} \phi_{R_1 *} \psi_{R_2}(v - v_*) | v - v_*|^{\gamma + 2} \Pi[v - v_*] (J_0^{\epsilon} - J_{0*}^{\epsilon}) d\mu(v_*) \} \\ \mu(0) = \mu_0 \end{cases}$$

The functions $0 \le \phi_{R_1}, \psi_{R_2} \le 1$ are smooth cut-off functions with the following properties

$$\phi_{R_1}(v) = \begin{cases} 1, & |v| \le R_1 \\ 0, & |v| \ge R_1 + 1 \end{cases}, \quad \psi_{R_2}(z) = \begin{cases} 0, & |z| \le 1/R_2 \\ 1, & |z| \ge 2/R_2 \end{cases}$$

For t > 0 we define J_t^{ϵ} to be the gradient of the first variation of \mathcal{H}_{ϵ} applied to μ_t , meaning

$$J_t^{\epsilon} = \nabla G^{\epsilon} * \log[\mu_t * G^{\epsilon}] \in C^{\infty}(\mathbb{R}^d; \mathbb{R}^d).$$

For this proof alone, we define the reduced ϵ -entropy-dissipation

$$D_{\epsilon}^{R_1,R_2}(\mu_0) := \frac{1}{2} \iint_{\mathbb{R}^{2d}} \phi_{R_1} \phi_{R_1*} \psi_{R_2}(v-v_*) |v-v_*|^{\gamma+2} |\Pi[v-v_*](J_0^{\epsilon}-J_{0*}^{\epsilon})|^2 d\mu_0(v) d\mu_0(v_*).$$

On the other hand, as the ϵ -entropy dissipation comes from the negative time derivative of entropy, we have

$$\begin{split} D_{\epsilon}^{R_{1},R_{2}}(\mu_{0}) &= \lim_{t\downarrow 0} \frac{\mathcal{H}_{\epsilon}(\mu_{0}) - \mathcal{H}_{\epsilon}(\mu_{t})}{t} = \lim_{t\downarrow 0} \frac{\mathcal{H}_{\epsilon}(\mu_{0}) - \mathcal{H}_{\epsilon}(\mu_{t})}{d_{L}(\mu_{0},\mu_{t})} \frac{d_{L}(\mu_{0},\mu_{t})}{t} \\ &\leq \lim_{t\downarrow 0} \left\{ \frac{\mathcal{H}_{\epsilon}(\mu_{0}) - \mathcal{H}_{\epsilon}(\mu_{t})}{d_{L}(\mu_{0},\mu_{t})} \times \frac{1}{t} \right. \\ &\left. \times \left(\int_{0}^{t} \sqrt{\frac{1}{2} \iint_{\mathbb{R}^{2d}} \phi_{R_{1}}^{2} \phi_{R_{1}}^{2} \psi_{R_{2}}^{2} |v - v_{*}|^{\gamma+2} (J_{s}^{\epsilon} - J_{s*}^{\epsilon}) \cdot \Pi[v - v_{*}] (J_{0}^{\epsilon} - J_{0*}^{\epsilon}) d\mu_{s}(v) d\mu_{s}(v_{*}) ds} \right) \right\} \\ &\leq |\partial^{-}\mathcal{H}_{\epsilon}|(\mu_{0}) \sqrt{D_{\epsilon}^{R_{1},R_{2}}(\mu_{0})}. \end{split}$$

In the last inequality, we have used the Lebesgue differentiation theorem with strong-weak convergence since μ is continuous in time as well as the fact that $\phi_{R_1}^2 \leq \phi_{R_1}$ and $\psi_{R_2}^2 \leq \psi_{R_2}$ since $0 \leq \phi_{R_1}, \psi_{R_2} \leq 1$. We are left with the inequality

$$\sqrt{D_{\epsilon}^{R_1,R_2}(\mu_0)} \le |\partial^- \mathcal{H}_{\epsilon}|(\mu_0), \quad \forall R_1, R_2 > 0.$$

As functions of R_1, R_2 individually, $D_{\epsilon}^{R_1, R_2}(\mu_0)$ is non-decreasing. Furthermore, the integrand of $D_{\epsilon}^{R_1, R_2}(\mu_0)$ converges to the integrand of $D_{\epsilon}(\mu_0)$ pointwise μ_0 -almost every v, v_* . Thus, an application of the monotone convergence theorem in the limit $R_1, R_2 \to \infty$ on the above inequality completes the proof. **Lemma 35.** $|\partial^{-}\mathcal{H}_{\epsilon}|$ is a strong upper gradient for \mathcal{H}_{ϵ} in $\mathscr{P}_{\mu_{0}}(\mathbb{R}^{d})$ where $\mu_{0} \in \mathscr{P}_{2,E}(\mathbb{R}^{d})$.

Proof. Fix $\lambda, \nu \in \mathscr{P}_{\mu_0}(\mathbb{R}^d)$ so that by the triangle inequality of Theorem 7, we have $d_L(\lambda, \nu) < \infty$. Now by Proposition 25, there exists a pair of curves $(\mu, M) \in \mathcal{GCE}_1^E$ connecting λ, ν and $\mathcal{A}(\mu_t, M_t) = d_L^2(\lambda, \nu)$ for almost every $t \in [0, 1]$. Using Remark 28 and Lemma 34, we have

$$|\mathcal{H}_{\epsilon}(\lambda) - \mathcal{H}_{\epsilon}(\nu)| \leq \int_{0}^{1} \sqrt{D_{\epsilon}(\mu_{t})} |\dot{\mu}|(t) dt \leq \int_{0}^{1} |\partial^{-}\mathcal{H}_{\epsilon}|(\mu_{t})|\dot{\mu}|(t) dt.$$

We now have all the ingredients to prove Theorem 9 so that we can relate curves of maximal slope to weak solutions of the ϵ -Landau equation.

Proof of Theorem 9. Take a limit curve μ_t constructed in Proposition 33. By the previous Lemma 35, the assumptions of [3, Theorem 2.3.3] are fulfilled so the curve is a maximal slope with respect to $|\partial^- \mathcal{H}_{\epsilon}|$ and satisfies the associated energy dissipation inequality

$$\mathcal{H}_{\epsilon}(\mu_r) - \mathcal{H}_{\epsilon}(\mu_s) + \frac{1}{2} \int_s^r |\partial^- \mathcal{H}_{\epsilon}(\mu_t)|^2 dt + \frac{1}{2} \int_s^r |\dot{\mu}|^2(t) dt \le 0.$$

The inequality of Lemma 34 gives

$$\mathcal{H}_{\epsilon}(\mu_r) - \mathcal{H}_{\epsilon}(\mu_s) + \frac{1}{2} \int_s^r D_{\epsilon}(\mu_t) dt + \frac{1}{2} \int_s^r |\dot{\mu}|^2(t) dt \le 0,$$

which is precisely the statement that the limit curve μ_t is a curve of maximal slope with respect to $\sqrt{D_{\epsilon}}$.

Remark 36. The results of Proposition 33 and Lemma 34 can be generalized to other regularization kernels $G^{s,\epsilon}$, in particular, the Maxwellian regularization. However, this is not the case for Lemma 35 since the proof relies on Proposition 27, see Remark 32.

6 Recovering the full Landau equation as $\epsilon \to 0$

Theorems 8 and 9 provide the basic existence theory for the $\epsilon > 0$ approximation of the Landau equation. In this section, we prove the $\epsilon \downarrow 0$ analogue of Theorem 8 which is Theorem 11.

Sketch proof of Theorem 11. By repeating the proof of Theorem 8, we see that the crucial ingredient is the chain rule (18) in Proposition 27. For now assume the following

Claim 37. Assume (A1), (A2), (A3) and let M be any grazing rate such that $(\mu, M) \in \mathcal{GCE}_T^E$ and

$$\int_0^T \mathcal{A}(\mu_t, M_t) dt < \infty.$$

Then we have the chain rule

$$\mathcal{H}[\mu_r] - \mathcal{H}[\mu_s] = \frac{1}{2} \int_s^r \iint_{\mathbb{R}^6} \tilde{\nabla} \left[\frac{\delta \mathcal{H}}{\delta \mu} \right] \cdot dM_t dt.$$
(28)

By following the steps of the proof of Theorem 8 and using (28) instead of (18), one completes the proof of Theorem 11. We dedicate this section to proving Claim 37.

Equation (28) is clearly the $\epsilon \downarrow 0$ limit of (18). The left-hand side of (28) can be obtained from the left-hand side of (18) using the finite entropy assumption (A2) and the fact that $\epsilon \mapsto \mathcal{H}_{\epsilon}[\mu_t]$ is non-increasing for every t. We refer to [25, Proof of Proposition 4.2; Step 4: part d)] for more details on a similar argument.

The difficulty remains in deducing that the right-hand side of (18) converges to the right-hand side of (28) as $\epsilon \downarrow 0$ given by

$$\int_{0}^{T} \iint_{\mathbb{R}^{6}} \tilde{\nabla} \frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \cdot dM_{t} dt \to \int_{0}^{T} \iint_{\mathbb{R}^{6}} \tilde{\nabla} \frac{\delta \mathcal{H}}{\delta \mu} \cdot dM_{t} dt, \quad \epsilon \downarrow 0$$
(29)

under the additional assumptions (A1), (A2), (A3) on f. The key result which we will use repeatedly in this section is the following theorem which is a specific case of the result in [38, Chapter 4, Theorem 17].

Theorem 38 (Extended Dominated Convergence Theorem (EDCT)). Let $(H_{\epsilon})_{\epsilon>0}$ and $(I_{\epsilon})_{\epsilon>0}$ be sequences of measurable functions on X satisfying $I_{\epsilon} \geq 0$ and suppose there exists measurable functions H, I satisfying

- 1. $|H_{\epsilon}| \leq I_{\epsilon}$ for every $\epsilon > 0$ and pointwise a.e.
- 2. H_{ϵ} and I_{ϵ} converge pointwise a.e. to H and I, respectively.

$$\lim_{\epsilon \downarrow 0} \int_X I_\epsilon = \int_X I < \infty.$$

Then, we have the convergence

$$\lim_{\epsilon \downarrow 0} \int_X H_\epsilon = \int_X H.$$

Setting $M = m\mathcal{L} \otimes \mathcal{L}$ (valid by Proposition 18) and using Young's inequality on the right-hand side of (18), we obtain the majorants

$$\tilde{\nabla}\left[\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right] \cdot m_{t} \leq \frac{1}{2}ff_{*}\left|\tilde{\nabla}\left[\frac{\delta\mathcal{H}_{\epsilon}}{\delta\mu}\right]\right|^{2} + \frac{1}{2}\frac{|m_{t}|^{2}}{ff_{*}}.$$

Notice that the first term is precisely the integrand of D_{ϵ} while the second term is the integrand of the action functional $\mathcal{A}(\mu_t, M_t)$ which has no dependence on ϵ and is henceforth ignored. We can apply EDCT 38 with $X = (0, T) \times \mathbb{R}^6$ to prove (29) once we show

$$\int_{0}^{T} \iint_{\mathbb{R}^{6}} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^{2} dv_{*} dv dt \to \int_{0}^{T} \iint_{\mathbb{R}^{6}} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}}{\delta \mu} \right] \right|^{2} dv_{*} dv dt, \quad \epsilon \downarrow 0.$$
(30)

The pointwise a.e. convergence hypothesis of EDCT 38 is straightforward based on the regularization of \mathcal{H}_{ϵ} through G^{ϵ} . Focusing on (30), we will use a standard Dominated

Convergence Theorem (DCT) for the integration in the t variable, by proving

$$\iint_{\mathbb{R}^{6}} \frac{1}{2} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^{2} dv_{*} dv \to \iint_{\mathbb{R}^{6}} \frac{1}{2} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}}{\delta \mu} \right] \right|^{2} dv_{*} dv, \quad \text{a.e. } t, \\
\iint_{\mathbb{R}^{6}} \frac{1}{2} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^{2} dv_{*} dv \leq C \iint_{\mathbb{R}^{6}} \frac{1}{2} ff_{*} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}}{\delta \mu} \right] \right|^{2} dv_{*} dv, \quad \text{a.e. } t \quad \forall \epsilon > 0,$$
(31)

where C > 0 is a constant independent of $\epsilon > 0$. The estimate of (31) guarantees the L_t^1 majorisation due to the finite entropy-dissipation assumption (A3).

Our estimates in this section accomplish both the convergence and the estimate of (31)by nested application of EDCT 38. The significance of all three assumptions (A1), (A2), and (A3) will be apparent in proving the convergence in (31).

Remark 39. In this section, the only properties of G^{ϵ} we use are that it is a non-negative radial approximate identity with sufficiently many moments. As in the construction of minimizing movement curves in Section 5, the results of this section can be achieved with other radial approximate identities.

6.1 Outline of technical strategy to prove (31)

The need to apply EDCT 38 instead of the more classical Lebesgue DCT is that we are unable to prove pointwise estimates in v for the function $v \to f \int_{\mathbb{R}^3} f_* \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta f} \right] \right|^2 dv_*$. Instead, our estimates in this section rely on the relief of the section $v \to f \int_{\mathbb{R}^3} f_* \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta f} \right] \right|^2 dv_*$. estimates in this section rely on the self-adjointness of convolution against radial exponentials (SACRE) to construct a convergent majorant in ϵ .

Step 1: Finding majorants and appealing to EDCT 38 We seek to find pointwise a.e. majorants in the v variable

$$f \int_{\mathbb{R}^3} f_* \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^2 dv_* \le I_{\epsilon}^1(v),$$

where $I_{\epsilon}^{1}(v)$ satisfies the hypothesis for the majorant in EDCT 38. We show that I_{ϵ}^{1} converges pointwise to some I^1 , since I^1_{ϵ} depends on ϵ only through convolutions against G^{ϵ} , which is an approximation of the identity. Hence, we are left with showing the integral convergence Item 3 of EDCT 38

$$\int_{\mathbb{R}^3} I^1_{\epsilon}(v) dv dt \to \int_{\mathbb{R}^3} I^1(v) dv, \quad \epsilon \to 0$$

Step 2: Use SACRE with G^{ϵ} To show the integral convergence for I^{1}_{ϵ} , we find functions A^{1} and B^{1} such that

$$I^1_{\epsilon}(v) \le A^1(v)(G^{\epsilon} * B^1)(v)$$

and apply EDCT 38. As in the previous step, the pointwise convergence is easily proved. Hence, we are left to show the integral convergence

$$\int_{\mathbb{R}^3} A^1(G^{\epsilon} * B^1) dv \to \int_{\mathbb{R}^3} A^1 B^1, \quad \epsilon \to 0.$$

The key observation is applying SACRE to obtain

$$\int_{\mathbb{R}^3} A^1(G^{\epsilon} * B^1) = \int_{\mathbb{R}^3} \underbrace{(G^{\epsilon} * A^1)B^1}_{=:I^2_{\epsilon}}.$$

Therefore, we have reduced the problem to showing integral convergence Item 3 of EDCT for I_{ϵ}^2 (as the pointwise convergence is easily proved).

Step 3: Reiterate step 2

We repeat the process outlined in Step 2 by finding functions A^2 and B^2 such that we have the pointwise bound

$$I_{\epsilon}^2(v) \le A^2(v)(G^{\epsilon} * B^2)(v).$$

Again the pointwise convergence for the majorant follows easily, hence we only need to check the integral convergence Item 3 of EDCT 38 given by

$$\int_{\mathbb{R}^3} A^2(G^{\epsilon} * B^2) \to \int_{\mathbb{R}^3} A^2 B^2.$$

Using SACRE, we study instead the integral convergence of

$$I^3_{\epsilon}(v) = (G^{\epsilon} * A^2)B^2.$$

Eventually, after a finite number of times of finding majorants and applying SACRE, we will obtain a majorant I_{ϵ}^{i} for which the estimates and the convergence as $\epsilon \to 0$ follows from the standard Lebesgue DCT, using the bound of the weighted Fisher information in terms of the entropy-dissipation (see Theorem 40) and assumption (A3).

6.2 Preparatory results

As mentioned in the previous section, for the final step of the proof we need a bound on the weighted Fisher information and a closely related variant in terms of the entropy-dissipation originally discovered by the third author in [19].

Theorem 40. Suppose $\gamma \in (-4, 0]$ and let $f \ge 0$ be a probability density belong to $L^1_{2-\gamma} \cap L \log L(\mathbb{R}^3)$. We have

$$\int_{\mathbb{R}^3} f(v) \langle v \rangle^{\gamma} \left| \nabla \frac{\delta \mathcal{H}}{\delta f} \right|^2 dv + \int_{\mathbb{R}^3} f(v) \left\langle v \right\rangle^{\gamma} \left| v \times \nabla \frac{\delta \mathcal{H}}{\delta f} \right|^2 dv \le C(1 + D_{w,\mathcal{H}}(f)),$$

where C > 0 is a constant depending only on the bounds of $m_{2-\gamma}(f)$ and the Boltzmann entropy, $\mathcal{H}(f)$, of f.

The estimate in this precise form can be found in [18, Proposition 4, p. 10]. We will refer to the second term on the left-hand side as a 'cross Fisher information'.

To decompose the entropy-dissipation in a manageable way that makes the cross Fisher term more apparent, we have the following linear algebra fact.

Lemma 41. For $x, y \in \mathbb{R}^3$, we have

$$|x|^2(y \cdot \Pi[x]y) = |x \times y|^2$$

Proof. Without loss of generality, we assume neither x, y = 0 or else the statement holds trivially. Let θ be an oriented angle between x and y. We expand the definition of $\Pi[x]$ and observe

$$\begin{aligned} |x|^2(y \cdot \Pi[x]y) &= y \cdot (|x|^2 I - x \otimes x)y = |x|^2 |y|^2 - |x \cdot y|^2 = |x|^2 |y|^2 (1 - \cos^2 \theta) = |x|^2 |y|^2 \sin^2 \theta \\ &= |x \times y|^2. \end{aligned}$$

The following lemma shows how we use assumption (A1) to control the singularity of the weight.

Lemma 42. Given $\gamma \in (-3, 0]$, assume that f satisfies (A1) for some $0 < \eta \le \gamma + 3$, then we have for a.e. t

$$\int_{\mathbb{R}^3} f_*(t) |v - v_*|^{\gamma} dv_* \le C_1(t) \langle v \rangle^{\gamma}, \quad \int_{\mathbb{R}^3} f_*(t) |v_*|^2 |v - v_*|^{\gamma} dv_* \le C_2(t) \langle v \rangle^{\gamma}, \quad (32)$$

where

$$||C_1||_{L^{\infty}(0,T)} \lesssim_{\gamma,\eta} ||\langle\cdot\rangle^{-\gamma} f(t)||_{L^{\infty}\left(0,T;L^1\cap L^{\frac{3-\eta}{3+\gamma-\eta}}(\mathbb{R}^3)\right)} ||C_2||_{L^{\infty}(0,T)} \lesssim_{\gamma,\eta} ||\langle\cdot\rangle^{2-\gamma} f(t)||_{L^{\infty}\left(0,T;L^1\cap L^{\frac{3-\eta}{3+\gamma-\eta}}(\mathbb{R}^3)\right)}.$$

Proof. We will only prove the first inequality of (32) since the second inequality uses the same procedure. We split the estimation for local $|v| \leq 1$ and far-field $|v| \geq 1$. $|v| \leq 1$

We split the integral over v_* into two regions

$$\begin{split} \int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* &= \int_{|v - v_*| \ge 1} f_* |v - v_*|^{\gamma} dv_* + \int_{|v - v_*| \le 1} f_* |v - v_*|^{\gamma} dv_* \\ &\le 1 + \int_{|v - v_*| \le 1} f_* |v - v_*|^{\gamma} dv_*, \end{split}$$

where we have used that $\int_{\mathbb{R}^3} f = 1$ and $\gamma \leq 0$. For the integral with the singularity, we apply Young's convolution inequality with conjugate exponents $\left(\frac{3-\eta}{3+\gamma-\eta}, \frac{-3+\eta}{\gamma}\right)$

$$\int_{|v-v_*|\leq 1} f_* |v-v_*|^{\gamma} dv_* \leq ||f^*(\chi_{B_1}|\cdot|^{\gamma})||_{L^{\infty}} \leq ||f||_{L^{\frac{3-\eta}{3+\gamma-\eta}}} ||\chi_{B_1}|\cdot|^{\gamma}||_{L^{\frac{-3+\eta}{\gamma}}} \leq \left(\frac{\omega_2}{\eta}\right)^{\frac{-3+\eta}{\gamma}} ||f||_{L^{\frac{3-\eta}{3+\gamma-\eta}}} ||f||_{L^$$

Here, ω_2 is the volume of the unit sphere in \mathbb{R}^3 . $|v| \ge 1$

Once again, we split the integral into two parts

$$\begin{split} \int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* &= \int_{|v_*| \le \frac{1}{2} |v|} f_* |v - v_*|^{\gamma} dv_* + \int_{|v_*| \ge \frac{1}{2} |v|} f_* |v - v_*|^{\gamma} dv_* \\ &\le 2^{-\gamma} |v|^{\gamma} \int_{|v_*| \le \frac{1}{2} |v|} f_* dv_* + 2^{-\gamma} |v|^{\gamma} \int_{|v_*| \ge \frac{1}{2} |v|} f_* |v_*|^{-\gamma} |v - v_*|^{\gamma} dv_*. \end{split}$$

The first term and second term come from the following inequalities based on their respective integration regions

$$|v - v_*| \ge |v| - |v_*| \ge \frac{1}{2}|v|, \quad 1 \le 2^{-\gamma}|v|^{\gamma}|v_*|^{-\gamma}.$$

We estimate the first integral using the unit mass of f, while the second integral is more delicate but again uses the splitting of the previous step to obtain

$$\int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* \le 2^{-\gamma} |v|^{\gamma} + 2^{-\gamma} |v|^{\gamma} \left(\int_{|v - v_*| \ge 1} f_* |v_*|^{-\gamma} |v - v_*|^{\gamma} dv_* + \int_{|v - v_*| \le 1} f_* |v_*|^{-\gamma} |v - v_*|^{\gamma} dv_* \right)$$

In the large brackets, the first integral can be estimated by $m_{-\gamma}(f)$. Now we use the same Young's inequality argument for the remaining integral to obtain

$$\int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* \le 2^{-\gamma} |v|^{\gamma} + 2^{-\gamma} |v|^{\gamma} \left(m_{-\gamma}(f) + \left(\frac{\omega_2}{\eta}\right)^{\frac{-3+\eta}{\gamma}} ||| \cdot |^{-\gamma} f||_{L^{\frac{3-\eta}{3+\gamma-\eta}}(\mathbb{R}^3)} \right).$$

The proof is complete by combining the estimates for $|v| \leq 1$ and $|v| \geq 1$.

Lemma 43 (Peetre). For any $p \in \mathbb{R}$ and $x, y \in \mathbb{R}^d$, we have

$$\frac{\langle x \rangle^p}{\langle y \rangle^p} \le 2^{|p|/2} \langle x - y \rangle^{|p|}.$$

Proof. Our proof follows [5]. Starting with the case p = 2, for fixed vectors $a, b \in \mathbb{R}^d$ we have, with the help of Young's inequality,

$$1 + |a - b|^{2} \le 1 + |a|^{2} + 2|a||b| + |b|^{2} \le 1 + 2|a|^{2} + 2|b|^{2}$$
$$\le 2 + 2|a|^{2} + 2|a|^{2}|b|^{2} + 2|b|^{2} = 2(1 + |a|^{2})(1 + |b|^{2}).$$

Dividing by $\langle b \rangle^2$ and setting a = x - y, b = -y, we obtain the inequality for p = 2

$$\frac{\langle x \rangle^2}{\langle y \rangle^2} \le 2 \langle x - y \rangle^2.$$

By taking non-negative powers, this proves the inequality for $p \ge 0$. On the other hand, when we divided by $\langle b \rangle^2$ we could have also set a = x - y, b = x to obtain

$$\frac{\langle y \rangle^2}{\langle x \rangle^2} \le 2 \langle x - y \rangle^2.$$

Taking strictly non-negative powers here proves the inequality for p < 0.

Next, we prove an estimate for algebraic functions (growing or decaying) convoluted against G^{ϵ} with respect to the original function.

Lemma 44. For any $p \in \mathbb{R}$, we have

$$\int_{\mathbb{R}^d} \langle w \rangle^p \, G^{\epsilon}(v-w) dw \le C \, \langle v \rangle^p \,,$$

where C > 0 is a constant depending only on |p| and $m_{|p|}(G)$.

Proof. We use Peetre's inequality in Lemma 43 to introduce v - w into the angle brackets

$$\int_{\mathbb{R}^d} \langle w \rangle^p G^{\epsilon}(v-w) dw \leq 2^{|p|/2} \langle v \rangle^p \int_{\mathbb{R}^d} \langle v-w \rangle^{|p|} G^{\epsilon}(v-w) dw$$

$$= 2^{|p|/2} \langle v \rangle^p \int_{\mathbb{R}^d} (1+|w|^2)^{\frac{|p|}{2}} \epsilon^{-d} G(w/\epsilon) dw = 2^{|p|/2} \langle v \rangle^p \int_{\mathbb{R}^d} (1+\epsilon^2 |w|^2)^{\frac{|p|}{2}} G(w) dw$$

$$\leq C_{|p|} \langle v \rangle^p \left[1+\epsilon^{|p|} \int_{\mathbb{R}^d} |w|^{|p|} G(w) dw \right] \leq C_{|p|} \left[1+\epsilon^{|p|} m_{|p|}(G) \right] \langle v \rangle^p$$

We stress that Peetre's inequality 43 is necessary for the estimate of Lemma 44 with *non-positive* powers p which we apply in the sequel. Finally, the last result we will need is an integration by parts formula for the differential operator associated to the cross Fisher information.

Lemma 45 (Twisted integration by parts). Let f, g be smooth scalar functions of \mathbb{R}^3 which are sufficiently integrable. Then, we have the formula

$$\int_{\mathbb{R}^3} (v \times \nabla_v g(v)) f(v) dv = -\int_{\mathbb{R}^3} g(v) (v \times \nabla_v f(v)) dv.$$

Here, the meaning of $v \times \nabla_v$ is

$$v \times \nabla_v f(v) = (v^2 \partial^3 f(v) - v^3 \partial^2 f(v), v^3 \partial^1 f(v) - v^1 \partial^3 f(v), v^1 \partial^2 f(v) - v^2 \partial^1 f(v))$$

6.3 Proof of (31) using EDCT 38

We start by decomposing and estimating the integrand of D_{ϵ} . With the help of Lemma 41, we expand the square term of the integrand to see

$$\begin{split} \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^{2} &= |v - v_{*}|^{2+\gamma} |\Pi[v - v_{*}](b^{\epsilon} * a^{\epsilon} - b^{\epsilon} * a^{\epsilon}_{*})|^{2} \\ &\leq |v - v_{*}|^{\gamma} (4|v \times (b^{\epsilon} * a^{\epsilon})|^{2} + 4|v_{*} \times (b^{\epsilon} * a^{\epsilon}_{*})|^{2} \\ &+ 4|v \times (b^{\epsilon} * a^{\epsilon}_{*})|^{2} + 4|v_{*} \times (b^{\epsilon} * a^{\epsilon})|^{2}) \\ &\leq 4|v - v_{*}|^{\gamma} \underbrace{|v \times (b^{\epsilon} * a^{\epsilon})|^{2}}_{(1)} + 4|v - v_{*}|^{\gamma} \underbrace{|v_{*} \times (b^{\epsilon} * a^{\epsilon}_{*})|^{2}}_{(2)} \\ &+ 4|v|^{2}|v - v_{*}|^{\gamma} \underbrace{|b^{\epsilon} * a^{\epsilon}_{*}|^{2}}_{(3)} + 4|v_{*}|^{2}|v - v_{*}|^{\gamma} \underbrace{|b^{\epsilon} * a^{\epsilon}|^{2}}_{(4)}, \end{split}$$

where we use the shorthand notation

$$b^{\epsilon} = G^{\epsilon}$$
 and $a^{\epsilon} = \nabla \log(G^{\epsilon} * f).$ (33)

By using that G^{ϵ} is an approximation of the identity, we know that the integrand of D_{ϵ} converges pointwise a.e. to the integrand of D as $\epsilon \downarrow 0$. As well, each (1) for i = 1, 2, 3, 4 converge pointwise a.e. to

$$() \to \frac{|v \times \nabla f|^2}{f^2}, () \to \frac{|v_* \times \nabla_* f_*|^2}{f_*^2}, () \to \frac{|\nabla_* f_*|^2}{f_*^2}, () \to \frac{|\nabla f|^2}{f^2}.$$

By EDCT 38, to show the integral convergence in (31), it suffices to show, for example,

$$\iint_{\mathbb{R}^6} ff_* |v - v_*|^{\gamma} \mathbb{D} dv dv_* \to \iint_{\mathbb{R}^6} ff_* |v - v_*|^{\gamma} \frac{|v \times \nabla f|^2}{f^2} dv dv_*,$$

and similarly for each (1) for i = 2, 3, 4. By symmetry considerations when swapping the variables $v \leftrightarrow v_*$, the convergence for the terms (1) and (4) controls the convergence for (2) and (3), respectively. Hence we will focus on the term (4) first and then on term (1).

6.3.1 Term ④

We seek to show in the limit $\epsilon \downarrow 0$,

$$\iint_{\mathbb{R}^{6}} ff_{*} |v_{*}|^{2} |v - v_{*}|^{\gamma} |b^{\epsilon} * a^{\epsilon}|^{2} dv_{*} dv = \int_{\mathbb{R}^{3}} \left(\int_{\mathbb{R}^{3}} f_{*} |v_{*}|^{2} |v - v_{*}|^{\gamma} dv_{*} \right) f |b^{\epsilon} * a^{\epsilon}|^{2} dv$$

$$\rightarrow \int_{\mathbb{R}^{3}} \left(\int_{\mathbb{R}^{3}} f_{*} |v_{*}|^{2} |v - v_{*}|^{\gamma} dv_{*} \right) \frac{|\nabla f|^{2}}{f} dv.$$
(34)

By the reordering of integrations written above, we now think of the double integral over v, v_* of $ff_*|v_*|^2|v-v_*|^{\gamma}|b^{\epsilon}*a^{\epsilon}|^2$ as a single integral of the function $\left(\int_{\mathbb{R}^d} f_*|v_*|^2|v-v_*|^{\gamma}dv_*\right)f|b^{\epsilon}*a^{\epsilon}|^2$ over v. For this is the single integral convergence that we will use EDCT 38 for. We can use Cauchy-Schwarz on the convolution integral to absorb the power term as follows

$$\begin{split} |b^{\epsilon} * a^{\epsilon}|^{2} &= \left| \int_{\mathbb{R}^{3}} b^{\epsilon}(v-w) a^{\epsilon}(w) dw \right|^{2} \leq \left(\int_{\mathbb{R}^{3}} \langle w \rangle^{-\gamma} b^{\epsilon}(v-w) dw \right) \left(\int_{\mathbb{R}^{3}} b^{\epsilon}(v-w) \langle w \rangle^{\gamma} |a^{\epsilon}(w)|^{2} dw \right) \\ &\leq C \left\langle v \right\rangle^{-\gamma} b^{\epsilon} * [\langle \cdot \rangle^{\gamma} |a^{\epsilon}(\cdot)|^{2}], \end{split}$$

where the last inequality comes from Lemma 44. Continuing with Lemma 42, we have

$$\left(\int_{\mathbb{R}^3} f_* |v_*|^2 |v - v_*|^{\gamma} dv_*\right) f |b^{\epsilon} * a^{\epsilon}|^2 \le C f b^{\epsilon} * [\langle \cdot \rangle^{\gamma} |a^{\epsilon}|^2]$$

By EDCT 38, we reduce the problem to showing in the limit $\epsilon \downarrow 0$

$$\int_{\mathbb{R}^3} f b^{\epsilon} * [\langle \cdot \rangle^{\gamma} | a^{\epsilon} |^2] dv \to \int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \frac{|\nabla f|^2}{f} dv.$$

This is were we use SACRE, Step 2 of our general strategy 6.1. Application of SACRE and further simplification using the specific forms of a^{ϵ} and b^{ϵ} (see (33)) yields

$$\int_{\mathbb{R}^3} f b^{\epsilon} * [\langle \cdot \rangle^{\gamma} | a^{\epsilon} |^2] dv = \int_{\mathbb{R}^3} [b^{\epsilon} * f] \langle v \rangle^{\gamma} | a^{\epsilon} |^2 dv = \int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \frac{|b^{\epsilon} * \nabla f|^2}{b^{\epsilon} * f} dv.$$
(35)

We work with this simplified expression and note that pointwise convergence is still valid

$$\frac{|b^{\epsilon} * \nabla f|^2}{b^{\epsilon} * f} \to \frac{|\nabla f|^2}{f}.$$

Next, we notice that the function $(F, f) \mapsto \frac{|F|^2}{f}$ is jointly convex in $F \in \mathbb{R}^3$ and f > 0, so we can use Jensen's inequality to obtain a further pointwise majorant for the integrand of (35)

$$\frac{|b^{\epsilon} * \nabla f|^2}{b^{\epsilon} * f} \le b^{\epsilon} * \left[\frac{|\nabla f|^2}{f}\right].$$

Using EDCT 38 again, we reduce the problem to showing in the limit $\epsilon \downarrow 0$

$$\int_{\mathbb{R}^3} \langle v \rangle^{\gamma} b^{\epsilon} * \left[\frac{|\nabla f|^2}{f} \right] dv \to \int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \frac{|\nabla f|^2}{f} dv.$$

We use SACRE once more and place the convolution onto the weight term

$$\int_{\mathbb{R}^3} \langle v \rangle^{\gamma} b^{\epsilon} * \left[\frac{|\nabla f|^2}{f} \right] dv = \int_{\mathbb{R}^3} [b^{\epsilon} * \langle \cdot \rangle^{\gamma}] \frac{|\nabla f|^2}{f} dv.$$

Now, we are in a position to apply the classical Dominated Convergence Theorem. We notice that we have the pointwise convergence

$$[b^{\epsilon} * \langle \cdot \rangle^{\gamma}] \to \langle v \rangle^{\gamma}.$$

Furthermore, using Lemma 44, we can estimate $b^{\epsilon} * \langle \cdot \rangle^{\gamma}$ uniformly in ϵ to find the domination

$$[b^{\epsilon} * \langle \cdot \rangle^{\gamma}] \frac{|\nabla f|^2}{f} \le C \langle v \rangle^{\gamma} \frac{|\nabla f|^2}{f}.$$

Using Theorem 40 and the finite entropy-dissipation assumption (A3), we know that the right-hand side belongs to L_v^1 a.e. $t \in (0, T)$. Therefore, for a.e. $t \in (0, T)$ the conditions of the Dominated Convergence Theorem are satisfied so we have the integral convergence

$$\int_{\mathbb{R}^3} [b^\epsilon * \langle \cdot \rangle^\gamma] \frac{|\nabla f|^2}{f} dv \to \int_{\mathbb{R}^3} \langle v \rangle^\gamma \frac{|\nabla f|^2}{f} dv.$$

We have closed the argument for the convergence of (34) after retracing the previous estimates with EDCT 38.

6.3.2 Term ①

We seek to show in the limit $\epsilon \downarrow 0$,

$$\iint_{\mathbb{R}^6} ff_* |v - v_*|^{\gamma} |v \times (b^{\epsilon} * a^{\epsilon})|^2 dv_* dv = \int_{\mathbb{R}^3} \left(\int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* \right) f |v \times (b^{\epsilon} * a^{\epsilon})|^2 dv$$

$$\rightarrow \int_{\mathbb{R}^3} \left(\int_{\mathbb{R}^3} f_* |v - v_*|^{\gamma} dv_* \right) \frac{|v \times \nabla f|^2}{f} dv$$
(36)

using the same strategy of nested applications of EDCT 38 like in the previous Section 6.3.1. We will encounter difficulty when trying to use Jensen's inequality due to the cross Fisher information term. As in the previous Section 6.3.1, we have written this double integral over v, v_* as a single integral over v. By EDCT 38 and Lemma 42, it suffices to show the integral convergence of

$$\int_{\mathbb{R}^3} \langle v \rangle^{\gamma} f | v \times (b^{\epsilon} * a^{\epsilon}) |^2 dv \to \int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \frac{|v \times \nabla f|^2}{f}$$
(37)

to obtain the integral convergence of (36). Pointwise, we can make the following manipulations

$$v \times (b^{\epsilon} * a^{\epsilon}) = v \times \left(\int_{\mathbb{R}^3} G^{\epsilon}(v - w) \nabla \log(f * G^{\epsilon}(w)) dw \right)$$

$$= v \times \left(\int_{\mathbb{R}^3} \nabla G^{\epsilon}(v - w) \log(f * G^{\epsilon}(w)) dw \right)$$

$$= \int_{\mathbb{R}^3} w \times \nabla G^{\epsilon}(v - w) \log(f * G^{\epsilon}(w)) dw$$

$$= \int_{\mathbb{R}^3} G^{\epsilon}(v - w) w \times \nabla \log(f * G^{\epsilon}(w)) dw,$$
(38)

where we have used the radial symmetry of G^{ϵ} to get the cancellation $(v-w) \times \nabla G^{\epsilon}(v-w) = 0$ and the twisted integration by parts Lemma 45 (we note that we not pick any signs in the integration by parts, as the variable w appears with a minus sign in the arguments of G^{ϵ}).

We apply Cauchy-Schwarz, multiply and divide by $\langle w \rangle^{\gamma}$, and use Lemma 44 to obtain

$$\begin{split} |v \times (b^{\epsilon} * a^{\epsilon})|^{2} &\leq \left(\int_{\mathbb{R}^{3}} G^{\epsilon}(v-w) \left\langle w \right\rangle^{-\gamma} dw \right) \left(\int_{\mathbb{R}^{3}} G^{\epsilon}(v-w) \left\langle w \right\rangle^{\gamma} \left| w \times \frac{\nabla f * G^{\epsilon}(w)}{f * G^{\epsilon}(w)} \right|^{2} dw \right) \\ &\lesssim_{\gamma} \left\langle v \right\rangle^{-\gamma} \left(\int_{\mathbb{R}^{3}} G^{\epsilon}(v-w) \left\langle w \right\rangle^{\gamma} \left| w \times \frac{\nabla f * G^{\epsilon}(w)}{f * G^{\epsilon}(w)} \right|^{2} dw \right). \end{split}$$

Remembering that this majorant holds pointwise on the integrand of (37), we multiply by $\langle v \rangle^{\gamma} f(v)$ and obtain

$$\langle v \rangle^{\gamma} f(v) | v \times (b^{\epsilon} * a^{\epsilon}) |^2 \lesssim f\left(\int_{\mathbb{R}^3} G^{\epsilon}(v-w) \langle w \rangle^{\gamma} \left| w \times \frac{\nabla f * G^{\epsilon}(w)}{f * G^{\epsilon}(w)} \right|^2 dw \right).$$

Now, we recognise a convolution inside the brackets. Hence, using SACRE we can re-write

$$\int_{\mathbb{R}^3} f\left(\int_{\mathbb{R}^3} G^{\epsilon}(v-w) \langle w \rangle^{\gamma} \left| w \times \frac{\nabla f * G^{\epsilon}(w)}{f * G^{\epsilon}(w)} \right|^2 dw \right) dv = \int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \frac{|v \times \nabla f * G^{\epsilon}(v)|^2}{f * G^{\epsilon}(v)} dv.$$

Using EDCT 38, we need to show the convergence of the right-hand side. Here, it is now possible to use Jensen's inequality after some more manipulations.

Claim 46.

$$\frac{|v \times \nabla f * G^{\epsilon}(v)|^2}{f * G^{\epsilon}(v)} \le \int_{\mathbb{R}^3} G^{\epsilon}(v - w) \frac{|w \times \nabla f(w)|^2}{f(w)} dw.$$
(39)

Proof of Claim 46. We start by repeating a similar argument to (38). Using that G^{ϵ} is radially symmetric and the twisted integration by parts Lemma 45 we obtain

$$\begin{aligned} v \times \nabla f * G^{\epsilon}(v) &= v \times \left(\int_{\mathbb{R}^3} \nabla G^{\epsilon}(v-w) f(w) dw \right) \\ &= \int_{\mathbb{R}^3} w \times \nabla G^{\epsilon}(v-w) f(w) dw \\ &= \int_{\mathbb{R}^3} G^{\epsilon}(v-w) \underbrace{(w \times \nabla_w f(w))}_{=:F(w)} dw. \end{aligned}$$

Therefore, since $(F, f) \mapsto \frac{|F|^2}{f}$ is jointly convex in $F \in \mathbb{R}^3$ and f > 0, we apply Jensen's inequality to the left-hand side of (39) to see

$$\frac{|v \times \nabla f \ast G^{\epsilon}(v)|^2}{f \ast G^{\epsilon}(v)} = \frac{|F \ast G^{\epsilon}|^2}{f \ast G^{\epsilon}}(v) \le \frac{|F|^2}{f} \ast G^{\epsilon}(v) = \int_{\mathbb{R}^3} G^{\epsilon}(v-w) \frac{|w \times \nabla f(w)|^2}{f(w)} dw,$$

which proves the claim.

Continuing, by EDCT 38, we seek to establish the integral convergence of

$$\int_{\mathbb{R}^3} \langle v \rangle^{\gamma} \left[\frac{|F|^2}{f} * G^{\epsilon} \right] (v) dv = \int_{\mathbb{R}^3} [\langle \cdot \rangle^{\gamma} * G^{\epsilon}](v) \frac{|v \times \nabla f(v)|^2}{f(v)} dv.$$

Finally, the integrand of the right-hand side has a majorant due to Lemma 44

$$[\langle \cdot \rangle^{\gamma} * G^{\epsilon}](v) \frac{|v \times \nabla f(v)|^2}{f(v)} \lesssim \langle v \rangle^{\gamma} \frac{|v \times \nabla f(v)|^2}{f(v)}.$$

Once again using Theorem 40 and Assumption (A3), we obtain that for a.e. $t \in (0, T)$ the right hand side belongs to $L_v^1(\mathbb{R}^3)$. Using Dominated Convergence theorem, we see that the integral converges. Tracing back the estimates, this takes care of the convergence of the term (I) and establishes the convergence in (37).

We note that the estimates in the previous subsections not only establish the a.e. pointwise convergence of (31), but also the majorisation

$$\iint_{\mathbb{R}^6} \frac{1}{2} f f_* \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}_{\epsilon}}{\delta \mu} \right] \right|^2 dv_* dv \le C \iint_{\mathbb{R}^6} \frac{1}{2} f f_* \left| \tilde{\nabla} \left[\frac{\delta \mathcal{H}}{\delta \mu} \right] \right|^2 dv_* dv, \quad \text{a.e. } t \quad \forall \epsilon > 0,$$

where

$$C \lesssim \left|\left|\left\langle\cdot\right\rangle^{-\gamma} f(t)\right|\right|_{L^{\infty}\left(0,T;L^{1}\cap L^{\frac{3-\eta}{3+\gamma-\eta}}(\mathbb{R}^{3})\right)} + \left|\left|\left\langle\cdot\right\rangle^{2-\gamma} f(t)\right|\right|_{L^{\infty}\left(0,T;L^{1}\cap L^{\frac{3-\eta}{3+\gamma-\eta}}(\mathbb{R}^{3})\right)}$$

by Lemma 42. Hence, using assumption (A3) and (31) we can apply Lebesgue DCT to pass to the limit in the time integral and show the desired chain rule Claim 37.

A An auxiliary PDE for Lemma 34

In this section, we study weak solutions to the following PDE

$$\begin{cases} \partial_t \mu = \nabla \cdot \{ \mu \phi_{R_1} \int_{\mathbb{R}^d} \phi_{R_1 *} \psi_{R_2}(v - v_*) | v - v_*|^{\gamma + 2} \Pi[v - v_*] (J_0^{\epsilon} - J_{0*}^{\epsilon}) d\mu(v_*) \} \\ \mu(0) = \mu_0 \end{cases}$$
(40)

We assume the initial data μ_0 belongs to $\mathscr{P}_2(\mathbb{R}^d)$. For $R_1, R_2 > 0$, the functions $0 \leq \phi_{R_1}, \psi_{R_2} \leq 1$ are smooth cut-off functions used to approximate the identity function in different ways.

$$\phi_{R_1}(v) = \begin{cases} 1, & |v| \le R_1 \\ 0, & |v| \ge R_1 + 1 \end{cases}, \quad \psi_{R_2}(z) = \begin{cases} 0, & |z| \le 1/R_2 \\ 1, & |z| \ge 2/R_2 \end{cases}$$

For $\epsilon > 0$, J_0^{ϵ} is the gradient of first variation of \mathcal{H}_{ϵ} applied to μ_0 , meaning

$$J_0^{\epsilon} = \nabla G^{\epsilon} * \log[\mu_0 * G^{\epsilon}] \in C^{\infty}(\mathbb{R}^d; \mathbb{R}^d).$$

The main result of this section is

Theorem 47. Fix $\epsilon, R_1, R_2 > 0$, $\gamma \in \mathbb{R}$, and $\mu_0 \in \mathscr{P}_2(\mathbb{R}^d)$. Then, there exists a T > 0 which depends on $\epsilon, \gamma, R_1, R_2$, and μ_0 such that equation (40) has a unique weak solution $\mu \in C([0, T]; \mathscr{P}_2(\mathbb{R}^d))$.

By Lemma 30, we know that J_0^{ϵ} is uniformly (with constant depending on ϵ and μ_0). The purpose of ϕ_{R_1}, ϕ_{R_1*} is to cut off the growth of $J_0^{\epsilon}, J_{0*}^{\epsilon}$ to ensure that the 'velocity field' in the right-hand side of (40) is globally Lipschitz (it is, in fact, smooth and compactly supported). The $\psi_{R_2}(v-v_*)$ term avoids the possible singularities coming from the weight $|v-v_*|^{\gamma+2}$ for soft potentials $\gamma < 0$.

The time, T > 0 in Theorem 47 has an explicit upper bound. Our strategy is to employ a fixed point argument in the space $C([0, T]; \mathscr{P}_2(\mathbb{R}^d))$ which we will equip with the following metric

$$d(\mu,\nu) := \sup_{t \in [0,T]} W_2(\mu(t),\nu(t)), \quad \mu,\nu \in C([0,T];\mathscr{P}_2(\mathbb{R}^d)),$$

where W_2 is the 2-Wasserstein distance on $\mathscr{P}_2(\mathbb{R}^d)$. We have closely followed the procedure in [9] with appropriate modifications for this setting.

Remark 48. Since we are cutting off the 'velocity' field at radius R_1, R_2 , the growth of J_0^{ϵ} is inconsequential. Hence the results of this section can be applied when replacing the convolution kernel of J_0^{ϵ} with general tailed exponential distributions $G^{s,\epsilon}(v)$ for s > 0.

For $\mu \in \mathscr{P}_2(\mathbb{R}^d)$, we will denote by $U[\mu](v)$ the following function

$$U[\mu](v) := -\phi_{R_1} \int_{\mathbb{R}^d} \phi_{R_1*} \psi_{R_2}(v - v_*) |v - v_*|^{\gamma+2} \Pi[v - v_*] (J_0^{\epsilon} - J_{0*}^{\epsilon}) d\mu(v_*),$$

so that the PDE in (40) can be written as a nonlinear transport/continuity equation

$$\partial_t \mu(t) = -\nabla \cdot \{\mu(t)U[\mu(t)]\}$$

To fix ideas, the weak formulation of (40) is such that the following equality holds for all test functions $\tau \in C_c^{\infty}(\mathbb{R}^d)$ and times $t \in [0, T]$

$$\int_{\mathbb{R}^d} \tau(v) d\mu_r(v) - \int_{\mathbb{R}^d} \tau(v) d\mu_0(v) = \int_0^t \int_{\mathbb{R}^d} \phi_{R_1} \nabla \tau(v) \cdot \int_{\mathbb{R}^d} \phi_{R_1*} \psi_{R_2}(v - v_*) |v - v_*|^{\gamma + 2} \Pi[v - v_*] (J_0^\epsilon - J_{0*}^\epsilon) d\mu_s(v) d\mu_s(v) ds.$$

Thanks to all the smooth cutoffs from ϕ_{R_1}, ϕ_{R_1*} , and ψ_{R_2} and $\mu_0 \in \mathscr{P}_2(\mathbb{R}^d)$, we can enlarge the class of test functions to smooth functions with quadratic growth. In particular, by choosing $\tau(v) = |v|^2$ and symmetrising the right-hand side by swapping $v \leftrightarrow v_*$, we see that the second moment of μ_0 is conserved along the evolution of (40).

Our first step is to look at the level of the characteristic equation associated to (40).

Lemma 49 (Characteristic equation). For any T > 0, $\mu \in C([0,T]; \mathscr{P}_2(\mathbb{R}^d))$ and $v_0 \in \mathbb{R}^d$, there exists a unique solution $v \in C^1((0,T); \mathbb{R}^d) \cap C([0,T]; \mathbb{R}^d)$ to the following ODE

$$\frac{dv}{dt} = U[\mu(t)](v), \quad v(0) = v_0.$$

Furthermore, the growth rate satisfies

$$|v(t)| \le \max\{|v_0|, R_1 + 1\}, \quad \forall t \in [0, T].$$

Proof. $U[\mu(t)](\cdot)$ is smooth and compactly supported uniformly in t, so classical Cauchy-Lipschitz theory gives existence and uniqueness of solution v with the promised regularity.

For the estimate on the growth rate, note that $U[\mu]$ has support contained in B_{R_1+1} . Points outside this ball do not change in time according to this ODE.

We will denote by Φ^t_{μ} the flow map associated to this ODE, so that

$$\frac{d}{dt}\Phi^t_{\mu}(v_0) = U[\mu(t)](\Phi^t_{\mu}(v_0)), \quad \Phi^0_{\mu}(v_0) = v_0.$$

It is known that, given $\nu \in C([0,T]; \mathscr{P}_2(\mathbb{R}^d))$, the curve of probability measures $\mu(t) = \Phi_{\nu}^t \# \mu_0$ is a weak solution to

$$\partial_t \mu(t) = -\nabla \cdot \{\mu(t)U[\nu(t)]\}, \quad \mu(0) = \mu_0.$$

Here, $\Phi^t_{\nu} \# \mu_0$ is the push-forward measure of μ_0 defined in duality with $\tau \in C_b(\mathbb{R}^d)$ by

$$\int_{\mathbb{R}^d} \tau(v) d(\Phi^t_\nu \# \mu_0)(v) = \int_{\mathbb{R}^d} \tau(\Phi^t_\nu(v)) d\mu_0(v).$$

We seek to find a fixed point to the map $\mu \mapsto \Phi^t_{\mu} \# \mu_0$ as it would weakly solve (40). To better understand the properties of this map, we need to establish estimates on the flow map through U as a function of time and measures.

Lemma 50 (L^{∞} estimate for velocity field). There exists a constant $C = C(\epsilon, \gamma, R_1, R_2, \mu_0) > 0$ such that for every T > 0 and $\nu \in C([0, T]; \mathscr{P}_2(\mathbb{R}^d))$, we have

$$|U[\nu(t)](v)| \le C, \quad \forall t \in [0, T], v \in \mathbb{R}^d.$$

Proof. Estimate for $\gamma \ge -2$:

We have the following three inequalities

$$|v - v_*|^{\gamma+2} \lesssim_{\gamma} |v|^{\gamma+2} + |v_*|^{\gamma+2}, \quad ||\Pi[v - v_*]|| \le 1, \quad J_0^{\epsilon} \lesssim_{\epsilon,\mu_0} 1$$

due to the range of γ , boundedness of Π , and Lemma 30, respectively. These three inequalities provide the estimate

$$|U[\nu(t)](v)| \lesssim_{\gamma,\epsilon,\mu_0} \phi_{R_1}(v) \int_{\mathbb{R}^d} \phi_{R_1}(v_*) (|v|^{\gamma+2} + |v_*|^{\gamma+2}) d\nu_t(v_*),$$

where we have dropped ψ_{R_2} altogether. For the integral term, we apply Hölder's inequality taking advantage of the compact support of ϕ_{R_1} and the unit mass of ν_t to further obtain

$$|U[\nu(t)](v)| \lesssim_{\gamma,\epsilon,\mu_0} \phi_{R_1}(v)(R_1^{2+\gamma} + \langle v \rangle^{2+\gamma}) \int_{\mathbb{R}^d} d\nu_t(v_*) \lesssim_{R_1} \phi_{R_1}(v) \langle v \rangle^{2+\gamma}$$

Again, since ϕ_{R_1} has compact support, we can brutally estimate the polynomial to conclude. Estimate for $\gamma < -2$:

Unlike the previous case, we change one of the inequalities due to the unavailability of a triangle inequality and use

$$\psi_{R_2}(v-v_*)|v-v_*|^{\gamma+2} \lesssim 1/R_2^{\gamma+2}, \quad ||\Pi[v-v_*]|| \le 1, \quad J_0^{\epsilon} \lesssim_{\epsilon,\mu_0} 1.$$

From these inequalities and the compact support of ϕ_{R_1} , we have

$$|U[\nu(t)](v)| \lesssim_{\gamma,\epsilon,\mu_0,R_2} \phi_{R_1}(v) \int_{\mathbb{R}^d} \phi_{R_1}(v_*) d\nu_t(v_*) \le 1,$$

which concludes the proof.

The next result follows exactly as in [9].

Lemma 51 (Time continuity of flow map). Let $C = C(\epsilon, \gamma, R_1, R_2, \mu_0) > 0$ be the same constant from Lemma 50. Then for any T > 0, and $\nu \in C([0, T]; \mathscr{P}_2(\mathbb{R}^d))$ we have

$$||\Phi_{\nu}^t - \Phi_{\nu}^s||_{L^{\infty}(\mathbb{R}^d)} \le C|t - s|.$$

Our next objective is to establish the regularity of the flow map with respect to the measures in the subscript. To simplify the subsequent lemmata, let us use the notation in the following

Lemma 52. Define

$$F: (v, w) \in \mathbb{R}^d \times \mathbb{R}^d \mapsto \phi_{R_1}(v)\phi_{R_1}(w)\psi_{R_2}(v-w)|v-w|^{\gamma+2}\Pi[v-w](J_0^{\epsilon}(v) - J_0^{\epsilon}(w))$$

The function F is smooth and compactly supported. In particular, for every $k, l \in \mathbb{N}$, there is a constant $C = C(\epsilon, \gamma, R_1, R_2, \mu_0, k, l) > 0$ such that

$$||D_v^k D_w^l F||_{L^{\infty}(\mathbb{R}^d \times \mathbb{R}^d)} \le C.$$

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Proof. The compact support property comes from the factor of $\phi_{R_1}(v)\phi_{R_1}(w)$ in the definition. The regularity comes from the avoidance of v = w due to the factor $\psi_{R_2}(v - w)$.

Corollary 53 (Pointwise and measurewise regularity of U). Consider the constant $C = C(\epsilon, \gamma, R_1, R_2, \mu_0, k, l) > 0$ from Lemma 52 above. We have the following

1. Take $C_1 = C(\epsilon, \gamma, R_1, R_2, \mu_0, 0, 1) > 0$. For every $T > 0; \nu^1, \nu^2 \in C([0, T]; \mathscr{P}_2(\mathbb{R}^d)); t \in [0, T]; v \in \mathbb{R}^d$ we have the estimate

$$|U[\nu^{1}(t)](v) - U[\nu^{2}(t)](v)| \le C_{1}W_{2}(\nu_{t}^{1},\nu_{t}^{2}).$$

2. Take $C_2 = C(\epsilon, \gamma, R_1, R_2, \mu_0, 1, 0) > 0$. For every $T > 0; \nu \in C([0, T]; \mathscr{P}_2(\mathbb{R}^d)); t \in [0, T]; v_1, v_2 \in \mathbb{R}^d$ we have the estimate

$$|U[\nu(t)](v_1) - U[\nu(t)](v_2)| \le C_2 |v_1 - v_2|.$$

Remark 54. By considering the anti-symmetric property of F when swapping variables $v \leftrightarrow w$, one really obtains $C_1 = C_2$. Their distinction in this corollary is artificial.

Proof. Item 1:

Firstly, for every $t \in [0, T]$ take $\pi(t) \in \mathscr{P}_2(\mathbb{R}^d \times \mathbb{R}^d)$ the 2-Wasserstein optimal transportation plan connecting $\nu^1(t)$ and $\nu^2(t)$ which exists, see [46]. We estimate the difference with notation from Lemma 52

$$|U[\nu^{1}(t)](v) - U[\nu^{2}(t)](v)| = \left| \int_{\mathbb{R}^{d}} F(v, w) d\nu_{t}^{1}(w) - \int_{\mathbb{R}^{d}} F(v, \bar{w}) d\nu_{t}^{2}(\bar{w}) \right|$$
$$= \left| \iint_{\mathbb{R}^{2d}} F(v, w) - F(v, \bar{w}) d\pi_{t}(w, \bar{w}) \right|$$
$$\leq C_{1} \iint_{\mathbb{R}^{2d}} |w - \bar{w}| d\pi_{t}(w, \bar{w})$$
$$\leq C_{1} W_{2}(\nu_{t}^{1}, \nu_{t}^{2}).$$

The first inequality uses a mean-value type estimate (in the second variable of F) and the second inequality uses Cauchy-Schwarz or equivalently, that W_2 is stronger than W_1 . Item 2:

As with item 1, we estimate the difference using F to find

$$|U[\nu(t)](v_1) - U[\nu(t)](v_2)| = \left| \int_{\mathbb{R}^d} F(v_1, w) - F(v_2, w) d\nu_t(w) \right|$$

$$\leq \int_{\mathbb{R}^d} |F(v_1, w) - F(v_2, w)| d\nu_t(w)$$

$$\leq C_2 |v_1 - v_2|.$$

Once more, a mean-value type estimate is applied (in the first variable of F) and we recall ν_t is a probability measure.

The next result combines both items of Corollary 53 to estimate the regularity of the flow map with respect to measures and follows exactly as in [9].

Lemma 55 (Continuity of flow map with respect to measures). For T > 0 fix any $\nu^1, \nu^2 \in C([0,T]; \mathscr{P}_2(\mathbb{R}^d))$ and $t \in [0,T]$. With $C := C_1 = C_2$ the same constants in Corollary 53, we have the estimate

$$||\Phi_{\nu^{1}}^{t} - \Phi_{\nu^{2}}^{t}||_{L^{\infty}(\mathbb{R}^{d})} \leq (e^{Ct} - 1)d(\nu^{1}, \nu^{2}),$$

recalling that $d(\nu^{1}, \nu^{2}) = \sup_{t \in [0,T]} W_{2}(\nu_{t}^{1}, \nu_{t}^{2}).$

It is by now classical how to obtain Theorem 47 from Corollary 53 and Lemma 55, see [9, 13, 28] for instance. The time of existence can be given by any $0 < T < \frac{1}{C} \log 2$ where C > 0 is chosen as in Lemma 55 and the result follows by a fixed point argument and a classical extension to show that the solution is defined for all times.

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References

- [1] R. Alexandre, J. Liao, and C. Lin. Some a priori estimates for the homogeneous Landau equation with soft potentials. *Kinet. Relat. Models*, 8(4):617–650, 2015.
- [2] L. Ambrosio. Minimizing movements. Rend. Accad. Naz. Sci. XL Mem. Mat. Appl. (5), 19:191-246, 1995.
- [3] L. Ambrosio, N. Gigli, and G. Savaré. Gradient flows in metric spaces and in the space of probability measures. Lectures in Mathematics ETH Zürich. Birkhäuser Verlag, Basel, second edition, 2008.
- [4] A. A. Arsen'ev and N. V. Peskov. The existence of a generalized solution of Landau's equation. Ž. Vyčisl. Mat i Mat. Fiz., 17(4):1063–1068, 1096, 1977.
- [5] J. Barros-Neto. An introduction to the theory of distributions. Marcel Dekker, Inc. New York, 1973. Pure and Applied Mathematics, 14.

- [6] J.-D. Benamou and Y. Brenier. A computational fluid mechanics solution to the Monge-Kantorovich mass transfer problem. *Numer. Math.*, 84(3):375–393, 2000.
- [7] A. V. Bobylev, M. Pulvirenti, and C. Saffirio. From particle systems to the Landau equation: a consistency result. *Comm. Math. Phys.*, 319(3):683–702, 2013.
- [8] G. Buttazzo. Semicontinuity, relaxation and integral representation in the calculus of variations, volume 207 of Pitman Research Notes in Mathematics Series. Longman Scientific & Technical, Harlow; copublished in the United States with John Wiley & Sons, Inc., New York, 1989.
- [9] J. A. Cañizo, J. A. Carrillo, and J. Rosado. A well-posedness theory in measures for some kinetic models of collective motion. *Math. Models Methods Appl. Sci.*, 21(3):515– 539, 2011.
- [10] E. A. Carlen and M. C. Carvalho. Strict entropy production bounds and stability of the rate of convergence to equilibrium for the Boltzmann equation. J. Statist. Phys., 67(3-4):575–608, 1992.
- [11] K. Carrapatoso, L. Desvillettes, and L. He. Estimates for the large time behavior of the Landau equation in the Coulomb case. Arch. Ration. Mech. Anal., 224(2):381–420, 2017.
- [12] K. Carrapatoso and S. Mischler. Landau equation for very soft and Coulomb potentials near Maxwellians. Ann. PDE, 3(1):Paper No. 1, 65, 2017.
- [13] J. A. Carrillo, Y.-P. Choi, and M. Hauray. The derivation of swarming models: mean-field limit and Wasserstein distances. In *Collective dynamics from bacteria to crowds*, volume 553 of *CISM Courses and Lect.*, pages 1–46. Springer, Vienna, 2014.
- [14] J. A. Carrillo, K. Craig, and F. S. Patacchini. A blob method for diffusion. Calc. Var. Partial Differential Equations, 58(2):Paper No. 53, 53, 2019.
- [15] J. A. Carrillo, J. Hu, L. Wang, and J. Wu. A particle method for the homogeneous Landau equation. J. Comput. Phys. X, 7:100066, 2020.
- [16] P. Degond and B. Lucquin-Desreux. The Fokker-Planck asymptotics of the Boltzmann collision operator in the Coulomb case. *Math. Models Methods Appl. Sci.*, 2(2):167–182, 1992.
- [17] L. Desvillettes. Entropy dissipation estimates for the Landau equation in the Coulomb case and applications. J. Funct. Anal., 269(5):1359–1403, 2015.
- [18] L. Desvillettes. Autour du théorème H de Boltzmann. To appear at Séminaire Laurent Schwartz, EDP et Applications, 2020.
- [19] L. Desvillettes and K. Fellner. Exponential decay toward equilibrium via entropy methods for reaction-diffusion equations. J. Math. Anal. Appl., 319(1):157–176, 2006.

- [20] L. Desvillettes, L. He, and J.-C. Jiang. A new monotonicity formula for spatially homogeneous Landau equation with Coulomb potential and its applications. In preparation.
- [21] L. Desvillettes and C. Villani. On the spatially homogeneous Landau equation for hard potentials. I. Existence, uniqueness and smoothness. *Comm. Partial Differential Equations*, 25(1-2):179–259, 2000.
- [22] L. Desvillettes and C. Villani. On the spatially homogeneous Landau equation for hard potentials. II. H-theorem and applications. Comm. Partial Differential Equations, 25(1-2):261–298, 2000.
- [23] J. Dolbeault, B. Nazaret, and G. Savaré. A new class of transport distances between measures. Calc. Var. Partial Differential Equations, 34(2):193–231, 2009.
- [24] M. Erbar. Gradient flows of the entropy for jump processes. Ann. Inst. Henri Poincaré Probab. Stat., 50(3):920–945, 2014.
- [25] M. Erbar. A gradient flow approach to the Boltzmann equation. arXiv preprint arXiv:1603.00540, 2016.
- [26] M. Erbar and J. Maas. Gradient flow structures for discrete porous medium equations. Discrete Contin. Dyn. Syst., 34(4):1355–1374, 2014.
- [27] N. Fournier and H. Guérin. Well-posedness of the spatially homogeneous Landau equation for soft potentials. J. Funct. Anal., 256(8):2542–2560, 2009.
- [28] F. Golse. On the dynamics of large particle systems in the mean field limit. In Macroscopic and large scale phenomena: coarse graining, mean field limits and ergodicity, volume 3 of Lect. Notes Appl. Math. Mech., pages 1–144. Springer, [Cham], 2016.
- [29] F. Golse, M. P. Gualdani, C. Imbert, and A. Vasseur. Partial regularity in time for the space homogeneous Landau equation with Coulomb potential. arXiv preprint arXiv:1906.02841, 2019.
- [30] F. Golse, C. Imbert, C. Mouhot, and A. F. Vasseur. Harnack inequality for kinetic Fokker-Planck equations with rough coefficients and application to the Landau equation. Ann. Sc. Norm. Super. Pisa Cl. Sci. (5), 19(1):253–295, 2019.
- [31] M. Gualdani and N. Guillen. Estimates for radial solutions of the homogeneous Landau equation with Coulomb potential. *Analysis & PDE*, 9(8):1773–1810, 2016.
- [32] M. Gualdani and N. Zamponi. A review for an isotropic Landau model. In PDE models for multi-agent phenomena, volume 28 of Springer INdAM Ser., pages 115–144. Springer, Cham, 2018.
- [33] M. P. Gualdani and N. Zamponi. Spectral gap and exponential convergence to equilibrium for a multi-species Landau system. Bull. Sci. Math., 141(6):509–538, 2017.
- [34] M. P. Gualdani and N. Zamponi. Global existence of weak even solutions for an isotropic Landau equation with Coulomb potential. *SIAM J. Math. Anal.*, 50(4):3676–3714, 2018.

- [35] Y. Guo. The Landau equation in a periodic box. Comm. Math. Phys., 231(3):391–434, 2002.
- [36] R. Jordan, D. Kinderlehrer, and F. Otto. The variational formulation of the Fokker-Planck equation. SIAM J. Math. Anal., 29(1):1–17, 1998.
- [37] E. M. Lifshitz and L. P. Pitaevskii. Course of theoretical physics ["Landau-Lifshits"]. Vol. 10. Pergamon International Library of Science, Technology, Engineering and Social Studies. Pergamon Press, Oxford-Elmsford, N.Y., 1981. Translated from the Russian by J. B. Sykes and R. N. Franklin.
- [38] H. L. Royden. *Real analysis*. Macmillan Publishing Company, New York, third edition, 1988.
- [39] E. Sandier and S. Serfaty. Gamma-convergence of gradient flows with applications to Ginzburg-Landau. Comm. Pure Appl. Math., 57(12):1627–1672, 2004.
- [40] F. Santambrogio. {Euclidean, metric, and Wasserstein} gradient flows: an overview. Bull. Math. Sci., 7(1):87–154, 2017.
- [41] S. Serfaty. Gamma-convergence of gradient flows on Hilbert and metric spaces and applications. Discrete Contin. Dyn. Syst., 31(4):1427–1451, 2011.
- [42] L. Silvestre. Upper bounds for parabolic equations and the Landau equation. J. Differential Equations, 262(3):3034–3055, 2017.
- [43] R. M. Strain and Z. Wang. Uniqueness of bounded solutions for the homogeneous relativistic Landau equation with Coulomb interactions. *Quart. Appl. Math.*, 78(1):107– 145, 2020.
- [44] C. Villani. On a new class of weak solutions to the spatially homogeneous Boltzmann and Landau equations. Arch. Rational Mech. Anal., 143(3):273–307, 1998.
- [45] C. Villani. On the spatially homogeneous Landau equation for Maxwellian molecules. Math. Models Methods Appl. Sci., 8(6):957–983, 1998.
- [46] C. Villani. Optimal transport, volume 338 of Grundlehren der Mathematischen Wissenschaften [Fundamental Principles of Mathematical Sciences]. Springer-Verlag, Berlin, 2009. Old and new.
- [47] K.-C. Wu. Global in time estimates for the spatially homogeneous Landau equation with soft potentials. J. Funct. Anal., 266(5):3134–3155, 2014.