

Existence and regularity of a weak function-solution for some Landau equations with a stochastic approach

Hélène Guérin*

March 13, 2001

Abstract

Using the Malliavin Calculus, this paper proves the existence of a weak function-solution of class C^∞ of the Landau equation for a generalization of Maxwellian molecules when the initial data is a probability measure.

Key Words : Landau Equation, Malliavin Calculus, White Noise, Nonlinear Stochastic Differential Equation.

MSC 2000 : 60H07, 60H30, 82C40

1 Introduction

The Landau equation, also called the Fokker-Planck-Landau equation, is obtained as limit of the Boltzmann equation when all the collisions become grazing. Its expression, in the spatially homogeneous case, is:

$$\frac{\partial f}{\partial t}(v, t) = \frac{1}{2} \sum_{i,j=1}^d \frac{\partial}{\partial v_i} \left\{ \int_{\mathbb{R}^d} dv_* a_{ij}(v - v_*) \left[f(v_*, t) \frac{\partial f}{\partial v_j}(v, t) - f(v, t) \frac{\partial f}{\partial v_{*j}}(v_*, t) \right] \right\} \quad (1)$$

where $f(v, t) \geq 0$ is the density of particles with velocity $v \in \mathbb{R}^d$ at time $t \in \mathbb{R}^+$, and $(a_{ij}(z))_{1 \leq i, j \leq d}$ is a nonnegative symmetric matrix depending on the interaction between the particles.

In this paper, we study the Landau equation for a generalization of Maxwell gas. We consider a matrix a of the form

$$a_{ij}(z) = h(|z|^2) \left(|z|^2 \delta_{ij} - z_i z_j \right) \quad (2)$$

where h is a positive continuous function on \mathbb{R}_+ such that there exist $m, M > 0$ with $\forall z \in \mathbb{R}^d$

$$m \leq h(|z|^2) \leq M \quad (3)$$

When h is a constant, we recognize the coefficient of the Landau equation for Maxwellian molecules.

We define the vector b by

$$\begin{aligned} b_i(z) &= \sum_{j=1}^d \partial_j a_{ij}(z) \\ &= -(d-1) h(|z|^2) z_i \end{aligned} \quad (4)$$

Then, by integration by parts, we can give a weak formulation of the equation (1), and consequently we define the notion of weak function-solution:

*Université Paris 10, MODALX, UFR SEGMI, 200 avenue de la République, 92000 Nanterre, hguerin@ccr.jussieu.fr

Definition 1 Let $f(., 0)$ be a nonnegative function on \mathbb{R}^d with finite mass and energy. A nonnegative function f on $\mathbb{R}^d \times \mathbb{R}^+$ is a weak function-solution of the Landau equation with initial data $f(., 0)$, if f satisfies the following equation for any test function $\varphi \in \mathcal{C}_b^2(\mathbb{R}^d, \mathbb{R})$

$$\begin{aligned} \frac{d}{dt} \int \varphi(v) f(v, t) dv &= \frac{1}{4} \sum_{i,j=1}^d \int_{\mathbb{R}^d \times \mathbb{R}^d} dv dv_* f(v, t) f(v_*, t) a_{ij}(v - v_*) (\partial_{ij} \varphi(v) + \partial_{ij} \varphi(v_*)) \\ &+ \frac{1}{2} \sum_{i=1}^d \int_{\mathbb{R}^d \times \mathbb{R}^d} dv dv_* f(v, t) f(v_*, t) b_i(v - v_*) (\partial_i \varphi(v) - \partial_i \varphi(v_*)) \end{aligned} \quad (5)$$

where $\partial_i \varphi = \frac{\partial \varphi}{\partial v_i}$ and $\partial_{ij} \varphi = \frac{\partial^2 \varphi}{\partial v_i \partial v_j}$.

The equation (5) conserves mass, momentum and energy. Thus, if there exists a weak function-solution f of (5) with an initial data satisfying $\int_{\mathbb{R}^d} f(v, 0) dv = 1$, the measure P_t on \mathbb{R}^d given by $P_t(dv) = f(v, t) dv$ is a probability measure, for any $t \geq 0$. Thus, we define a probabilistic notion of solutions of the Landau equation :

Definition 2 Let P_0 be a probability measure on \mathbb{R}^d with a finite two-order moment (i.e. $\int_{\mathbb{R}^d} |v|^2 P_0(dv) < \infty$). A measure-solution of the Landau equation (6) with initial data P_0 is a probability flow $(P_t)_{t \geq 0}$ on \mathbb{R}^d satisfying

$$\begin{aligned} \frac{d}{dt} \int \varphi(v) P_t(dv) &= \frac{1}{2} \sum_{i,j=1}^d \int_{\mathbb{R}^d} P_t(dv) \left(\int_{\mathbb{R}^d} P_t(dv_*) a_{ij}(v - v_*) \right) \partial_{ij} \varphi(v) \\ &+ \sum_{i=1}^d \int_{\mathbb{R}^d} P_t(dv) \left(\int_{\mathbb{R}^d} P_t(dv_*) b_i(v - v_*) \right) \partial_i \varphi(v) \end{aligned} \quad (6)$$

for any function $\varphi \in \mathcal{C}_b^2(\mathbb{R}^d, \mathbb{R})$.

This approach allows us to have weaker conditions on the initial data, i.e. we can assume that the initial data is a probability measure and not necessarily a density of probability.

Remark 3 With an abuse of notation, we will still say that a probability measure P on $\mathcal{C}([0, T], \mathbb{R}^d)$ is a measure-solution of the Landau equation when its time-marginals flow is a measure-solution in the sense of Definition 2.

We have already proved, in [3], the existence of a probability measure-solution of the Landau equation (6). We are here interested in proving with a stochastic approach the existence of a weak function-solution of (5) of class \mathcal{C}^∞ .

We recall briefly the main results of [3]. We have associated with the Landau equation (6) a nonlinear stochastic differential equation driven by a space-time white noise. We highlight the nonlinearity using two probability spaces: let $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ be a filtered probability space and $([0, 1], \mathcal{B}([0, 1]), d\alpha)$ be an auxiliary probability space, where $d\alpha$ is the Lebesgue measure on $[0, 1]$. In order to avoid any confusion, we will denote by E the expectation and \mathcal{L} the distribution of a random variable on $(\Omega, \mathcal{F}, \mathbb{P})$ and $E_\alpha, \mathcal{L}_\alpha$ for a random variable on $([0, 1], \mathcal{B}([0, 1]), d\alpha)$.

For $k \geq 2$, we define \mathcal{P}_k the space of continuous adapted processes $X = (X_t)_{t \geq 0}$ from $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ to \mathbb{R}^d , such that $E \left[\sup_{0 \leq t \leq T} |X_t|^k \right] < \infty \forall T > 0$, and $\mathcal{P}_{k,\alpha}$ the space of continuous processes $Y = (Y_t)_{t \geq 0}$ from $([0, 1], \mathcal{B}([0, 1]), d\alpha)$ to \mathbb{R}^d , such that $E_\alpha \left[\sup_{0 \leq t \leq T} |Y_t|^k \right] < \infty \forall T > 0$.

a is a nonnegative symmetric matrix, then there exists a matrix σ of order $d \times d'$ such that

$$a = \sigma \cdot \sigma^* \quad (7)$$

where σ^* is the adjoint matrix of σ .

We define a d' - dimensional space-time white noise on $[0, 1] \times [0, \infty)$ by

$$W^{d'} = \begin{pmatrix} W_1 \\ \vdots \\ W_{d'} \end{pmatrix} \quad (8)$$

where the W_i are independent space-time white noises with covariance measure $d\alpha dt$ on $[0, 1] \times [0, \infty)$ (according to Walsh's Definition, [8]).

Let X_0 be a random vector on \mathbb{R}^d , independent of $W^{d'}$, with a finite moment of order 2.

We consider the following nonlinear stochastic differential equation:

Definition 4 Let X_0 and $W^{d'}$ be defined as below. A couple of process (X, Y) on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}) \times ([0, 1], \mathcal{B}([0, 1]), d\alpha)$ is solution of the nonlinear stochastic differential equation (NSDE(σ, b)) if for any $t \geq 0$

$$X_t = X_0 + \int_0^t \int_0^1 \sigma(X_s - Y_s(\alpha)) \cdot W^{d'}(d\alpha, ds) + \int_0^t \int_0^1 b(X_s - Y_s(\alpha)) d\alpha ds$$

and $\mathcal{L}(X) = \mathcal{L}_\alpha(Y)$.

We notice, using Ito's Formula, that the distribution of a solution of (NSDE(σ, b)) is a weak measure-solution of the Landau equation (6) with initial data $P_0 = \mathcal{L}(X_0)$.

In [3] (Theorem 10), we have proved the following theorem for $k = 2$, but, adapting the proofs, it is still true for any $k \geq 2$.

Theorem 5 Assume that $W^{d'}$ is a d' -dimensional space-time white noise and assume that X_0 is an independent random vector on \mathbb{R}^d with finite moment of order k . If the functions σ and b , defined by (7), (2) and (4), are Lipschitz continuous, there exists a couple (X, Y) , unique in law, solution of the nonlinear equation (NSDE(σ, b)) with $(X, Y) \in \mathcal{P}_k \times \mathcal{P}_{k, \alpha}$.

Corollary 6 Assume that P_0 is a probability measure with a finite moment of order 2. There exists a measure-solution $(P_t)_{t \geq 0}$ with initial data P_0 to the Landau equation (6) when σ and b are Lipschitz continuous functions.

Corollary 7 Assume that P_0 is a probability measure with a finite moment of order 2. There is uniqueness of the measure-solution $(P_t)_{t \geq 0}$ with initial data P_0 to the Landau equation (6) when σ and b are Lipschitz continuous functions.

Proof.

- We just have to prove the uniqueness of the solution $(Q_t^\mu)_{t \geq 0}$ of the linear Landau equation, i.e.

$$\begin{aligned} \frac{d}{dt} \int \varphi(v) Q_t(dv) &= \frac{1}{2} \sum_{i,j=1}^d \int_{\mathbb{R}^d} Q_t(dv) \left(\int_{\mathbb{R}^d} \mu_t(dv_*) a_{ij}(v - v_*) \right) \partial_{ij} \varphi(v) \\ &+ \sum_{i=1}^d \int_{\mathbb{R}^d} Q_t(dv) \left(\int_{\mathbb{R}^d} \mu_t(dv_*) b_i(v - v_*) \right) \partial_i \varphi(v) \end{aligned} \quad (9)$$

where $(\mu_t)_{t \geq 0}$ is a probability flow on \mathbb{R}^d . In fact, as a measure-solution $(P_t)_{t \geq 0}$ to the (nonlinear) Landau equation is also a solution of the linear equation (9) with $\mu_t = P_t$ for any $t \geq 0$, by uniqueness of the measure-solution of (9), $(P_t)_{t \geq 0}$ is uniquely determined.

- The linear Landau equation (9) satisfies the assumptions of [2] Theorem 5.2, then there is uniqueness of the measure-solution of (9).

■

Remark 8 We notice that we can choose

$$\sigma(z) = \sqrt{h(|z|^2)} \begin{bmatrix} z_2 \\ -z_1 \end{bmatrix} \quad (10)$$

in dimension two, and

$$\sigma(z) = \sqrt{h(|z|^2)} \begin{bmatrix} z_2 & -z_3 & 0 \\ -z_1 & 0 & z_3 \\ 0 & z_1 & -z_2 \end{bmatrix} \quad (11)$$

in dimension three. Then, if h is a bounded function of class \mathcal{C}^1 with $h'(x) = O(\frac{1}{x^2})$ when $x \rightarrow +\infty$, σ and b are Lipschitz continuous functions of class \mathcal{C}^1 on \mathbb{R}^d , for $d = 2, 3$. This property can be generalized in higher dimension.

The aim of this article is to find a weak function-solution of the Landau equation when the initial data is a probability measure. To state the existence of a weak function-solution of (5) from a measure-solution, it is enough to show that the measure-solution is absolutely continuous with respect the Lebesgue measure. Indeed, if $(P_t)_{t \geq 0}$ is a measure-solution of (6) with initial data P_0 and if there exists a nonnegative function f_t on \mathbb{R}^d such that $P_t(dv) = f_t(v) dv$ for any $t > 0$, then the function f defined by $f(v, t) = f_t(v)$ for any $v \in \mathbb{R}^d$, $t > 0$, is a weak function-solution of (5) with initial data P_0 .

The idea consists in using the relation between the Landau equation and the nonlinear differential equation ($NSDE(\sigma, b)$). In fact, we develop a Malliavin Calculus for the value X_t , $t > 0$, of the solution X of ($NSDE(\sigma, b)$) obtained in Theorem 5, inspired by the methods used by V. Bally and E. Pardoux in [1] and by D. Nualart in [6].

The Maxwellian case (i.e., when the function h is a constant) is studied in detail with an analytic approach by C. Villani in [7]. When the initial data is a nonnegative function f_0 with finite mass and energy, C. Villani has proved the existence and the uniqueness of a solution of (1) of class \mathcal{C}^∞ .

We prove here the existence of a weak function-solution of the Landau equation (5) of class \mathcal{C}^∞ when the initial data is a probability measure with finite moments for some bounded functions h .

1.1 About the Malliavin calculus for a white noise

The Malliavin calculus is almost the same for the white noise as for the Brownian Motion. We use in the following the same notation as D. Nualart, [6]. We just recall the definitions of the main spaces for the Malliavin calculus.

Let $W^{d'}$ be a d' -dimensional space-time white noise.

Let \mathcal{S} be the class of random variables F having the following form

$$F = f\left(W^{d'}(g_1), \dots, W^{d'}(g_n)\right)$$

where f is a \mathcal{C}^∞ real values function on \mathbb{R}^n with partial derivatives having polynomial growth, $g = (g_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq d'}}$ is a matrix with components in $\mathbb{L}^2([0, 1] \times [0, \infty), d\alpha ds)$, and

$$W^{d'}(g_i) = \sum_{j=1}^{d'} \int_0^\infty \int_0^1 g_{ij}(s, \alpha) W_j(d\alpha, ds)$$

Assuming that $(r, z) \in [0, \infty) \times [0, 1]$, we define the first order Malliavin derivative $D_{(r,z)}^l F$ of F in relation to the l^{th} white noise W_l at point (r, z) , with $l \in \{1, \dots, d'\}$, by

$$D_{(r,z)}^l F = \sum_{i=1}^n \partial_i f \left(W^{d'}(g_1), \dots, W^{d'}(g_n) \right) g_{il}(r, z)$$

We will say that F is differentiable when all its first derivatives exist.

For $k \geq 1$, $\lambda_k = ((r_1, z_1), \dots, (r_k, z_k))$ with $r_m \in [0, \infty)$ and $z_m \in [0, 1]$, $m = 1, \dots, k$. We define by iteration the derivatives of order k . Let (l_1, \dots, l_k) be a k -uplet of $\{1, \dots, d'\}$, we denote by

$$D_{\lambda_k}^{l_1, \dots, l_k} F = D_{(r_k, z_k)}^{l_k} D_{(r_{k-1}, z_{k-1})}^{l_{k-1}} \dots D_{(r_1, z_1)}^{l_1} F$$

We will say that F has a derivative of order k when all its derivatives of order k exist.

We denote by $\mathbb{D}^{k,p}$ the completion of \mathcal{S} with respect to the norm

$$\|F\|_{k,p} = \left[E(|F|^p) + \sum_{m=1}^k \sum_{l_1, \dots, l_m=1}^{d'} E \left(\|D_{\lambda_m}^{l_1, \dots, l_m} F\|_{\mathbb{L}^2(\Lambda_m)}^p \right) \right]^{\frac{1}{p}}$$

where

$$\|D_{\lambda_m}^{l_1, \dots, l_m} F\|_{\mathbb{L}^2(\Lambda_m)}^2 = \int_{\Lambda_m} \left| D_{\lambda_m}^{l_1, \dots, l_m} F \right|^2 d\lambda_m$$

with $\Lambda_m = ([0, \infty) \times [0, 1])^m$ and $\lambda_m = ((r_1, z_1), \dots, (r_m, z_m)) \in \Lambda_m$.

We also denote by \mathbb{D}^∞ the subspace of the infinitely differentiable variables:

$$\mathbb{D}^\infty = \bigcap_{p \geq 1} \bigcap_{k \geq 1} \mathbb{D}^{k,p}$$

When F is a random vector in \mathbb{R}^d , we derive component by component and we denote by DF the matrix $(DF)_{i,l} = D^l F_i$, $1 \leq i \leq d, 1 \leq l \leq d'$. The Malliavin matrix is defined by

$$I = \int_0^T \int_0^1 D_{(r,z)} F \cdot (D_{(r,z)} F)^* dz dr$$

In this paper, under suitable assumptions on σ and b and integrability conditions on the initial data X_0 , we show that for any $t > 0$ the value X_t of X obtained in Theorem 5 satisfies the conditions of one of those two following theorems.

Theorem (a) (see [6] Theorem 2.1.2)

Let $F = (F_1, \dots, F_d)$ be a random vector on $(\Omega, \mathcal{F}, \mathbb{P})$ satisfying the following conditions :

(i) F_i belongs to the space $\mathbb{D}^{1,p}$, $p > 1$, for any $i = 1, \dots, d$.

(ii) The Malliavin matrix $I = \int_0^\infty \int_0^1 D_{(r,z)} F \cdot (D_{(r,z)} F)^* dz dr$ is invertible a.s..

Then the distribution of F is absolutely continuous with respect the Lebesgue measure on \mathbb{R}^d .

Theorem (b) (see [6] Corollary 2.1.2)

Let $F = (F_1, \dots, F_d)$ be a random vector on $(\Omega, \mathcal{F}, \mathbb{P})$ satisfying the following conditions :

(i) F_i belongs to \mathbb{D}^∞ , for any $i = 1, \dots, d$.

(ii) The Malliavin matrix $I = \int_0^\infty \int_0^1 D_{(r,z)} F \cdot (D_{(r,z)} F)^* dz dr$ satisfies

$$(\det I)^{-1} \in \bigcap_{p > 1} \mathbb{L}^p(\Omega)$$

Then F has an infinitely differentiable density.

1.2 Notations

- $\mathcal{C}([0, T], \mathbb{R}^d)$ is the space of continuous functions from $[0, T]$ to \mathbb{R}^d , and for $k \in \mathbb{N}$, $\mathcal{C}_b^k([0, T], \mathbb{R}^d)$ is the space of functions of class \mathcal{C}^k with all its derivatives bounded up to order k .
- $\mathcal{M}_{d,d'}(\mathbb{R})$ is the set of $d \times d'$ matrix on \mathbb{R} .
- For $k \geq 2$, a random variable Z on $(\Omega, \mathcal{F}, \mathbb{P})$ belongs to \mathbb{L}^k if Z has a finite moment of order k , i.e. $E[|Z|^k] < \infty$.
- K is an arbitrary notation for a positive constant (K can change from line to line).

2 Computation of the derivatives of X

2.1 The first derivative

Assumption (H^1): σ and b are Lipschitz continuous functions of class \mathcal{C}^1 from \mathbb{R}^d to $\mathcal{M}_{d,d'}(\mathbb{R})$ and \mathbb{R}^d respectively.

We denote by K_σ and K_b their Lipschitz constants.

Theorem 9 *We assume that X_0 has a finite 2-order moment. Let (X, Y) be the solution of the nonlinear stochastic differential equation (NSDE(σ, b)) obtained in theorem 5. (Y will play a parameter role in the following.)*

Under Assumption (H^1), $\forall t \in [0, T] \forall i = 1, \dots, d$, $X_{i,t} \in \mathbb{D}^{1,2}$. The i^{th} component of its derivative in relation to the l^{th} white noise at point $(r, z) \in [0, \infty) \times [0, 1]$ is given by

$$\begin{aligned} D_{(r,z)}^l X_{i,t} &= \sigma_{i,l}(X_r - Y_r(z)) \\ &+ \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k}(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} W_k(d\alpha, ds) \\ &+ \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \end{aligned}$$

if $t \geq r$ and $D_{(r,z)}^l X_{i,t} = 0$ if $t < r$.

Proof. We consider the Picard sequence of \mathcal{P}_2 -processes defined by

$$\begin{aligned} X_t^0 &= X_0 \\ X_t^{n+1} &= X_0 + \int_0^t \int_0^1 \sigma(X_s^n - Y_s(\alpha)) \cdot W^{d'}(d\alpha, ds) + \int_0^t \int_0^1 b(X_s^n - Y_s(\alpha)) d\alpha ds \end{aligned} \quad (12)$$

Then, the i^{th} component writes

$$X_{i,t}^{n+1} = X_{i,0} + \int_0^t \int_0^1 \sum_{k=1}^{d'} \sigma_{i,k}(X_s^n - Y_s(\alpha)) W_k(d\alpha, ds) + \int_0^t \int_0^1 b_i(X_s^n - Y_s(\alpha)) d\alpha ds$$

According to [3] Theorem 8, the sequence (X^n) satisfies

$$\sup_n E \left[\sup_{0 \leq r \leq T} |X_r^n|^2 \right] < \infty \quad (13)$$

and converges for the norm $\|U\| = \left\| \sup_{0 \leq t \leq T} U_t \right\|_{\mathbb{L}^2}$ toward X .

Let $T > 0$ be arbitrary fixed. Let $t \in [0, T]$ and $(r, z) \in [0, T] \times [0, 1]$ be fixed.

We show firstly by recurrence that for any $n \geq 0$ X_t^n is differentiable at point (r, z) in the Malliavin sense.

Recurrence Hypothesis:

- (i) $X_{i,t}^n \in \mathbb{D}^{1,2} \forall t \in [0, T] \quad \forall i = 1, \dots, d.$
- (ii) $\sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left(\int_0^\infty \int_0^1 |D_{(r,z)}^l X_t^n|^2 dz dr \right) < \infty$ where $|D_{(r,z)}^l X_t^n|^2 = \sum_{i=1}^d \left(D_{(r,z)}^l X_{i,t}^n \right)^2.$

For $n = 0$, Recurrence Hypothesis is satisfied.

We assume that it is true at rank n . Since σ and b are functions of class \mathcal{C}_b^1 , according to [6] proposition 1.2.2, $\forall i = 1, \dots, d \quad \forall k = 1, \dots, d'$, we have

$$\begin{aligned} \sigma_{i,k}(X_t^n - Y_t(\alpha)) &\in \mathbb{D}^{1,2} \\ b_i(X_t^n - Y_t(\alpha)) &\in \mathbb{D}^{1,2} \end{aligned}$$

As for the Brownian Motion, we can show that derivative and integral commute (see [6]), then $X_{i,t}^{n+1} \in \mathbb{D}^{1,2} \forall t \in [0, T] \quad \forall i = 1, \dots, d.$ Moreover, its derivative at point $(r, z) \in [0, T] \times [0, 1]$ in relation to the l^{th} white noise W_l is given by

$$\begin{aligned} D_{(r,z)}^l X_{i,t}^{n+1} &= \sigma_{i,l}(X_r^n - Y_r(z)) \\ &+ \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k}(X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n W_k(d\alpha, ds) \\ &+ \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i(X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n d\alpha ds \end{aligned}$$

if $r \leq t$, and $D_{(r,z)}^l X_{i,t}^{n+1} = 0$ else.

We still have to check that

$$\sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left[\int_0^\infty \int_0^1 |D_{(r,z)}^l X_t^{n+1}|^2 dz dr \right] < \infty$$

We define

$$\begin{aligned} S_n &= \sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left[\int_0^\infty \int_0^1 |D_{(r,z)}^l X_t^n|^2 dz dr \right] \\ &= \sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left[\int_0^t \int_0^1 |D_{(r,z)}^l X_t^n|^2 dz dr \right] \end{aligned}$$

According to Recurrence Hypothesis, $S_n < \infty$. Let us study S_{n+1} .

Let $l \in \{1, \dots, d'\}$ be arbitrary fixed. We divide in three parts the expectation $E \left[\int_0^t \int_0^1 |D_{(r,z)}^l X_{i,t}^{n+1}|^2 dz dr \right].$

We define

$$\begin{aligned} E_1 &= E \left[\int_0^t \int_0^1 |\sigma_{i,l}(X_r^n - Y_r(z))|^2 dz dr \right] \\ &\leq 2K_\sigma^2 E \left[\int_0^t \int_0^1 |X_r^n|^2 + |Y_r(z)|^2 dz dr \right] + T |\sigma(0)| \\ &\quad \text{since } \sigma \text{ is Lipschitz continuous} \\ &\leq 2K_\sigma^2 T \left[\sup_n E \left[\sup_{0 \leq r \leq T} |X_r^n|^2 \right] + E_\alpha \left[\sup_{0 \leq r \leq T} |Y_r|^2 \right] \right] + T |\sigma(0)| \end{aligned}$$

According to (13), we have $\sup_{0 \leq t \leq T} E_1 < \infty$.
We define

$$\begin{aligned}
E_2 &= E \left[\int_0^t \int_0^1 \left| \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k} (X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n W_k(d\alpha, ds) \right|^2 dzdr \right] \\
&= \int_0^t \int_0^1 E \left[\int_r^t \int_0^1 \sum_{k=1}^{d'} \left(\sum_{m=1}^d \partial_m \sigma_{i,k} (X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n \right)^2 d\alpha ds \right] dzdr \\
&\quad \text{since } W_k \text{ are independent} \\
&\leq d \sum_{k=1}^{d'} \sum_{m=1}^d \int_0^t \int_0^1 E \left[\int_r^t \int_0^1 \left(\partial_m \sigma_{i,k} (X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n \right)^2 d\alpha ds \right] dzdr \\
&\quad \text{using Hölder's Inequality}
\end{aligned}$$

Since the partial derivatives of σ are bounded by K_σ ,

$$\begin{aligned}
E_2 &\leq dK_\sigma^2 \sum_{k=1}^{d'} \sum_{m=1}^d \int_0^t \int_0^1 E \left[\int_r^t \int_0^1 \left[D_{(r,z)}^l X_{m,s}^n \right]^2 d\alpha ds \right] dzdr \\
&= d' dK_\sigma^2 \int_0^t E \left[\int_0^s \int_0^1 \left| D_{(r,z)}^l X_s^n \right|^2 dzdr \right] ds \\
&\quad \text{using Fubini's Theorem} \\
&\leq d' dK_\sigma^2 \int_0^t \sup_{s \in [0, T]} E \left[\int_0^s \int_0^1 \left| D_{(r,z)}^l X_s^n \right|^2 dzdr \right] ds \\
&\leq d' dK_\sigma^2 T S_n
\end{aligned}$$

Then, by Recurrence Hypothesis, we have $\sup_{0 \leq t \leq T} E_2 < \infty$.

We consider now

$$E_3 = E \left[\int_0^t \int_0^1 \left| \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i (X_s^n - Y_s(\alpha)) D_{(r,z)}^l X_{m,s}^n d\alpha ds \right|^2 dzdr \right]$$

Using the same method as for integral E_2 , we also have $\sup_{0 \leq t \leq T} E_3 < \infty$.

Finally, we have proved that for any $l \in \{1, \dots, d'\}$

$$\sup_{t \in [0, T]} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^{n+1} \right|^2 dzdr \right) \leq C_0 + C_1 T S_n < \infty$$

with

$$\begin{aligned}
C_0 &= 6dK_\sigma^2 T \left(\sup_n E \left(\sup_{0 \leq r \leq T} |X_r^n|^2 \right) + E_\alpha \left(\sup_{0 \leq r \leq T} |Y_r|^2 \right) \right) + 3dT |\sigma(0)| \\
C_1 &= 6d^2 \max(d' K_\sigma^2, K_b^2 T)
\end{aligned}$$

Thus, Recurrence Hypothesis is satisfied for any $n \geq 0$.

We notice that we have in fact a stronger result concerning property (ii):

Lemma 10 *The sequence of the first derivatives of $(X^n)_{n \geq 0}$ satisfies*

$$\sup_{n \geq 0} \sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^n \right|^2 dzdr \right) \leq C < \infty$$

Proof. We have already checked the following estimate

$$\begin{aligned} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^{n+1} \right|^2 dz dr \right) &\leq C_0 + C_1 \int_0^t S_n ds \\ &= C_0 + C_1 t S_n \end{aligned}$$

Since $S_0 = 0$,

$$\begin{aligned} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^1 \right|^2 dz dr \right) &\leq C_0 \\ &\vdots \\ E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^n \right|^2 dz dr \right) &\leq C_0 + C_0 C_1 t + \dots + C_0 C_1^{n-1} \frac{t^{n-1}}{(n-1)!} \end{aligned}$$

If we define $C = d' C_0 e^{C_1 T}$, we have

$$\sup_{n \geq 0} \sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^n \right|^2 dz dr \right) \leq C$$

■

Finally, we have proved

$$\forall n \geq 0 \quad \forall t \in [0, T] \quad \forall i = 1, \dots, d \quad X_{i,t}^n \in \mathbb{D}^{1,2} \quad (14)$$

$$\sup_{n \geq 0} \sup_{t \in [0, T]} \sum_{l=1}^{d'} E \left(\int_0^t \int_0^1 \left| D_{(r,z)}^l X_t^n \right|^2 dz dr \right) < \infty \quad (15)$$

Since the sequence (X^n) converges uniformly on $[0, T]$ in \mathbb{L}^2 toward X and thanks to (14) and (15), we deduce that X is differentiable (see [6] lemma 1.2.3). Moreover, the sequence of derivatives (DX^n) converges toward DX for the weak topology on $\mathbb{L}^2([0, T] \times [0, 1] \times \Omega)$. Thus, the theorem is proved. ■

2.2 The upper order derivatives

We state that X belongs to \mathbb{D}^∞ under a stronger assumption on σ and b .

Assumption (H^∞): σ and b are Lipschitz continuous functions of class \mathcal{C}^∞ with bounded derivatives from \mathbb{R}^d to $\mathcal{M}_{d,d'}(\mathbb{R})$ and \mathbb{R}^d respectively.

Notations: Let $k \geq 1$. We define $\lambda_k = ((r_1, z_1), \dots, (r_k, z_k))$ and

$$\hat{\lambda}_m = ((r_1, z_1), \dots, (r_{m-1}, z_{m-1}), (r_{m+1}, z_{m+1}), \dots, (r_k, z_k))$$

with $r_m \in [0, t]$ and $z_m \in [0, 1]$ for $m = 1, \dots, k$.

Let us now define $l(E) = l_{\varepsilon_1}, \dots, l_{\varepsilon_\eta}$ and $\lambda(E) = ((r_{\varepsilon_1}, z_{\varepsilon_1}), \dots, (r_{\varepsilon_\eta}, z_{\varepsilon_\eta}))$ for any subspace $E = \{\varepsilon_1, \dots, \varepsilon_\eta\}$ of $\{1, \dots, k\}$. We consider

$$\begin{aligned} \Sigma_{j, (l_1, \dots, l_k)}^i((s, \alpha), \lambda_k) &= \sum_{k_1, \dots, k_\nu=1}^d \sum_{k_1, \dots, k_\nu=1}^d (\partial_{k_1} \dots \partial_{k_\nu} \sigma_{i,j})(X_s - Y_s(\alpha)) \times D_{\lambda(E_1)}^{l(E_1)} X_{k_1, s} \dots \times D_{\lambda(E_\nu)}^{l(E_\nu)} X_{k_\nu, s} \\ \beta_{(l_1, \dots, l_k)}^i((s, \alpha), \lambda_k) &= \sum_{k_1, \dots, k_\nu=1}^d \sum_{k_1, \dots, k_\nu=1}^d (\partial_{k_1} \dots \partial_{k_\nu} b_i)(X_s - Y_s(\alpha)) \times D_{\lambda(E_1)}^{l(E_1)} X_{k_1, s} \dots \times D_{\lambda(E_\nu)}^{l(E_\nu)} X_{k_\nu, s} \end{aligned}$$

where the first sum is taken on all partitions $E_1 \cup \dots \cup E_\nu = \{1, \dots, k\}$.

We define at last,

$$\Sigma_j^i((s, \alpha)) = \sigma_{ij}(X_s - Y_s(\alpha))$$

We denote by $r_1 \vee \dots \vee r_k = \sup\{r_1, \dots, r_k\}$.

Theorem 11 *Assume that $X_0 \in \mathbb{L}^p$, for any $p \geq 1$. Under Assumption (H^∞) , $\forall t \geq 0$ $X_t \in \mathbb{D}^\infty$. Moreover, the i^{th} component of one of its derivative of order k at point $\lambda_k = ((r_1, z_1), \dots, (r_k, z_k))$ is given by the following equation*

$$\begin{aligned} D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t} &= \sum_{m=1}^k \Sigma_{l_m, (l_1, \dots, l_{m-1}, l_{m+1}, \dots, l_k)}^i \left((r_m, z_m), \hat{\lambda}_m \right) + \int_{r_1 \vee \dots \vee r_k}^t \int_0^1 \sum_{j=1}^{d'} \Sigma_{j, (l_1, \dots, l_k)}^i((s, \alpha), \lambda_k) W_j(d\alpha, ds) \\ &\quad + \int_{r_1 \vee \dots \vee r_k}^t \int_0^1 \beta_{(l_1, \dots, l_k)}^i((s, \alpha), \lambda_k) d\alpha ds \end{aligned} \quad (16)$$

if $t \geq r_1 \vee \dots \vee r_k$ and $D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t} = 0$ if $t < r_1 \vee \dots \vee r_k$.

Remark 12 *In expression (16) of the k^{th} derivative, the terms in the first sum with $r_m < r_1 \vee \dots \vee r_k$ are equal to 0.*

Proof. We use again the Picard sequence $(X^n)_{n \geq 0}$ defined by (12). For any $p \geq 2$, $n \geq 0$, $X^n \in \mathbb{L}^p$ and (X^n) converges uniformly toward X in \mathbb{L}^p . As σ and b satisfy Assumption (H^∞) , using the same method as in the previous paragraph, we prove that $X_t^n \in \mathbb{D}^{1,p} \forall p \geq 1$ for any $t \geq 0$. By recurrence, we prove that $\forall t \geq 0, \forall n \geq 0$ $X_t^n \in \mathbb{D}^\infty$.

Let us fix $T > 0$.

Recurrence Hypothesis (h_n) :

(i) $X_{i,t}^n \in \mathbb{D}^\infty, \forall t \in [0, T], \forall i = 1, \dots, d$.

(ii) $\sup_{t \in [0, T]} \sum_{l_1, \dots, l_k=1}^{d'} E \left(\int_{([0, t] \times [0, 1])^k} \left| D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t}^n \right|^p d\lambda_k \right) < \infty \quad \forall p \geq 1, \forall k \geq 1$.

(iii) the derivatives of order k have the following expression:

$$\begin{aligned} D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t}^{n+1} &= \sum_{m=1}^k \Sigma_{l_m, (l_1, \dots, l_{m-1}, l_{m+1}, \dots, l_k)}^{n,i} \left((r_m, z_m), \hat{\lambda}_m \right) + \int_{r_1 \vee \dots \vee r_k}^t \int_0^1 \sum_{j=1}^{d'} \Sigma_{j, (l_1, \dots, l_k)}^{n,i}((s, \alpha), \lambda_k) W_j(d\alpha, ds) \\ &\quad + \int_{r_1 \vee \dots \vee r_k}^t \int_0^1 \beta_{(l_1, \dots, l_k)}^{n,i}((s, \alpha), \lambda_k) d\alpha ds \end{aligned} \quad (17)$$

if $t \geq r_1 \vee \dots \vee r_k$ and $D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t}^n = 0$ else, where Σ^n and β^n are defined as Σ and β replacing X with X^n .

Hypothesis (h_0) is satisfied.

Let us assume that Hypothesis (h_n) is true, and let us study (h_{n+1}) . According to Assumption (H^∞) and adapting the computation of the first derivative, it is easy to state that the two first properties are satisfied. We just check the expression of the k^{th} derivative by recurrence on k .

For $k = 1$, we have

$$\begin{aligned} D_{(r,z)}^l X_{i,t}^{n+1} &= \Sigma_l^{n,i}((r, z)) + \int_r^t \int_0^1 \sum_{j=1}^{d'} \Sigma_{j, (l)}^{n,i}((s, \alpha), (r, z)) W_j(d\alpha, ds) \\ &\quad + \int_r^t \int_0^1 \beta_{(l)}^{n,i}((s, \alpha), (r, z)) d\alpha ds \end{aligned}$$

then the expression (17) is satisfied.

We assume that the expression (17) of the k^{th} derivative is true, and we now compute the derivative of order $k + 1$

$$\begin{aligned}
D_{(r_{k+1}, z_{k+1})}^{l_{k+1}} \left(D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t}^{n+1} \right) &= D_{(r_{k+1}, z_{k+1})}^{l_{k+1}} \left(\sum_{m=1}^k \Sigma_{l_m, (l_1, \dots, l_{m-1}, l_{m+1}, \dots, l_k)}^{n,i} \left((r_m, z_m), \hat{\lambda}_m \right) \right) \\
&\quad + \Sigma_{l_{k+1}, (l_1, \dots, l_k)}^{n,i} \left((r_{k+1}, z_{k+1}), \lambda_k \right) \\
&\quad + \int_{r_1 \vee \dots \vee r_k \vee r_{k+1}}^t \int_0^1 \sum_{j=1}^{d'} D_{(r_{k+1}, z_{k+1})}^{l_{k+1}} \left(\Sigma_{j, (l_1, \dots, l_k)}^{n,i} \left((s, \alpha), \lambda_k \right) \right) W_j (d\alpha, ds) \\
&\quad + \int_{r_1 \vee \dots \vee r_k \vee r_{k+1}}^t \int_0^1 D_{(r_{k+1}, z_{k+1})}^{l_{k+1}} \left(\beta_{(l_1, \dots, l_k)}^{n,i} \left((s, \alpha), \lambda_k \right) \right) d\alpha ds
\end{aligned}$$

Using some elementary computations, we obtain

$$\begin{aligned}
D_{(r_{k+1}, z_{k+1})}^{l_{k+1}} \left(D_{\lambda_k}^{l_1, \dots, l_k} X_{i,t}^{n+1} \right) &= \sum_{m=1}^k \Sigma_{l_m, (l_1, \dots, l_{m-1}, l_{m+1}, \dots, l_k, l_{k+1})}^{n,i} \left((r_m, z_m), \hat{\lambda}_m \right) \\
&\quad + \Sigma_{l_{k+1}, (l_1, \dots, l_k)}^{n,i} \left((r_{k+1}, z_{k+1}), \hat{\lambda}_{k+1} \right) \\
&\quad + \int_{r_1 \vee \dots \vee r_{k+1}}^t \int_0^1 \sum_{j=1}^{d'} \Sigma_{j, (l_1, \dots, l_k, l_{k+1})}^{n,i} \left((s, \alpha), \lambda_{k+1} \right) W_j (d\alpha, ds) \\
&\quad + \int_{r_1 \vee \dots \vee r_{k+1}}^t \int_0^1 \beta_{(l_1, \dots, l_k, l_{k+1})}^{n,i} \left((s, \alpha), \lambda_k \right) d\alpha ds
\end{aligned}$$

So by recurrence, the property (iii) of (h_{n+1}) is proved and consequently for any $n \geq 0$ Recurrence Hypothesis (h_n) is satisfied.

Moreover, as in the computation of the first derivative, we have a stronger property than (ii) in (h_n) :

Lemma 13 *If we denote by*

$$\begin{aligned}
S_{n,k}(t) &= \sum_{l_1, \dots, l_k=1}^{d'} E \left(\int_{([0,t] \times [0,1])^k} \left| D_{\lambda_k}^{l_1, \dots, l_k} X_t^n \right|^p d\lambda_k \right) \\
M_k &= \sup_{0 \leq q \leq k} \sup_{n \geq 0} \sup_{t \in [0, T]} S_{n,q}(t)
\end{aligned}$$

then for any $k \geq 1$

$$M_k < \infty$$

Proof. The proof is similar to the proof of Lemma 10. ■

As (X^n) converges toward X in \mathbb{L}^p uniformly on $[0, T]$ for any $T > 0$, the process X satisfies the conditions of lemma 1.5.4 in [6]. Then, the theorem is proved. ■

3 Existence of a weak function-solution of the Landau equation

Under some suitable conditions on the function h , the Landau coefficients satisfy Assumption (H^1) (see Remark 8). Consequently, if X_0 belongs to \mathbb{L}^2 , the process X solution of $(NSDE(\sigma, b))$ is differentiable in the Malliavin sense. Let us now study the Malliavin matrix $I_t = \int_0^T \int_0^1 D_{(r,z)} X_t \cdot (D_{(r,z)} X_t)^* dz dr$ for any $t > 0$ to state the following theorem.

Theorem 14 Assume that X_0 is a \mathbb{R}^d -valued random vector with a finite 2-order moment. Let σ and b be the coefficients of the Landau equation defined respectively by (7), (2) and (4). We assume that σ and b are Lipschitz continuous of class \mathcal{C}^1 . If the distribution of X_0 is not a Dirac mass and if we denote by (X, Y) the solution of the nonlinear stochastic differential equation

$$X_t = X_0 + \int_0^t \int_0^1 \sigma(X_s - Y_s(\alpha)) \cdot W^{d'}(d\alpha, ds) + \int_0^t \int_0^1 b(X_s - Y_s(\alpha)) d\alpha ds \quad (NSDE(\sigma, b))$$

then, for any $t > 0$ the distribution P_t of X_t is absolutely continuous with respect the Lebesgue measure.

Corollary 15 Let P_0 be a probability measure such that $\int |x|^2 P_0(dx) < \infty$. Let σ and b be the coefficients of the Landau equation defined respectively by (7), (2) and (4). We assume that σ and b are Lipschitz continuous of class \mathcal{C}^1 . If P_0 is not a Dirac measure, there exists a unique weak function-solution of the Landau equation with initial data P_0 .

Proof. (Corollary 15)

Let X_0 be a random vector with distribution P_0 and X be a solution of $(NSDE(\sigma, b))$ with initial data X_0 . If we denote by f_t the density of the distribution of X_t , then, using Itô's Formula, the function f , defined by $f(x, t) = f_t(x)$ for $t > 0$, is a weak function solution of the Landau equation (5) with initial data P_0 .

The uniqueness is given by Corollary 7. ■

Remark 16 Without any restriction, we can assume that $\mathbf{E}[X_0] = \mathbf{0}$ to simplify the computations.

Proof. (Remark 16)

By conservation of momentum, if we define for any $t \geq 0$, $X'_t = X_t - E[X_0]$, the expectation of X' is equal to 0 and X' satisfies the following equation

$$X'_t = X'_0 + \int_0^t \int_0^1 \sigma(X'_s - Y'_s(\alpha)) \cdot W^{d'}(d\alpha, ds) + \int_0^t \int_0^1 b(X'_s - Y'_s(\alpha)) d\alpha ds$$

with $Y'_s(\alpha) = Y_s(\alpha) - E[X_0]$.

As $\mathcal{L}(X) = \mathcal{L}_\alpha(Y)$, we also have $\mathcal{L}(X') = \mathcal{L}_\alpha(Y')$.

If we prove that the distribution of X'_t has a density f'_t with respect the Lebesgue measure, then X_t has a density given by $f_t(z) = f'_t(z - E[X_0])$. ■

Proof. (Theorem 14)

We recall the expression of the first Malliavin derivative at point $(r, z) \in [0, \infty) \times [0, 1]$ of X :

$$\begin{aligned} D_{(r,z)}^l X_{i,t} &= \sigma_{i,l}(X_r - Y_r(z)) \\ &+ \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k}(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} W_k(d\alpha, ds) \\ &+ \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \end{aligned}$$

if $t \geq r$ and $D_{(r,z)}^l X_{i,t} = 0$ else.

We fix $(r, z) \in [0, \infty) \times [0, 1]$ and we define

$$\begin{aligned} S_k(\cdot) &= (\partial_m \sigma_{i,k}(\cdot))_{1 \leq i, m \leq d} \\ B(\cdot) &= (\partial_m b_i(\cdot))_{1 \leq i, m \leq d} \end{aligned}$$

Thus we give a matricial expression of the derivative of X

$$\begin{aligned} D_{(r,z)}X_t &= \begin{cases} \sigma(X_r - Y_r(z)) + \int_r^t \int_0^1 \sum_{k=1}^{d'} S_k(X_s - Y_s(\alpha)) \cdot D_{(r,z)}X_s W_k(d\alpha, ds) \\ + \int_r^t \int_0^1 B(X_s - Y_s(\alpha)) \cdot D_{(r,z)}X_s d\alpha ds \end{cases} \\ &= \begin{cases} \sigma(X_r - Y_r(z)) + \int_r^t \int_0^1 \sum_{k=1}^{d'} S_k(X_s - Y_s(\alpha)) \cdot D_{(r,z)}X_s W_k(d\alpha, ds) \\ + \int_r^t \int_0^1 B(X_s - Y_s(\alpha)) \cdot D_{(r,z)}X_s d\alpha ds \end{cases} \\ D_{(r,z)}X_t &= 0 \end{aligned} \quad \text{if } t < r$$

Let us define the semimartingale Z^r , for any $t \geq r$,

$$Z_t^r = \int_r^t \int_0^1 \sum_{k=1}^{d'} S_k(X_s - Y_s(\alpha)) W_k(d\alpha, ds) + \int_r^t \int_0^1 B(X_s - Y_s(\alpha)) d\alpha ds$$

As S and B are bounded, $(Z_t^r)_{t \geq r}$ is a continuous semimartingale and the first derivative satisfies the equation

$$D_{(r,z)}X_t = \sigma(X_r - Y_r(z)) + \int_r^t dZ_s^r \cdot D_{(r,z)}X_s \quad (18)$$

Using the results of [4], there is a unique solution of (18) defined almost surely, for any $t \geq r$ by

$$D_{(r,z)}X_t = \mathcal{E}(Z_t^r) \cdot \sigma(X_r - Y_r(z))$$

with $\mathcal{E}(Z_t^r)$ invertible for any $t \geq r$.

We fix now $t > 0$.

We want to apply Theorem (a), thus study if the Malliavin matrix I_t is invertible a.s..

$$\begin{aligned} I_t &= \int_0^\infty \int_0^1 D_{(r,z)}X_t \cdot (D_{(r,z)}X_t)^* dr dz \\ &= \int_0^t \int_0^1 D_{(r,z)}X_t \cdot (D_{(r,z)}X_t)^* dr dz \\ &= \int_0^t \int_0^1 \mathcal{E}(Z_t^r) \cdot \sigma(X_r - Y_r(z)) \cdot \sigma^*(X_r - Y_r(z)) \cdot (\mathcal{E}(Z_t^r))^* dr dz \\ &= \int_0^t \mathcal{E}(Z_t^r) \cdot \left(\int_0^1 a(X_r - Y_r(z)) dz \right) \cdot (\mathcal{E}(Z_t^r))^* dr \end{aligned}$$

I_t is a nonnegative symmetric matrix, then I_t is invertible if and only if $V^* \cdot I_t \cdot V > 0$ for any $V \in \mathbb{R}^d \setminus \{0\}$.

We define $\Gamma_r = \int_0^1 a(X_r - Y_r(z)) dz$.

We prove the theorem by contradiction.

Assumption : let us suppose that I_t is not ‘‘invertible a.s.’’.

Then, there exists a subset $\Omega_1 \subset \Omega$, $\mathbb{P}(\Omega_1) > 0$, such that $\forall \omega \in \Omega_1$ $I_t(\omega)$ is not invertible.

Let Ω_2 be such that $\mathbb{P}(\Omega_2) = 1$ and $\forall \omega \in \Omega_2$ $\forall r \leq t$, $\mathcal{E}(Z_t^r)(\omega)$ is invertible.

We define $\Omega_0 = \Omega_1 \cap \Omega_2$, and we notice that $\mathbb{P}(\Omega_0) > 0$.

We fix $\omega \in \Omega_0$.

As $I_t(\omega)$ is not invertible, there exists a vector $V_\omega \in \mathbb{R}^d \setminus \{0\}$ such that

$$\begin{aligned} V_\omega^* \cdot I_t(\omega) \cdot V_\omega &= \int_0^t V_\omega^* \cdot \mathcal{E}(Z_t^r)(\omega) \cdot \Gamma_r(\omega) \cdot (\mathcal{E}(Z_t^r)(\omega))^* \cdot V_\omega dr \\ &= 0 \end{aligned}$$

As for any $r \leq t$, $\mathcal{E}(Z)_t^r(\omega) \cdot \Gamma_r(\omega) \cdot (\mathcal{E}(Z)_t^r(\omega))^*$ is a nonnegative symmetric matrix, we notice that

$$V_\omega^* \cdot \mathcal{E}(Z)_t^r(\omega) \cdot \Gamma_r(\omega) \cdot (\mathcal{E}(Z)_t^r(\omega))^* \cdot V_\omega \geq 0$$

Then, on a subset J_ω of full measure in $[0, t]$, $V_\omega^* \cdot \mathcal{E}(Z)_t^r(\omega) \cdot \Gamma_r(\omega) \cdot (\mathcal{E}(Z)_t^r(\omega))^* \cdot V_\omega = 0$. This implies that $\forall r \in J_\omega$, $\mathcal{E}(Z)_t^r(\omega) \cdot \Gamma_r(\omega) \cdot (\mathcal{E}(Z)_t^r(\omega))^*$ is a non invertible matrix. However, since $\Omega_0 \subset \Omega_2$, $\mathcal{E}(Z)_t^r(\omega)$ is invertible for any $r \leq t$, and consequently $\Gamma_r(\omega)$ is not invertible for $r \in J_\omega$.

Let us now study if the situation “ $\Gamma_r(\omega)$ non invertible for almost all r ” is possible.

Using Lebesgue’s Theorem, we notice that the mapping $r \rightarrow \Gamma_r(\omega)$ is continuous. Consequently, $\Gamma_r(\omega)$ non invertible for almost all r implies that $\Gamma_r(\omega)$ is non invertible for any $r \in [0, t]$.

Let $V = (V_i)_{1 \leq i \leq d}$ be a vector in $\mathbb{R}^d \setminus \{0\}$,

$$\begin{aligned} V^* \cdot \Gamma_r(\omega) \cdot V &= \int_0^1 h(|X_r(\omega) - Y_r(z)|^2) \left[|V|^2 |X_r(\omega) - Y_r(z)|^2 - \left(\sum_{i=1}^d V_i (X_{i,r}(\omega) - Y_{i,r}(z)) \right)^2 \right] dz \\ &\geq m \left[|V|^2 |X_r(\omega)|^2 - \left(\sum_{i=1}^d V_i X_{i,r}(\omega) \right)^2 + |V|^2 E|X_r|^2 - \sum_{i=1}^d V_i^2 E[X_{i,r}^2] + \sum_{i \neq j} V_i V_j E[X_{i,r} X_{j,r}] \right] \\ &\quad \text{using the lower bound (3) of } h, E[X_t] = 0 \text{ and } \mathcal{L}(X_t) = \mathcal{L}_\alpha(Y_t) \quad \forall t \geq 0 \end{aligned}$$

Using Cauchy-Schwarz’s inequality, we notice that $|V|^2 |X_r(\omega)|^2 - \left(\sum_{i=1}^d V_i X_{i,r}(\omega) \right)^2 \geq 0$. Then,

$$V^* \cdot \Gamma_r(\omega) \cdot V \geq m \left(E[|X_r|^2] |V|^2 - \sum_{i=1}^d V_i^2 E[X_{i,r}^2] - \sum_{i \neq j} V_i V_j E[X_{i,r} X_{j,r}] \right) \quad (19)$$

$\Gamma_r(\omega)$ non invertible means that for any $r \in [0, t]$ there exists $V_r(\omega) \in \mathbb{R}^d \setminus \{0\}$ such that $V_r(\omega)^* \cdot \Gamma_r(\omega) \cdot V_r(\omega) = 0$. Nevertheless, using expression (19), this implies that there is equality in Cauchy-Schwarz, i.e. for any i, j such that $V_{i,r}(\omega) \neq 0$ and $V_{j,r}(\omega) \neq 0$

$$E[X_{i,r} X_{j,r}]^2 = E[X_{i,r}^2] E[X_{j,r}^2] \quad (20)$$

We notice that the equality in Cauchy-Schwarz (20) is in fact satisfied for any $i, j \in \{1, \dots, d\}$. Indeed, if one of the components of $V_r(\omega)$ is equal to 0, for example $V_{1,r}(\omega) = 0$, then

$$\begin{aligned} V_r(\omega)^* \cdot \Gamma_r(\omega) \cdot V_r(\omega) &\geq m \left[E[|X_r|^2] |V_r(\omega)|^2 - \left(\sum_{i=1}^d |V_{i,r}(\omega)| \sqrt{E[X_{i,r}^2]} \right)^2 \right] \\ &= m \left[E[(X_{1,r})^2] |V_r(\omega)|^2 + |V_r(\omega)|^2 \sum_{i=2}^d E[(X_{i,r})^2] - \left(\sum_{i=2}^d |V_{i,r}(\omega)| \sqrt{E[X_{i,r}^2]} \right)^2 \right] \\ &\geq m E[(X_{1,r})^2] |V_r(\omega)|^2 \end{aligned}$$

As $V_r(\omega)^* \cdot \Gamma_r(\omega) \cdot V_r(\omega) = 0$ and $|V_r(\omega)| \neq 0$, $E[(X_{1,r})^2] = 0$. Consequently, for any j , we still have $E[X_{1,r} X_{j,r}]^2 = E[X_{1,r}^2] E[X_{j,r}^2]$.

Finally, for any $r \in [0, t]$, there exists a vector $\lambda_r \in \mathbb{R}^d$ and a real-valued random variable U_r such that, for any $\omega \in \Omega$,

$$X_r(\omega) = U_r(\omega) \lambda_r \quad (21)$$

As $E \left[|X_0|^2 \right] \neq 0$ (the distribution of X_0 is not a Dirac mass), using the conservation of energy, $X_r \neq 0$ a.s., then $\lambda_r \neq 0$ and $U_r \neq 0$ a.s.. We can suppose $|\lambda_r| = 1$ for any $r \geq 0$. Then, by conservation of momentum and energy, we notice that $E[U_r] = 0$ and $E[U_r^2] = E[|X_0|^2]$.

The distribution of a solution of $(NSDE(\sigma, b))$ is a measure-solution of the Landau equation (6). Then, we will now study if the distribution of a process defined by (21) can be a solution of the Landau equation. We denote by Q the distribution of U and by Q_t the distribution on \mathbb{R} of U_t . Using (21), the equation (6) writes

$$\begin{aligned} \frac{d}{dt} \int_{\mathbb{R}} \varphi(\lambda_t x) Q_t(dx) &= \frac{1}{2} \sum_{i,j=1}^d \int_{\mathbb{R} \times \mathbb{R}} a_{ij}(\lambda_t(x-y)) \partial_{ij} \varphi(\lambda_t x) Q_t(dx) Q_t(dy) \\ &+ \sum_{i=1}^d \int_{\mathbb{R} \times \mathbb{R}} b_i(\lambda_t(x-y)) \partial_i \varphi(\lambda_t x) Q_t(dx) Q_t(dy) \end{aligned}$$

for any test function $\varphi \in \mathcal{C}_b^2(\mathbb{R}^d, \mathbb{R})$.

As $|\lambda_t| = 1$, for $i, j \in \{1, \dots, d\}$

$$\begin{aligned} a_{ij}(\lambda_t(x-y)) &= (x-y)^2 h\left((x-y)^2\right) (\delta_{ij} - \lambda_{i,t} \lambda_{j,t}) \\ b_i(\lambda_t(x-y)) &= -(d-1)(x-y) h\left((x-y)^2\right) \lambda_{i,t} \end{aligned}$$

Then,

$$\begin{aligned} \frac{d}{dt} \int_{\mathbb{R}} \varphi(\lambda_t x) Q_t(dx) &= \frac{1}{2} \sum_{i,j=1}^d (\delta_{ij} - \lambda_{i,t} \lambda_{j,t}) \int_{\mathbb{R} \times \mathbb{R}} (x-y)^2 h\left((x-y)^2\right) \partial_{ij} \varphi(\lambda_t x) Q_t(dx) Q_t(dy) \\ &- (d-1) \sum_{i=1}^d \lambda_{i,t} \int_{\mathbb{R} \times \mathbb{R}} (x-y) h\left((x-y)^2\right) \partial_i \varphi(\lambda_t x) Q_t(dx) Q_t(dy) \end{aligned}$$

We now explicit the equation satisfied by the 2-order moments of X : let $k, l \in \mathbb{N}$, $k \neq l$. Using $\varphi(v) = v_k^2$ or $\varphi(v) = v_k v_l$, $v \in \mathbb{R}^d$, we obtain

$$\begin{aligned} \frac{d}{dt} \int_{\mathbb{R}} \lambda_{k,t}^2 x^2 Q_t(dx) &= E \left[|X_0|^2 \right] \frac{d}{dt} \lambda_{k,t}^2 \\ &= (1 - d\lambda_{k,t}^2) \int_{\mathbb{R} \times \mathbb{R}} (x-y)^2 h\left((x-y)^2\right) Q_t(dx) Q_t(dy) \\ \frac{d}{dt} \int_{\mathbb{R}} \lambda_{k,t} \lambda_{l,t} x^2 Q_t(dx) &= E \left[|X_0|^2 \right] \frac{d}{dt} \lambda_{k,t} \lambda_{l,t} \\ &= -d\lambda_{k,t} \lambda_{l,t} \int_{\mathbb{R} \times \mathbb{R}} (x-y)^2 h\left((x-y)^2\right) Q_t(dx) Q_t(dy) \end{aligned}$$

Let us define $f(t) = \int_{\mathbb{R} \times \mathbb{R}} (x-y)^2 h\left((x-y)^2\right) Q_t(dx) Q_t(dy)$. As h satisfies (3) and $E[|X_0|^2] \neq 0$, for any $t \geq 0$, we notice that $f(t) > 0$.

Let us now compute $\frac{d}{dt} (\lambda_{k,t}^2 \lambda_{l,t}^2)$, using two different ways :

$$\begin{aligned} E \left[|X_0|^2 \right] \frac{d}{dt} (\lambda_{k,t}^2 \lambda_{l,t}^2) &= \lambda_{k,t}^2 E \left[|X_0|^2 \right] \frac{d}{dt} \lambda_{l,t}^2 + \lambda_{l,t}^2 E \left[|X_0|^2 \right] \frac{d}{dt} \lambda_{k,t}^2 \\ &= \lambda_{k,t}^2 f(t) + \lambda_{l,t}^2 f(t) - 2d\lambda_{k,t}^2 \lambda_{l,t}^2 f(t) \\ E \left[|X_0|^2 \right] \frac{d}{dt} (\lambda_{k,t} \lambda_{l,t}) &= 2\lambda_{k,t} \lambda_{l,t} E \left[|X_0|^2 \right] \frac{d}{dt} (\lambda_{k,t} \lambda_{l,t}) \\ &= -2d\lambda_{k,t} \lambda_{l,t} f(t) \end{aligned}$$

Then $\lambda_t = 0$, consequently $X_t = 0$ which is impossible.

Finally, I_t is invertible a.s. for any $t > 0$ and according to theorem (a), the theorem is proved.

■

Remark 17 We notice that the matrix $\Gamma_r = \int_0^1 a(X_r - Y_r(z)) dz$ is invertible a.s., whereas $\det(a(X_r - Y_r(z))) = 0$ for any r, z . In fact, thanks to the nonlinearity of equation (NSDE(σ, b)), we can conclude that the Malliavin matrix has an inverse a.s..

Remark 18 A consequence of Theorem 14 is, for any $i, j \in \{1, \dots, d\}$, $i \neq j$, for any $t > 0$

$$E[X_{i,t}X_{j,t}]^2 < E[X_{i,t}^2] E[X_{j,t}^2]$$

if the random vector X_0 is not a constant.

4 Regularity of the weak function-solution

Theorem 19 Let X_0 be a random vector such that $E[|X_0|^p] < \infty$ for any $p \geq 1$. Let σ and b be the coefficients of the Landau equation respectively defined by (7), (2) and (4). We assume that σ and b are Lipschitz continuous and infinitely differentiable with bounded derivatives. If the distribution of X_0 is not a Dirac mass and if we denote by X the solution of the nonlinear stochastic differential equation

$$X_t = X_0 + \int_0^t \int_0^1 \sigma(X_s - Y_s(\alpha)) . W^d(d\alpha, ds) + \int_0^t \int_0^1 b(X_s - Y_s(\alpha)) d\alpha ds \quad (\text{NSDE}(\sigma, b))$$

then for any $t > 0$ the distribution of X_t has a density of class \mathcal{C}^∞ with respect the Lebesgue measure on \mathbb{R}^d .

Corollary 20 Let P_0 be a probability measure such that $\int |x|^p P_0(dx) < \infty$ for any $p \geq 1$. Let σ and b be the coefficients of the Landau equation defined respectively by (7), (2) and (4). We assume that σ and b are Lipschitz continuous and infinitely differentiable with bounded derivatives. If P_0 is not a Dirac measure, there exists a unique weak function solution of the Landau equation of class \mathcal{C}^∞ with initial data P_0 .

Remark 21 Using the expressions (10) or (11), we notice that, if h is a bounded function of class \mathcal{C}^∞ such that $h^{(l)}(x) = O\left(\frac{1}{x^{l+1}}\right)$ when $x \rightarrow +\infty$ for any $l \geq 1$, σ and b are Lipschitz continuous functions of class \mathcal{C}^∞ with bounded derivatives.

Proof. As in the previous part, we assume that $E[X_0] = 0$ to simplify the computations. As σ and b satisfy Assumption (H^∞), the process X is infinitely differentiable in the Malliavin sense. We need to study the moments of the inverse of the determinant of the Malliavin matrix I_t at time t , for any $t > 0$, to apply Theorem (b). The expression of the determinant is complex, nevertheless we can notice that in dimension d ,

$$(\det I_t)^{1/d} \geq \inf_{|V|=1} \langle I_t V, V \rangle$$

where $\langle \cdot, \cdot \rangle$ is the euclidean scalar product in \mathbb{R}^d .

Moreover, see P. L. Morien [5] lemma 10.5.1, the property (ii) of theorem (b) is satisfied as soon as for any $k \in \mathbb{N}$

$$\lim_{\varepsilon \rightarrow 0} \varepsilon^{-k} \mathbb{P}\left((\det I_t)^{1/d} < c\varepsilon\right) = 0 \quad (22)$$

where c is a positive constant which will be computed later. Indeed, let us fix $p \geq 1$ and $k \in \mathbb{N}$ such that $k > d(p+1)$. If (22) is satisfied, there exists $\varepsilon_0 > 0$ such that $\varepsilon^{-k} \mathbb{P}\left((\det I_t)^{1/d} < c\varepsilon\right) < 1$ for any $\varepsilon < \varepsilon_0$. Let $m_0 \in \mathbb{N}$ such that $\forall m > m_0, \frac{1}{m} < (c\varepsilon_0)^d$, then

$$\begin{aligned} E\left[\frac{1}{(\det I_t)^p}\right] &\leq 1 + \sum_{m=1}^{\infty} (m+1)^p \mathbb{P}\left(m \leq \frac{1}{\det I_t} \leq m+1\right) \\ &\leq 1 + 2^p \sum_{m=1}^{\infty} m^p \mathbb{P}\left(\det I_t \leq \frac{1}{m}\right) \\ &\leq 1 + 2^p \sum_{m=1}^{m_0} m^p \mathbb{P}\left(\det I_t \leq \frac{1}{m}\right) + 2^p c^{-k} \sum_{m=m_0+1}^{\infty} \frac{1}{m^{\frac{k}{d}-p}} \end{aligned}$$

since $\frac{k}{d} - p > 1$, we obtain $E\left[\frac{1}{(\det I_t)^p}\right] < \infty$.

Let $t > 0$ be fixed.

As $(\det I_t)^{\frac{1}{d}} \geq \inf_{|V|=1} \langle I_t V, V \rangle$, we want to find a lower bound for $\inf_{|V|=1} \langle I_t V, V \rangle$.

Let ε be such that $0 < \varepsilon < \frac{t}{2}$. We consider $V = (V_i)_{1 \leq i \leq d} \in \mathbb{R}^d$ such that $|V| = 1$.

$$\begin{aligned} \langle I_t V, V \rangle &= \int_0^t \int_0^1 \sum_{l=1}^{d'} \left(\sum_{i=1}^d D_{r,z}^l X_{i,t} V_i \right)^2 dz dr \\ &\geq \int_{t-\varepsilon}^t \int_0^1 \sum_{l=1}^{d'} \left(\sum_{i=1}^d D_{r,z}^l X_{i,t} V_i \right)^2 dz dr \\ &\geq \frac{2}{3} I_1 - 2 I_2 \end{aligned}$$

with

$$\begin{aligned} I_1 &= \int_{t-\varepsilon}^t \int_0^1 \sum_{l=1}^{d'} \left(\sum_{i=1}^d \sigma_{i,l}(X_r - Y_r(z)) V_i \right)^2 dz dr \\ I_2 &= \int_{t-\varepsilon}^t \int_0^1 \sum_{l=1}^{d'} \left[\sum_{i=1}^d V_i \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k}(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} W_k(d\alpha, ds) \right. \\ &\quad \left. + \sum_{i=1}^d V_i \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \right]^2 dz dr \end{aligned}$$

Then

$$\inf_{|V|=1} \langle I_t V, V \rangle \geq \frac{2}{3} \inf_{|V|=1} I_1 - 2 \sup_{|V|=1} I_2$$

We want to minimize the first integral:

$$\begin{aligned}
I_1 &= \int_{t-\varepsilon}^t \int_0^1 \sum_{l=1}^{d'} \left(\sum_{i=1}^d \sigma_{i,l}(X_r - Y_r(z)) V_i \right)^2 dz dr \\
&= \int_{t-\varepsilon}^t \int_0^1 \sum_{i,j=1}^d \sum_{l=1}^{d'} \sigma_{i,l}(X_r - Y_r(z)) \sigma_{j,l}(X_r - Y_r(z)) V_i V_j dz dr \\
&= \int_{t-\varepsilon}^t \int_0^1 \sum_{i,j=1}^d a_{i,j}(X_r - Y_r(z)) V_i V_j dz dr \\
&= \int_{t-\varepsilon}^t V^* \cdot \Gamma_r \cdot V dr
\end{aligned}$$

Using the results of Section 3, we obtain

$$I_1 \geq m \int_{t-\varepsilon}^t \left(E[|X_r|^2] |V|^2 - \sum_{i=1}^d V_i^2 E[X_{i,r}^2] - \sum_{i \neq j} V_i V_j E[X_{i,r} X_{j,r}] \right) dr$$

We define the function

$$f(V, r) = E[|X_r|^2] |V|^2 - \sum_{i=1}^d V_i^2 E[X_{i,r}^2] - \sum_{i \neq j} V_i V_j E[X_{i,r} X_{j,r}]$$

We notice that f is a positive continuous function (see Remark 18) on the compact subset $D = \{V \in \mathbb{R}^d : |V| = 1\} \cup \{r : \frac{t}{2} \leq r \leq t\}$, then f reaches its minimum. So, if we denote by

$$\tilde{c} = \inf \left\{ f(V, r) : |V| = 1 \text{ and } \frac{t}{2} \leq r \leq t \right\}$$

we notice that \tilde{c} is independent of $\omega \in \Omega$, $\tilde{c} > 0$ and

$$I_1 \geq m \cdot \tilde{c} \cdot \varepsilon$$

Let us now study $E \left[\sup_{|V|=1} I_2^p \right]$ for $p \geq 1$.

$$\begin{aligned}
I_2 &= \int_{t-\varepsilon}^t \int_0^1 \sum_{l=1}^{d'} \left[\sum_{i=1}^d V_i \int_r^t \int_0^1 \sum_{k=1}^{d'} \sum_{m=1}^d \partial_m \sigma_{i,k}(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} W_k(d\alpha, ds) \right. \\
&\quad \left. + \sum_{i=1}^d V_i \int_r^t \int_0^1 \sum_{m=1}^d \partial_m b_i(X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \right]^2 dz dr
\end{aligned}$$

Using Burkholder-Davis-Gundy's and Hölder's inequalities, and the fact that $|V| = 1$, we notice that

$$\begin{aligned}
E \left[\sup_{|V|=1} I_2^p \right] &\leq K\varepsilon^{p-1} \left\{ \int_{t-\varepsilon}^t \int_0^1 \sum_{l,i,m} E \left[\left| \sum_{k=1}^{d'} \int_r^t \int_0^1 \partial_m \sigma_{i,k} (X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} W_k(d\alpha, ds) \right|^{2p} \right] dzdr \right. \\
&\quad \left. + \int_{t-\varepsilon}^t \int_0^1 \sum_{l,i,m} E \left[\left| \int_r^t \partial_m b_i (X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \right|^{2p} \right] dzdr \right\} \\
&\leq K\varepsilon^{p-1} \left\{ \int_{t-\varepsilon}^t \int_0^1 \sum_{l,i,k,m} E \left[\left| \sum_{k=1}^{d'} \int_r^t \int_0^1 (\partial_m \sigma_{i,k} (X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s})^2 d\alpha ds \right|^p \right] dzdr \right. \\
&\quad \left. + \int_{t-\varepsilon}^t \int_0^1 \varepsilon^{2p-1} \sum_{l,i,m} E \left[\left| \int_r^t \partial_m b_i (X_s - Y_s(\alpha)) D_{(r,z)}^l X_{m,s} d\alpha ds \right|^{2p} \right] dzdr \right\}
\end{aligned}$$

As the derivatives of σ and b are bounded, using Hölder's inequality, we obtain

$$\begin{aligned}
E \left[\sup_{|V|=1} I_2^p \right] &\leq K\varepsilon^{2p-2} \left\{ \int_{t-\varepsilon}^t \int_0^1 \sum_l E \left[\int_r^t |D_{(r,z)}^l X_s|^{2p} ds \right] dzdr \right\} \\
&\quad \text{using Fubini's Theorem} \\
&\leq K\varepsilon^{2p-2} \int_r^t E \left[\int_{t-\varepsilon}^s \int_0^1 |D_{(r,z)} X_s|^{2p} dzdr \right] ds
\end{aligned}$$

Then, for any $p \geq 2$ there exists a constant $K = K(p, d, d', t)$ such that

$$E \left[\sup_{|V|=1} I_2^p \right] \leq K\varepsilon^{2p-1} \sup_{0 \leq s \leq t} E \left[\int_0^s \int_0^1 |D_{(r,z)} X_s|^{2p} dzdr \right]$$

Let us now check that $\lim_{\varepsilon \rightarrow 0} \varepsilon^{-k} \mathbb{P} \left((\det I_t)^{\frac{1}{d}} < c\varepsilon \right) = 0$, where $k \in \mathbb{N}$ is fixed and $c = \frac{1}{3} m\tilde{c}$ with \tilde{c} the constant built in the study of the first integral I_1 .

Let $p \in \mathbb{N}$ such that $p > k + 1$.

$$\begin{aligned}
\mathbb{P} \left((\det I_t)^{\frac{1}{d}} < c\varepsilon \right) &\leq \mathbb{P} \left(\inf_{|V|=1} \langle I_t V, V \rangle < c\varepsilon \right) \\
&\leq \mathbb{P} \left(\frac{2}{3} \inf_{|V|=1} I_1 - 2 \sup_{|V|=1} I_2 < c\varepsilon \right) \\
&\leq \mathbb{P} \left(\sup_{|V|=1} I_2 > \frac{c\varepsilon}{2} \right) \\
&\quad \text{using Tchebychev's Inequality} \\
&\leq \left(\frac{2}{c} \right)^p \varepsilon^{-p} E \left[\sup_{|V|=1} I_2^p \right] \\
&\leq K\varepsilon^{p-1}
\end{aligned}$$

Then, $\lim_{\varepsilon \rightarrow 0} \varepsilon^{-k} \mathbb{P} \left((\det I_t)^{\frac{1}{d}} < c\varepsilon \right) = 0$ and we can apply Theorem (b). So the theorem is proved. ■

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