Integration by parts formula and applications to equations with jumps

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Abstract We establish an integration by parts formula in an abstract framework in order to study the regularity of the law for processes arising as the solution of stochastic differential equations with jumps, including equations with discontinuous coefficients for which the Malliavin calculus developed by Bichteler et al. (Stochastics Monographs, vol 2. Gordon & Breach, New York, 1987) and Bismut (Z Wahrsch Verw Gebiete 63(2):147–235, 1983) fails.

Keywords Integration by parts formula · Malliavin calculus · Stochastic equations · Poisson point measures

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1 Introduction

This paper is made up of two parts. In a first part we give an abstract, finite dimensional version of Malliavin calculus. Of course Malliavin calculus is known to be an infinite dimensional differential calculus (see [20] for the basic theory of Malliavin calculus) and so a finite dimensional version seems to be of a limited interest. We discuss later on the relation between the finite dimensional and the infinite dimensional frameworks, and we highlight the interest of the finite dimensional approach.

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In the second part of the paper, we use the results from the first section in order to give sufficient conditions for the regularity of the law of X_t , where X is the Markov process with infinitesimal operator

$$Lf(x) = \langle \nabla f(x), g(x) \rangle + \int_{R^d} (f(x + c(z, x)) - f(x))\gamma(z, x)\mu(dz).$$
(1)

Suppose for the moment that γ does not depend on x. Then it is well known that the process X may be represented as the solution of a stochastic equation driven by a Poisson point measure with intensity measure $\gamma(z)\mu(dz)$. Sufficient conditions for the regularity of the law of X_t using a Malliavin calculus for Poisson point measures are given in [6]. But in our framework γ depends on x, which roughly speaking means that the intensity of the jumps depends on the position of the particle when the jump occurs. Such processes are of interest in a lot of applications and unfortunately the standard Malliavin calculus developed in [6] does not apply in this framework. After the classical papers of Bismut [7] and Bichteler et al. [6] much work concerning the Malliavin calculus for Poisson point measures has been done and many different approaches have been developed: see for example [1,8,13,19,21]. But as far as we know they do not lead to a solution for our problem. If X is an one-dimensional process an analytical argument permits to solve the above problem, this is done in Fournier [9–11] but the argument there seems difficult to extend in the multi-dimensional case.

We come now back to the relation between the finite dimensional and the infinite dimensional framework. This seems to be the more interesting point in our approach so we try to explain the main idea. In order to prove Malliavin's regularity criterion for the law of a functional F on the Wiener space the main tool is the integration by parts formula

$$E(\partial^{\beta} f(F)) = E(f(F)H_{\beta})$$
⁽²⁾

where ∂^{β} denotes the derivative corresponding to a multi-index β and H_{β} is a random variable built using the Malliavin derivatives of *F*. Once such a formula is proved one may estimate the Fourier transform $\hat{p}_F(\xi) = E(\exp(i\xi F))$ in the following way. First we remark that $\partial_x^{\beta} \exp(i\xi x) = (i\xi)^{\beta} \exp(i\xi x)$ (with an obvious abuse of notation) and then, using the integration by parts formula

$$\begin{aligned} |\widehat{p}_F(\xi)| &= \frac{1}{|\xi|^{|\beta|}} \left| E(\partial_x^\beta \exp(i\xi F)) \right| \\ &= \frac{1}{|\xi|^{|\beta|}} \left| E(\exp(i\xi F)H_\beta) \right| \le \frac{1}{|\xi|^{|\beta|}} E\left| H_\beta \right|. \end{aligned}$$

If we know that $E |H_{\beta}| < \infty$ for every multi-index β then we have proved that $|\xi|^p |\hat{p}_F(\xi)|$ is integrable for every $p \in \mathbb{N}$ and consequently the law of *F* is absolutely continuous with respect to the Lebesgue measure and has an infinitely differentiable density.

Let us come back to the infinite dimensional differential calculus which allows the construction H_{β} . In order to define the Malliavin derivative of F one considers a sequence of simple functionals $F_n \to F$ in L^2 and, if $DF_n \to G$ in L^2 , then one defines DF = G. The simple functionals F_n are functions of a finite number of random variables (increments of the Brownian motion) and the derivative DF_n is a gradient type operator defined in an elementary way. Then one may take the following alternative route in order to prove the regularity of the law of F. For each fixed n one proves an analogue of the integration by parts formula (2): $E(\partial^{\beta} f(F_n)) = E(f(F_n)H_{\beta}^n)$. As F_n is a function which depends on a finite number m of random variables, such a formula is obtained using standard integration by parts on \mathbb{R}^m (this is done in the first section of this paper). Then the same calculus as above gives $|\widehat{p}_{F_n}(\xi)| \leq |\xi|^{-|\beta|} E |H_{\beta}^n|$. Passing to the limit one obtains

$$\left|\widehat{p}_{F}(\xi)\right| = \lim_{n} \left|\widehat{p}_{F_{n}}(\xi)\right| \le \left|\xi\right|^{-\left|\beta\right|} \sup_{n} E\left|H_{\beta}^{n}\right|$$

and, if we can prove that $\sup_n E |H_{\beta}^n| < \infty$, we are done. Notice that here we do not need that $F_n \to F$ in L^2 but only in law. And also, we do not need to construct H_{β} but only to prove that $\sup_n E |H_{\beta}^n| < \infty$. Anyway we are not very far from the standard Malliavin calculus. Things become different if $\sup_n E |H_{\beta}^n| = \infty$ and this is the case in our examples (because the Ornstein Uhlenbeck operators LF_n blow up as $n \to \infty$). But even in this case one may obtain estimates of the Fourier transform of F in the following way. One writes

$$\left|\widehat{p}_{F}(\xi)\right| \leq \left|\widehat{p}_{F}(\xi) - \widehat{p}_{F_{n}}(\xi)\right| + \left|\widehat{p}_{F_{n}}(\xi)\right| \leq \left|\xi\right| \times E\left|F - F_{n}\right| + \left|\xi\right|^{-\left|\beta\right|} E\left|H_{\beta}^{n}\right|.$$

And if one may obtain a good balance between the convergence to zero of the error $E |F - F_n|$ and the blow up to infinity of $E \left| H_{\beta}^n \right|$ then one obtains $|\hat{p}_F(\xi)| \le |\xi|^{-p}$ for some p. Examples in which such a balance works are given in Sect. 3. An other application of this methodology is given in Bally and Fournier [3] for the Boltzmann equation. In this case some specific and nontrivial difficulties appear due to the singularity and unboundedness of the coefficients of the equation.

The paper is organized as follows. In Sect. 2, we establish the abstract Malliavin calculus associated to a finite dimensional random variable and we obtain estimates of the weight H_{β} which appears in the integration by parts formula (we follow here some ideas which already appear in [2,5]). Section 3 is devoted to the study of the regularity of the law of the Markov process X of infinitesimal operator (1) and it contains our main results: Proposition 3 and Theorem 4. Finally, we provide in Sect. 4 the technical estimates which are needed to prove the results of Sect. 3.

2 Integration by parts formula

2.1 Notations-derivative operators

Throughout this paper, we consider a sequence of random variables $(V_i)_{i \in \mathbb{N}^*}$ on a probability space (Ω, \mathcal{F}, P) , a sub σ -algebra $\mathcal{G} \subseteq \mathcal{F}$ and a random variable J, \mathcal{G} measurable, with values in \mathbb{N} . We assume that the variables (V_i) and J satisfy the following integrability conditions: for all $p \ge 1$, $E(J^p) + E((\sum_{i=1}^J V_i^2)^p) < \infty$. Our aim is to establish a differential calculus based on the variables (V_i) , conditionally on \mathcal{G} , and we first define the class of functions on which this differential calculus will apply. More precisely, we consider in this paper functions $f : \Omega \times \mathbb{R}^{\mathbb{N}^*} \to \mathbb{R}$ which can be written as

$$f(\omega, v) = \sum_{j=1}^{\infty} f^{j}(\omega, v_{1}, \dots, v_{j}) \mathbf{1}_{\{J(\omega)=j\}}$$
(3)

where $f^j : \Omega \times \mathbb{R}^j \to \mathbb{R}$ are $\mathcal{G} \times \mathcal{B}(\mathbb{R}^j)$ -measurable functions. We denote by \mathcal{M} the class of functions f given by (3) such that there exists a random variable $C \in \bigcap_{q \ge 1} L^q(\Omega, \mathcal{G}, P)$ and a real number $p \ge 1$ satisfying $|f(\omega, v)| \le C(\omega)(1 + (\sum_{i=1}^{J(\omega)} v_i^2)^p)$. So conditionally on \mathcal{G} , the functions of \mathcal{M} have polynomial growth with respect to the variables (V_i) . We need some more notations. Let \mathcal{G}_i be the σ -algebra generated by $\mathcal{G} \cup \sigma(V_j, 1 \le j \le J, j \ne i)$ and let $(a_i(\omega))$ and $(b_i(\omega))$ be sequences of \mathcal{G}_i measurable random variables satisfying $-\infty \le a_i(\omega) < b_i(\omega) \le +\infty$, for all $i \in \mathbb{N}^*$. Now let O_i be the open set of $\mathbb{R}^{\mathbb{N}^*}$ defined by $O_i = P_i^{-1}(]a_i, b_i[]$, where P_i is the coordinate map $P_i(v) = v_i$. We localize the differential calculus on the sets (O_i) by introducing some weights (π_i) , satisfying the following hypothesis.

H0. For all $i \in \mathbb{N}^*$, $\pi_i \in \mathcal{M}$, $0 \le \pi_i \le 1$ and $\{\pi_i > 0\} \subset O_i$. Moreover for all $j \ge 1, \pi_i^j$ is infinitely differentiable with bounded derivatives with respect to the variables (v_1, \ldots, v_j) .

We associate to these weights (π_i) , the spaces $C_{\pi}^k \subset \mathcal{M}, k \in \mathbb{N}^*$ defined recursively as follows. For $k = 1, C_{\pi}^1$ denotes the space of functions $f \in \mathcal{M}$ such that for each $i \in \mathbb{N}^*$, f admits a partial derivative with respect to the variable v_i on the open set O_i . We then define

$$\partial_i^{\pi} f(\omega, v) := \pi_i(\omega, v) \frac{\partial}{\partial v_i} f(\omega, v)$$

and we assume that $\partial_i^{\pi} f \in \mathcal{M}$.

Note that the chain rule is verified : for each $\phi \in C^1(\mathbb{R}^d, \mathbb{R})$ and $f = (f^1, \dots, f^d) \in (C^1_{\pi})^d$ we have

$$\partial_i^{\pi} \phi(f) = \sum_{r=1}^d \partial_r \phi(f) \partial_i^{\pi} f^r.$$

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Suppose now that C_{π}^{k} is already defined. For a multi-index $\alpha = (\alpha_{1}, \ldots, \alpha_{k}) \in \mathbb{N}^{*k}$ we define recursively $\partial_{\alpha}^{\pi} = \partial_{\alpha_{k}}^{\pi} \ldots \partial_{\alpha_{1}}^{\pi}$ and C_{π}^{k+1} is the space of functions $f \in C_{\pi}^{k}$ such that for every multi-index $\alpha = (\alpha_{1}, \ldots, \alpha_{k}) \in \mathbb{N}^{*k}$ we have $\partial_{\alpha}^{\pi} f \in C_{\pi}^{1}$. Note that if $f \in C_{\pi}^{k}, \partial_{\alpha}^{\pi} f \in \mathcal{M}$ for each α with $|\alpha| \leq k$.

Finally we define $C_{\pi}^{\infty} = \bigcap_{k \in \mathbb{N}^*} C_{\pi}^k$. Roughly speaking the space C_{π}^{∞} is the analogue of C^{∞} with partial derivatives ∂_i replaced by localized derivatives ∂_i^{π} .

Simple functionals. A random variable *F* is called a simple functional if there exists $f \in C_{\pi}^{\infty}$ such that $F = f(\omega, V)$, where $V = (V_i)$. We denote by *S* the space of the simple functionals. Notice that *S* is an algebra. It is worth to remark that conditionally on \mathcal{G} , $F = f^J(V_1, \ldots, V_J)$.

Simple processes. A simple process is a sequence of random variables $U = (U_i)_{i \in \mathbb{N}^*}$ such that for each $i \in \mathbb{N}^*$, $U_i \in S$. Consequently, conditionally on \mathcal{G} , we have $U_i = u_i^J(V_1, \ldots, V_J)$. We denote by \mathcal{P} the space of the simple processes and we define the scalar product

$$\langle U, V \rangle_J = \sum_{i=1}^J U_i V_i.$$

Note that $\langle U, V \rangle_J \in \mathcal{S}$.

We can now define the derivative operator and state the integration by parts formula. \Box The derivative operator. We define $D : S \rightarrow P$: by

$$DF := (D_i F) \in \mathcal{P}$$
 where $D_i F := \partial_i^{\pi} f(\omega, V)$.

Note that $D_i F = 0$ for i > J. For $F = (F^1, ..., F^d) \in S^d$ the Malliavin covariance matrix is defined by

$$\sigma^{k,k'}(F) = \left\langle DF^k, DF^{k'} \right\rangle_J = \sum_{j=1}^J D_j F^k D_j F^{k'}.$$

We denote

$$\Lambda(F) = \{\det \sigma(F) \neq 0\} \text{ and } \gamma(F)(\omega) = \sigma^{-1}(F)(\omega), \omega \in \Lambda(F).$$

In order to derive an integration by parts formula, we need some additional assumptions on the random variables (V_i) . The main hypothesis is that conditionally on \mathcal{G} , the law of the vector (V_1, \ldots, V_J) admits a locally smooth density with respect to the Lebesgue measure on \mathbb{R}^J .

H1.

- (i) Conditionally on \mathcal{G} , the vector (V_1, \ldots, V_J) is absolutely continuous with respect to the Lebesgue measure on \mathbb{R}^J and we note p_J the conditional density.
- (ii) The set $\{p_J > 0\}$ is open in \mathbb{R}^J and on $\{p_J > 0\}$ ln $p_J \in C_{\pi}^{\infty}$.

(iii) $\forall q \ge 1$, there exists a constant C_q such that

$$(1+|v|^q)p_J \le C_q$$

where |v| stands for the euclidian norm of the vector (v_1, \ldots, v_J) .

Assumption iii) implies in particular that conditionally on \mathcal{G} , the functions of \mathcal{M} are integrable with respect to p_J and that for $f \in \mathcal{M}$:

$$E_{\mathcal{G}}(f(\omega, V)) = \int_{\mathbb{R}^J} f^J \times p_J(\omega, v_1, \dots, v_J) dv_1 \dots dv_J$$

 \Box The divergence operator Let $U = (U_i)_{i \in \mathbb{N}^*} \in \mathcal{P}$ with $U_i \in \mathcal{S}$. We define $\delta : \mathcal{P} \to \mathcal{S}$ by

$$\delta_i(U) := -(\partial_{v_i}(\pi_i U_i) + U_i \mathbf{1}_{\{p_J > 0\}} \partial_i^{\pi} \ln p_J),$$

$$\delta(U) = \sum_{i=1}^J \delta_i(U)$$

For $F \in S$, let $L(F) = \delta(DF)$.

2.2 Duality and integration by parts formulae

In our framework the duality between δ and D is given by the following proposition.

Proposition 1 Assume H0 and H1, then $\forall F \in S$ and $\forall U \in P$ we have

$$E_{\mathcal{G}}(\langle DF, U \rangle_J) = E_{\mathcal{G}}(F\delta(U)). \tag{4}$$

Proof By definition, we have $E_{\mathcal{G}}(\langle DF, U \rangle_J) = \sum_{i=1}^J E_{\mathcal{G}}(D_i F \times U_i)$ and from **H1**

$$E_{\mathcal{G}}(D_i F \times U_i) = \int_{\mathbb{R}^J} \partial_{v_i}(f^J) \pi_i \ u_i^J \ p_J(\omega, v_1, \dots, v_J) dv_1 \dots dv_J$$

recalling that $\{\pi_i > 0\} \subset O_i$, we obtain from Fubini's theorem

$$E_{\mathcal{G}}(D_i F \times U_i) = \int_{\mathbb{R}^{J-1}} \left(\int_{a_i}^{b_i} \partial_{v_i} (f^J) \pi_i \, u_i^J \, p_J(\omega, v_1, \dots, v_J) dv_i \right)$$
$$\times dv_1 \dots dv_{i-1} dv_{i+1} \dots dv_J.$$

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By using the classical integration by parts formula, we have

$$\int_{a_{i}}^{b_{i}} \partial_{v_{i}}(f^{J})\pi_{i} u_{i}^{J} p_{J}(\omega, v_{1}, \dots, v_{J}) dv_{i} = [f^{J}\pi_{i}u_{i}^{J}p_{J}]_{a_{i}}^{b_{i}} - \int_{a_{i}}^{b_{i}} f^{J} \partial_{v_{i}}(u_{i}^{J}\pi_{i}p_{J}) dv_{i}.$$

Now if $-\infty < a_i < b_i < +\infty$, we have $\pi_i(a_i) = 0 = \pi_i(b_i)$ and $[f^J \pi_i u_i^J p_J]_{a_i}^{b_i} = 0$. Moreover since f^J , u_i^J and π_i belong to \mathcal{M} , we deduce from H_1 iii) that $\lim_{|v_i| \to +\infty} (f^J \pi_i u_i^J p_J) = 0$ and we obtain that for all a_i , b_i such that $-\infty \le a_i < b_i \le +\infty$:

$$\int_{a_i}^{b_i} \partial_{v_i}(f^J)\pi_i u_i^J p_J(\omega, v_1, \dots, v_J)dv_i = -\int_{a_i}^{b_i} f^J \partial_{v_i}(u_i^J \pi_i p_J)dv_i,$$

Observing that $\partial_{v_i}(u_i^J \pi_i p_J) = (\partial_{v_i}(u_i^J \pi_i) + u_i^J \mathbf{1}_{\{p_J > 0\}} \partial_i^{\pi}(\ln p_J)) p_J$, the proposition is proved.

We have the following straightforward computation rules.

Lemma 1 Let $\phi : \mathbb{R}^d \to \mathbb{R}$ be a smooth function and $F = (F^1, \dots, F^d) \in S^d$. Then $\phi(F) \in S$ and

$$D\phi(F) = \sum_{r=1}^{d} \partial_r \phi(F) DF^r.$$
 (5)

If $F \in S$ and $U \in P$ then

$$\delta(FU) = F\delta(U) - \langle DF, U \rangle_J .$$
(6)

Moreover, for $F = (F^1, \ldots, F^d) \in S^d$, we have

$$L\phi(F) = \sum_{r=1}^{d} \partial_r \phi(F) LF^r - \sum_{r,r'=1}^{d} \partial_{r,r'} \phi(F) \left\langle DF^r, DF^{r'} \right\rangle_J.$$
(7)

The first equality is a consequence of the chain rule, the second one follows from the definition of the divergence operator δ . Combining these equalities (7) follows.

We can now state the main results of this section.

Theorem 1 We assume H0 and H1. Let $F = (F^1, \ldots, F^d) \in S^d$, $G \in S$ and $\phi : \mathbb{R}^d \to \mathbb{R}$ be a smooth bounded function with bounded derivatives. Let $\Lambda \in \mathcal{G}$, $\Lambda \subset \Lambda(F)$ such that

$$E(|\det \gamma(F)|^p \mathbf{1}_{\Lambda}) < \infty \quad \forall p \ge 1.$$
(8)

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Then, for every $r = 1, \ldots, d$,

$$E_{\mathcal{G}}\left(\partial_r \phi(F)G\right) \mathbf{1}_{\Lambda} = E_{\mathcal{G}}\left(\phi(F)H_r(F,G)\right) \mathbf{1}_{\Lambda} \tag{9}$$

with

$$H_{r}(F,G) = \sum_{r'=1}^{d} \delta(G\gamma^{r',r}(F)DF^{r'})$$

= $\sum_{r'=1}^{d} \left(G\delta(\gamma^{r',r}(F)DF^{r'}) - \gamma^{r',r}\left\langle DF^{r'}, DG \right\rangle_{J}\right).$ (10)

Proof Using the chain rule

$$\left\langle D\phi(F), DF^{r'} \right\rangle_J = \sum_{j=1}^J D_j \phi(F) D_j F^{r'}$$

=
$$\sum_{j=1}^J \left(\sum_{r=1}^d \partial_r \phi(F) D_j F^r \right) D_j F^{r'} = \sum_{r=1}^d \partial_r \phi(F) \sigma^{r,r'}(F)$$

so that $\partial_r \phi(F) \mathbf{1}_{\Lambda} = \mathbf{1}_{\Lambda} \sum_{r'=1}^d \left\langle D\phi(F), DF^{r'} \right\rangle_J \gamma^{r',r}(F)$. Since $F \in S^d$ it follows that $\phi(F) \in S$ and $\sigma^{r,r'}(F) \in S$. Moreover, since det $\gamma(F) \mathbf{1}_{\Lambda} \in \bigcap_{p \ge 1} L^p$ it follows that $\gamma^{r,r'}(F) \mathbf{1}_{\Lambda} \in S$. So $G\gamma^{r',r}(F)DF^{r'} \mathbf{1}_{\Lambda} \in \mathcal{P}$ and the duality formula gives:

$$E_{\mathcal{G}}(\partial_{r}\phi(F)G)\mathbf{1}_{\Lambda} = \sum_{r'=1}^{d} E_{\mathcal{G}}\left(\left\langle D\phi(F), G\gamma^{r',r}(F)DF^{r'}\right\rangle_{J}\right)\mathbf{1}_{\Lambda}$$
$$= \sum_{r'=1}^{d} E_{\mathcal{G}}\left(\phi(F)\delta(G\gamma^{r',r}(F)DF^{r'})\right)\mathbf{1}_{\Lambda}.$$

We can extend this integration by parts formula.

Theorem 2 Under the assumptions of Theorem 1, we have for every multi-index $\beta = (\beta_1, \ldots, \beta_q) \in \{1, \ldots, d\}^q$

$$E_{\mathcal{G}}\left(\partial_{\beta}\phi(F)G\right)\mathbf{1}_{\Lambda} = E_{\mathcal{G}}\left(\phi(F)H_{\beta}^{q}(F,G)\right)\mathbf{1}_{\Lambda}$$
(11)

where the weights H^q are defined recursively by (10) and

$$H^{q}_{\beta}(F,G) = H_{\beta_{1}}\left(F, H^{q-1}_{(\beta_{2},\dots,\beta_{q})}(F,G)\right).$$
(12)

Proof The proof is straightforward by induction. For q = 1, this is just Theorem 1. Now assume that Theorem 2 is true for $q \ge 1$ and let us prove it for q + 1. Let $\beta = (\beta_1, \ldots, \beta_{q+1}) \in \{1, \ldots, d\}^{q+1}$, we have

$$E_{\mathcal{G}}\left(\partial_{\beta}\phi(F)G\right)\mathbf{1}_{\Lambda} = E_{\mathcal{G}}\left(\partial_{(\beta_{2},\dots,\beta_{q+1})}(\partial_{\beta_{1}}\phi(F))G\right)\mathbf{1}_{\Lambda}$$
$$= E_{\mathcal{G}}\left(\partial_{\beta_{1}}\phi(F)H^{q}_{(\beta_{2},\dots,\beta_{q+1})}(F,G)\right)\mathbf{1}_{\Lambda}$$

and the result follows.

2.3 Estimations of H^q

2.3.1 Iterated derivative operators, Sobolev norms

In order to estimate the weights H^q appearing in the integration by parts formulae of the previous section, we need first to define iterations of the derivative operator. Let $\alpha = (\alpha_1, \ldots, \alpha_k)$ be a multi-index, with $\alpha_i \in \{1, \ldots, J\}$, for $i = 1, \ldots, k$ and $|\alpha| = k$. For $F \in S$, we define recursively $D_{(\alpha_1, \ldots, \alpha_k)}^k F = D_{\alpha_k}(D_{(\alpha_1, \ldots, \alpha_{k-1})}^{k-1}F)$ and

$$D^{k}F = \left(D_{(\alpha_{1},...,\alpha_{k})}^{k}F\right)_{\alpha_{i}\in\{1,...,J\}}$$

Remark that $D^k F \in \mathbb{R}^{J \otimes k}$ and consequently we define the norm of $D^k F$ as

$$|D^k F| = \sqrt{\sum_{\alpha_1,\dots,\alpha_k=1}^J |D^k_{(\alpha_1,\dots,\alpha_k)}F|^2}.$$

Moreover, we introduce the following norms, for $F \in S$:

$$|F|_{1,l} = \sum_{k=1}^{l} |D^{k}F|, \quad |F|_{l} = |F| + |F|_{1,l} = \sum_{k=0}^{l} |D^{k}F|.$$
(13)

For $F = (F^1, \ldots, F^d) \in \mathcal{S}^d$:

$$|F|_{1,l} = \sum_{r=1}^{d} |F^r|_{1,l}, \quad |F|_l = \sum_{r=1}^{d} |F^r|_l,$$

and similarly for $F = (F^{r,r'})_{r,r'=1,...,d}$

$$|F|_{1,l} = \sum_{r,r'=1}^{d} |F^{r,r'}|_{1,l}, \quad |F|_l = \sum_{r,r'=1}^{d} |F^{r,r'}|_l.$$

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Finally for $U = (U_i)_{i \leq J} \in \mathcal{P}$, we have $D^k U = (D^k U_i)_{i \leq J}$ and we define the norm of $D^k U$ as

$$|D^k U| = \sqrt{\sum_{i=1}^J |D^k U_i|^2}.$$

We can remark that for k = 0, this gives $|U| = \sqrt{\langle U, U \rangle_J}$. Similarly to (13), we set

$$|U|_{1,l} = \sum_{k=1}^{l} |D^{k}U|, \quad |U|_{l} = |U| + |U|_{1,l} = \sum_{k=0}^{l} |D^{k}U|.$$

Observe that for $F, G \in S$, we have $D(F \times G) = DF \times G + F \times DG$. This leads to the following useful inequalities

Lemma 2 Let $F, G \in S$ and $U, V \in P$, we have

$$|F \times G|_{l} \le 2^{l} \sum_{l_{1}+l_{2} < l} |F|_{l_{1}} |G|_{l_{2}},$$
(14)

$$|\langle U, V \rangle_J |_l \le 2^l \sum_{l_1 + l_2 \le l} |U|_{l_1} |V|_{l_2}.$$
(15)

We can remark that the first inequality is sharper than the following one $|F \times G|_l \le C_l |F|_l |G|_l$. Moreover from (15) with U = DF and V = DG ($F, G, \in S$) we deduce

$$|\langle DF, DG \rangle_J|_l \le 2^l \sum_{l_1+l_2 \le l} |F|_{1,l_1+1} |G|_{1,l_2+1}$$
(16)

and as an immediate consequence of (14) and (16), we have for $F, G, H \in S$:

$$|H \langle DF, DG \rangle_J|_l \le 2^{2l} \sum_{l_1+l_2+l_3 \le l} |F|_{1,l_1+1} |G|_{1,l_2+1} |H|_{l_3}.$$
 (17)

Proof We just prove (15), since (14) can be proved on the same way. We first give a bound for $D^k \langle U, V \rangle_J = (D^k_{\alpha} \langle U, V \rangle_J)_{\alpha \in \{1, ..., J\}^k}$. For a multi-index $\alpha = (\alpha_1, ..., \alpha_k)$, with $\alpha_i \in \{1, ..., J\}$, we note $\alpha(\Gamma) = (\alpha_i)_{i \in \Gamma}$, where $\Gamma \subset \{1, ..., k\}$ and $\alpha(\Gamma^c) = (\alpha_i)_{i \notin \Gamma}$. We have

$$D_{\alpha}^{k} \langle U, V \rangle_{J} = \sum_{i=1}^{J} D_{\alpha}^{k} (U_{i} V_{i}) = \sum_{k'=0}^{k} \sum_{|\Gamma|=k'} \sum_{i=1}^{J} D_{\alpha(\Gamma)}^{k'} U_{i} \times D_{\alpha(\Gamma^{c})}^{k-k'} V_{i}.$$

Let $W^{i,\Gamma} = (W^{i,\Gamma}_{\alpha})_{\alpha \in \{1,...,J\}^k} = (D^{k'}_{\alpha(\Gamma)}U_i \times D^{k-k'}_{\alpha(\Gamma^c)}V_i)_{\alpha \in \{1,...,J\}^k}$, we have the equality in $\mathbb{R}^{J \otimes k}$:

$$D^k \langle U, V \rangle_J = \sum_{k'=0}^k \sum_{|\Gamma|=k'} \sum_{i=1}^J W^{i,\Gamma}.$$

This gives

$$|D^{k} \langle U, V \rangle_{J}| \leq \sum_{k'=0}^{k} \sum_{|\Gamma|=k'} \left| \sum_{i=1}^{J} W^{i,\Gamma} \right|,$$

where

$$\left|\sum_{i=1}^{J} W^{i,\Gamma}\right| = \sqrt{\sum_{\alpha_1,\dots,\alpha_k=1}^{J} \left|\sum_{i=1}^{J} W^{i,\Gamma}_{\alpha}\right|^2}.$$

But from Cauchy-Schwarz inequality, we have

$$\left|\sum_{i=1}^{J} W_{\alpha}^{i,\Gamma}\right|^{2} = \left|\sum_{i=1}^{J} D_{\alpha(\Gamma)}^{k'} U_{i} \times D_{\alpha(\Gamma^{c})}^{k-k'} V_{i}\right|^{2} \le \sum_{i=1}^{J} |D_{\alpha(\Gamma)}^{k'} U_{i}|^{2} \times \sum_{i=1}^{J} |D_{\alpha(\Gamma^{c})}^{k-k'} V_{i}|^{2}.$$

Consequently we obtain

$$\left|\sum_{i=1}^{J} W^{i,\Gamma}\right| \leq \sqrt{\sum_{\alpha_1,\dots,\alpha_k=1}^{J} \sum_{i=1}^{J} |D_{\alpha(\Gamma)}^{k'} U_i|^2 \times \sum_{i=1}^{J} |D_{\alpha(\Gamma^c)}^{k-k'} V_i|^2}$$
$$= |D^{k'} U| \times |D^{k-k'} V|.$$

This last equality results from the fact that we sum on different index sets (Γ and Γ^c). This gives

$$\begin{split} \left| D^{k} \langle U, V \rangle_{J} \right| &\leq \sum_{k'=0}^{k} \sum_{|\Gamma|=k'} \left| D^{k'} U \right| \left| D^{k-k'} V \right| = \sum_{k'=0}^{k} C_{k}^{k'} \left| D^{k'} U \right| \left| D^{k-k'} V \right| \\ &\leq \sum_{k'=0}^{k} C_{k}^{k'} \left| U \right|_{k'} \left| V \right|_{k-k'} \leq 2^{k} \left(\sum_{l_{1}+l_{2}=k} \left| U \right|_{l_{1}} \left| V \right|_{l_{2}} \right). \end{split}$$

Summing on k = 0, ..., l we deduce (16).

2.3.2 Estimation of $|\gamma(F)|_l$

We give in this section an estimation of the derivatives of $\gamma(F)$ in terms of det $\sigma(F)$ and the derivatives of *F*. We assume that $\omega \in \Lambda(F)$.

In what follows $C_{l,d}$ is a constant depending eventually on the order of derivation l and the dimension d.

Proposition 2 Let $F \in S^d$, we have $\forall l \in \mathbb{N}$

$$|\gamma(F)|_{l} \leq C_{l,d} \sum_{l_{1}+l_{2} \leq l} |F|_{1,l_{2}+1}^{2(d-1)} \left(\frac{1}{|\det \sigma(F)|} + \sum_{k=1}^{l_{1}} \frac{|F|_{1,l_{1}+1}^{2kd}}{|\det \sigma(F)|^{k+1}} \right)$$
(18)

$$\leq C_{l,d} \frac{1}{|\det \sigma(F)|^{l+1}} (1 + |F|_{1,l+1}^{2d(l+1)}).$$
⁽¹⁹⁾

Before proving Proposition 2, we establish a preliminary lemma.

Lemma 3 for every $G \in S$, G > 0 we have

$$\left|\frac{1}{G}\right|_{l} \leq C_{l}\left(\frac{1}{G} + \sum_{k=1}^{l} \frac{1}{G^{k+1}} \sum_{\substack{k \leq r_{1} + \dots + r_{k} \leq l \ i=1}} \prod_{i=1}^{k} |D^{r_{i}}G|\right)$$
$$\leq C_{l}\left(\frac{1}{G} + \sum_{k=1}^{l} \frac{1}{G^{k+1}} |G|_{1,l}^{k}\right).$$
(20)

Proof For $F \in S^d$ and $\phi : \mathbb{R}^d \to \mathbb{R}$ a \mathcal{C}^{∞} function, we have from the chain rule

$$D_{(\alpha_1,\dots,\alpha_k)}^k \phi(F) = \sum_{|\beta|=1}^k \partial_\beta \phi(F) \sum_{\Gamma_1 \cup \dots \cup \Gamma_{|\beta|} = \{1,\dots,k\}} \left(\prod_{i=1}^{|\beta|} D_{\alpha(\Gamma_i)}^{|\Gamma_i|} F^{\beta_i} \right), \quad (21)$$

where $\beta \in \{1, \ldots, d\}^{|\beta|}$ and $\sum_{\Gamma_1 \cup \ldots \cup \Gamma_{|\beta|}}$ denotes the sum over all partitions of $\{1, \ldots, k\}$ with length $|\beta|$. In particular, for $G \in S$, G > 0 and for $\phi(x) = 1/x$, we obtain

$$\left| D_{\alpha}^{k} \left(\frac{1}{G} \right) \right| \leq C_{k} \sum_{k'=1}^{k} \frac{1}{G^{k'+1}} \sum_{\Gamma_{1} \cup \dots \cup \Gamma_{k'} = \{1,\dots,k\}} \left(\prod_{i=1}^{k'} |D_{\alpha(\Gamma_{i})}^{|\Gamma_{i}|} G| \right).$$
(22)

We deduce then that

$$\begin{split} D^{k}\left(\frac{1}{G}\right) & \bigg| \leq C_{k} \sum_{k'=1}^{k} \frac{1}{G^{k'+1}} \sum_{\Gamma_{1} \cup \ldots \cup \Gamma_{k'} = \{1, \ldots, k\}} \left| \prod_{i=1}^{k'} D_{\alpha(\Gamma_{i})}^{|\Gamma_{i}|} G \right|_{\mathbb{R}^{J \otimes k}}, \\ & = C_{k} \sum_{k'=1}^{k} \frac{1}{G^{k'+1}} \sum_{\Gamma_{1} \cup \ldots \cup \Gamma_{k'} = \{1, \ldots, k\}} \left(\prod_{i=1}^{k'} |D^{|\Gamma_{i}|}G| \right), \\ & = C_{k} \sum_{k'=1}^{k} \frac{1}{G^{k'+1}} \sum_{\substack{r_{1} + \cdots + r_{k'} = k \\ r_{1}, \ldots, r_{k'} \geq 1}} \left(\prod_{i=1}^{k'} |D^{r_{i}}G| \right), \end{split}$$

and the first part of (20) is proved. The proof of the second part is straightforward.

With this lemma, we can prove Proposition 2.

Proof **Proposition 2.** We have on $\Lambda(F)$

$$\gamma^{r,r'}(F) = \frac{1}{\det \sigma(F)} \hat{\sigma}^{r,r'}(F),$$

where $\hat{\sigma}(F)$ is the algebraic complement of $\sigma(F)$. But recalling that $\sigma^{r,r'}(F) = \langle D^r F, D^{r'} F \rangle_I$ we have

$$|\det \sigma(F)|_l \le C_{l,d}|F|_{1,l+1}^{2d}$$
 and $|\hat{\sigma}(F)|_l \le C_{l,d}|F|_{1,l+1}^{2(d-1)}$. (23)

Applying inequality (14), this gives

$$|\gamma(F)|_l \le C_{l,d} \sum_{l_1+l_2 \le l} |(\det \sigma(F))^{-1}|_{l_1} |\hat{\sigma}(F)|_{l_2}.$$

From Lemma 3 and (23), we have

$$|(\det \sigma(F))^{-1}|_{l_1} \le C_{l_1} \left(\frac{1}{|\det \sigma(F)|} + \sum_{k=1}^{l_1} \frac{|F|_{1,l_1+1}^{2kd}}{|\det \sigma(F)|^{k+1}} \right).$$

Putting together these inequalities, we obtain the inequality (18) and consequently (19).

2.3.3 Some bounds on H^q

Now our goal is to establish some estimates for the weights H^q in terms of the derivatives of G, F, LF and $\gamma(F)$. **Theorem 3** For $F \in S^d$, $G \in S$ and for all $q \in \mathbb{N}^*$ there exists an universal constant $C_{q,d}$ such that for every multi-index $\beta = (\beta_1, \ldots, \beta_q)$

$$\begin{split} \left| H_{\beta}^{q}(F,G) \right| &\leq \frac{C_{q,d} \left| G \right|_{q} \left(1 + |F|_{q+1} \right)^{(6d+1)q}}{\left| \det \sigma(F) \right|^{3q-1}} \left(1 + \sum_{j=1}^{q} \sum_{k_{1} + \dots + k_{j} \leq q-j} \prod_{i=1}^{j} |L(F)|_{k_{i}} \right), \\ &\leq \frac{C_{q,d} \left| G \right|_{q} \left(1 + |F|_{q+1} \right)^{(6d+1)q}}{\left| \det \sigma(F) \right|^{3q-1}} \left(1 + |LF|_{q-1}^{q} \right). \end{split}$$

Proof For $F \in S^d$, we define the linear operator $T_r : S \to S, r = 1, ..., d$ by

$$T_r(G) = \left\langle DG, \left(\gamma(F)DF\right)^r \right\rangle,$$

where $(\gamma(F)DF)^r = \sum_{r'=1}^d \gamma^{r',r}(F)DF^{r'}$. Notice that

$$T_r(G \times G') = GT_r(G') + G'T_r(G).$$
⁽²⁴⁾

Moreover, for a multi-index $\beta = (\beta_1, ..., \beta_q)$ we define by induction $T_{\beta}(G) = T_{\beta_q}$ $(T_{(\beta_1,...,\beta_{q-1})}(G))$. We also make the convention that if β is the void multi-index, then $T_{\beta}(G) = G$. Finally we denote by $L_r^{\gamma'}(F) = \sum_{r'=1}^d \delta(\gamma^{r',r}(F)DF^{r'})$. With this notation we have

$$H_r(F, G) = GL_r^{\gamma}(F) - T_r(G), H_{\beta}^q(F, G) = H_{\beta_1}(F, H_{(\beta_2, \dots, \beta_q)}^{q-1}(F, G)).$$

We will now give an explicite expression of $H_{\beta}^{q}(F, G)$. In order to do this we have to introduce some more notation. Let $\Lambda_{j} = \{\lambda_{1}, \ldots, \lambda_{j}\} \subset \{1, \ldots, q\}$ such that $|\Lambda_{j}| = j$. We denote by $\mathcal{P}(\Lambda_{j})$ the set of the partitions $\Gamma = (\Gamma_{0}, \Gamma_{1}, \ldots, \Gamma_{j})$ of $\{1, \ldots, q\} \setminus \Lambda_{j}$. Notice that we accept that $\Gamma_{i}, i = 0, 1, \ldots, j$ may be void sets. Moreover, for a multi-index $\beta = (\beta_{1}, \ldots, \beta_{q})$ we denote by $\Gamma_{i}(\beta) = (\beta_{k_{i}^{1}}, \ldots, \beta_{k_{i}^{p}})$ where $\Gamma_{i} = \{k_{i}^{1}, \ldots, k_{i}^{p}\}$. With this notation we can prove by induction and using (24) that

$$H^{q}_{\beta}(F,G) = T_{\beta}(G) + \sum_{j=1}^{q} \sum_{\Lambda_{j} \subset \{1,\dots,q\}} \sum_{\Gamma \in \mathcal{P}(\Lambda_{j})} c_{\beta,\Gamma} T_{\Gamma_{0}(\beta)}(G) \prod_{i=1}^{j} T_{\Gamma_{i}(\beta)}(L^{\gamma}_{\beta_{\lambda_{i}}}(F))$$

$$(25)$$

where $c_{\beta,\Gamma} \in \{-1, 0, 1\}$.

We first give an estimation of $|T_{\beta}(G)|_l$, for $l \ge 0$ and $\beta = (\beta_1, \ldots, \beta_q)$. We proceed by induction. For q = 1 and $1 \le r \le d$, we have

$$|T_r(G)|_l = |\langle DG, (\gamma(F)DF)^r \rangle|_l$$

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and using (17) we obtain

$$|T_r(G)|_l \le C_l \sum_{l_1+l_2+l_3 \le l} |\gamma(F)|_{l_1} |G|_{l_2+1} |F|_{l_3+1} \le |G|_{l+1} |F|_{l+1} \sum_{l_1=0}^l |\gamma(F)|_{l_1},$$

where C_l is a constant which depends on *l* only. We obtain then by induction for every multi-index $\beta = (\beta_1, \dots, \beta_q)$

$$\left| T_{\beta}(G) \right|_{l} \leq C_{l,q} |G|_{l+q} |F|_{l+q}^{q} \sum_{l_{1}+\dots+l_{q} \leq l+q-1} \prod_{i=1}^{q} |\gamma(F)|_{l_{i}}.$$
 (26)

In particular this gives for l = 0

$$\left|T_{\beta}(G)\right| \leq C_q |G|_q |F|_q^q P_q(\gamma(F)),$$

with

$$P_q(\gamma(F)) = \sum_{l_1 + \dots + l_q \le q-1} \prod_{i=1}^q |\gamma(F)|_{l_i}, \quad q \ge 1.$$

To complete the notation, we note $P_0(\gamma(F)) = 1$. We obtain

$$\left|T_{\Gamma_{i}(\beta)}(L^{\gamma}_{\beta_{\lambda_{i}}}(F))\right| \leq C_{q} \left|L^{\gamma}_{\beta_{\lambda_{i}}}(F)\right|_{|\Gamma_{i}(\beta)|} |F|^{|\Gamma_{i}(\beta)|}_{|\Gamma_{i}(\beta)|} P_{|\Gamma_{i}(\beta)|}(\gamma(F)).$$

We turn now to the estimation of $|L_r^{\gamma}(F)|_l$. From the properties of the divergence operator δ (see Lemma 1)

$$\delta(\gamma(F)DF) = \gamma(F)\delta(DF) - \langle D\gamma(F), DF \rangle_{I}.$$

It follows from (14) and (16) that

$$|L_r^{\gamma}(F)|_l \le C_l |\gamma(F)|_{l+1} (|\delta(DF)|_l + |F|_{l+1}) \le C_l |\gamma(F)|_{l+1} (1 + |LF|_l) (1 + |F|_{l+1}),$$

and we get

$$\begin{aligned} \left| T_{\Gamma_{i}(\beta)}(L^{\gamma}_{\beta_{\lambda_{i}}}(F)) \right| &\leq C_{q} \left| \gamma(F) \right|_{|\Gamma_{i}(\beta)|+1} \left(1 + |LF|_{|\Gamma_{i}(\beta)|} \right) \\ &\times (1 + |F|_{|\Gamma_{i}(\beta)|+1}) |F|^{|\Gamma_{i}(\beta)|}_{|\Gamma_{i}(\beta)|} P_{|\Gamma_{i}(\beta)|}(\gamma(F)). \end{aligned}$$

$$(27)$$

Reporting these inequalities in (25) and recalling that $|\Gamma_0(\beta)| + \cdots + |\Gamma_j(\beta)| = q - j$ we deduce:

$$|H_{\beta}^{q}(F,G)| \leq |T_{\beta}(G)| + C_{q,d} \sum_{j=1}^{q} \sum_{k_{0}+\dots+k_{j}=q-j} |G|_{k_{0}} |F|_{k_{0}}^{k_{0}} P_{k_{0}}(\gamma(F)) \\ \times \left(\prod_{i=1}^{j} |\gamma(F)|_{k_{i}+1} P_{k_{i}}(\gamma(F)) |F|_{k_{i}}^{k_{i}} (1+|F|_{k_{i}+1})(1+|LF|_{k_{i}}) \right)$$

$$(28)$$

Now, for $q \ge 1$, we have from (19):

$$P_q(\gamma(F)) \le C_q \frac{1}{|\det \sigma(F)|^{2q-1}} (1+|F|_q)^{4dq},$$

so the following inequality holds for $q \ge 0$:

$$P_q(\gamma(F)) \le C_q \frac{1}{|\det \sigma(F)|^{2q}} (1+|F|_q)^{4dq}.$$

We obtain then for $k_0, k_1, \ldots, k_j \in \mathbb{N}$ such that $k_0 + \cdots + k_j = q - j$

$$\prod_{i=0}^{j} P_{k_i}(\gamma(F)) \le C_q \frac{1}{|\det \sigma(F)|^{2(q-j)}} (1+|F|_{q-j})^{4d(q-j)}$$
(29)

and once again from (19)

$$\prod_{i=1}^{j} |\gamma(F)|_{k_i+1} \le C_q \frac{1}{|\det \sigma(F)|^{q+j}} (1+|F|_{q-j+2})^{2d(q+j)}$$
(30)

it yields finally

$$\prod_{i=0}^{j} P_{k_i}(\gamma(F)) \prod_{i=1}^{j} |\gamma(F)|_{k_i+1} \le C_q \frac{1}{|\det \sigma(F)|^{3q-j}} (1+|F|_{q-j+2})^{6dq-2dj}.$$

Turning back to (28), it follows that

$$\left| H_{\beta}^{q}(F,G) \right| \leq \frac{C_{q,d} |G|_{q} (1+|F|_{q+1})^{(6d+1)q}}{|\det \sigma(F)|^{3q-1}} \left(1 + \sum_{j=1}^{q} \sum_{k_{1}+\dots+k_{j} \leq q-j} \prod_{i=1}^{j} |L(F)|_{k_{i}} \right),$$

and Theorem 3 is proved.

3 Stochastic equations with jumps

3.1 Notations and hypotheses

We consider a Poisson point process p with state space (E, B(E)), where $E = \mathbb{R}^d \times \mathbb{R}_+$. We refer to [12] for the notation. We denote by N the counting measure associated to p, we have $N([0, t) \times A) = \#\{0 \le s < t; p_s \in A\}$ for $t \ge 0$ and $A \in B(E)$. We assume that the associated intensity measure is given by $\widehat{N}(dt, dz, du) = dt \times d\mu(z) \times 1_{[0,\infty)}(u)du$ where $(z, u) \in E = \mathbb{R}^d \times \mathbb{R}_+$ and $\mu(dz) = h(z)dz$.

We are interested in the solution of the d dimensional stochastic equation

$$X_{t} = x + \int_{0}^{t} \int_{E} c(z, X_{s-}) \mathbf{1}_{\{u < \gamma(z, X_{s-})\}} N(ds, dz, du) + \int_{0}^{t} g(X_{s}) ds.$$
(31)

We remark that the infinitesimal generator of the Markov process X_t is given by

$$L\psi(x) = g(x)\nabla\psi(x) + \int_{\mathbb{R}^d} (\psi(x+c(z,x)) - \psi(x))K(x,dz)$$

where $K(x, dz) = \gamma(z, x)h(z)dz$ depends on the variable $x \in \mathbb{R}^d$. We refer to [4,14–18] for the links between Markov processes and pseudo-differential operators.

Our aim is to give sufficient conditions in order to prove that the law of X_t is absolutely continuous with respect to the Lebesgue measure and has a smooth density. In this section we make the following hypotheses on the functions γ , g, h and c.

Hypothesis 3.0 We assume that γ , g, h and c are infinitely differentiable functions in both variables z and x. Moreover we assume that g and its derivatives are bounded and that $\ln h$ has bounded derivatives.

Hypothesis 3.1 We assume that there exist two functions $\overline{\gamma}, \gamma : \mathbb{R}^d \to \mathbb{R}_+$ such that

$$\overline{C} \ge \overline{\gamma}(z) \ge \gamma(z, x) \ge \gamma(z) \ge 0, \quad \forall x \in \mathbb{R}^d$$

where \overline{C} is a constant.

Hypothesis 3.2 (i) We assume that there exists a non negative and bounded function $\overline{c} : \mathbb{R}^d \to \mathbb{R}_+$ such that $\int_{\mathbb{R}^d} \overline{c}(z) d\mu(z) < \infty$ and

$$|c(z,x)| + \left|\partial_z^\beta \partial_x^\alpha c(z,x)\right| \le \overline{c}(z) \quad \forall z,x \in \mathbb{R}^d \quad \forall \alpha,\beta.$$

We need this hypothesis in order to estimate the Sobolev norms.

(ii) There exists a measurable function $\widehat{c} : \mathbb{R}^d \to \mathbb{R}_+$ such that $\int_{\mathbb{R}^d} \widehat{c}(z) d\mu(z) < \infty$ and

$$\left\| \nabla_{x} c \times (I + \nabla_{x} c)^{-1}(z, x) \right\| \leq \widehat{c}(z), \quad \forall (z, x) \in \mathbb{R}^{d} \times \mathbb{R}^{d}.$$

In order to simplify the notations we assume that $\hat{c}(z) = \bar{c}(z)$.

(iii) There exists a non negative function $c : \mathbb{R}^d \to \mathbb{R}_+$ such that for every $z \in \mathbb{R}^d$

$$\sum_{r=1}^{d} \left\langle \partial_{z_r} c(z, x), \xi \right\rangle^2 \ge \underline{c}^2(z) \, |\xi|^2 \,, \quad \forall \xi \in \mathbb{R}^d$$

and we assume that there exists $\theta > 0$ such that

$$\underline{\lim}_{a \to +\infty} \frac{1}{\ln a} \int_{\{\underline{c}^2 \ge 1/a\}} \underline{\gamma}(z) d\mu(z) = \theta.$$

Remark assumptions ii) and iii) give sufficient conditions to prove the non degeneracy of the Malliavin covariance matrix as defined in the previous section. In particular the second part of iii) implies that \underline{c}^2 is a (p, t) broad function (see [6]) for $p/t < \theta$. Notice that we may have $\underline{c}(z) = 0$ for some $z \in \mathbb{R}^d$.

We add to these hypotheses some assumptions on the derivatives of γ and $\ln \gamma$ with respect to x and z. For $l \ge 1$ we use the notation :

$$\overline{\gamma}^{x,l}(z) = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,x}\gamma(z,x)|,$$

$$\overline{\gamma}_{\ln}^{x,l}(z) = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,x}\ln\gamma(z,x)|,$$

$$\overline{\gamma}_{\ln}^{z,l}(z) = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,z}\ln\gamma(z,x)|.$$

Hypothesis 3.3 We assume that $\ln \gamma$ has bounded derivatives with respect to z (that is $\overline{\gamma}_{\ln}^{z,l}(z)$ is bounded) and that γ has bounded derivatives with respect to x such that $\forall z \in \mathbb{R}^d, \overline{\gamma}^{x,l}(z) \leq \overline{\gamma}^{x,l}$; moreover we assume that

$$\sup_{z^*\in\mathbb{R}^d}\int_{B(z^*,1)}\overline{\gamma}(z)d\mu(z)<+\infty.$$

We complete this hypothesis with two alternative hypotheses.

(a) (weak dependence on *x*) We assume that $\forall l \ge 1$

$$\int_{\mathbb{R}^d} \overline{\gamma}_{\ln}^{x,l}(z) \overline{\gamma}(z) d\mu(z) < \infty.$$

(b) (strong dependence on x) We assume that $\ln \gamma$ has bounded derivatives with respect to x such that $\forall l \ge 1$

$$\forall z \in \mathbb{R}^d, \quad \overline{\gamma}_{\ln}^{x,l}(z) \le \overline{\gamma}_{\ln}^{x,l}.$$

Remark If μ is the Lebesgue measure (case h = 1) and if γ does not depend on z then $\overline{\gamma}_{\ln}^{x,l}$ is constant and consequently Hypothesis 3.3.a fails. Conversely, if $\gamma(z, x) = \gamma(z)$ then Hypothesis 3.3.a is satisfied as soon as $\ln \gamma$ has bounded derivatives. This last case corresponds to the standard case where the law of the amplitude of the jumps does not depend on the position of X_t . Under Hypothesis 3.3.a we are in a classical situation where the divergence does not blow up and this leads to an integration by part formula with bounded weights (see Proposition 4 and Lemma 11). On the contrary under assumption 3.3.*b*, the divergence can blow up as well as the weights appearing in the integration by part formula.

3.2 Main results and examples

Our methodology to study the regularity of the law of the random variable X_t is based on the following result. Let $\hat{p}_X(\xi) = E(e^{i\langle\xi,X\rangle})$ be the Fourier transform of a *d*-dimensional random variable X then using the Fourier inversion formula, one can prove that if $\int_{\mathbb{R}^d} |\xi|^p |\hat{p}_X(\xi)| d\xi < \infty$ for p > 0 then the law of X is absolutely continuous with respect to the Lebesgue measure on \mathbb{R}^d and its density is $\mathcal{C}^{[p]}$, where [p] denotes the integer part of p.

To apply this result, we just have to bound the Fourier transform of X_t in terms of $1/|\xi|$. This is done in the next proposition. The proof of this proposition needs a lot of steps that we detail in the next sections and it will be given later.

Proposition 3 Let $B_M = \{z \in \mathbb{R}^d; |z| < M\}$, then under Hypotheses 3.0., 3.1. 3.2. and 3.3 we have for all $M \ge 1$, for $q \ge 1$ and t > 0 such that $4d(3q - 1)/t < \theta$

(a) *if* 3.3.*a* holds

$$|\hat{p}_{X_{t}}(\xi)| \leq t \int_{B_{M-1}^{c}} \underline{c}^{2}(z)\underline{\gamma}(z)d\mu(z)\frac{1}{2}|\xi|^{2} + |\xi|te^{Ct} \int_{B_{M}^{c}} \overline{c}(z)\overline{\gamma}(z)d\mu(z) + \frac{C_{q}}{|\xi|^{q}}.$$

(b) *if* 3.3.*b* holds

$$\begin{split} |\hat{p}_{X_{t}}(\xi)| &\leq t \int\limits_{B_{M-1}^{c}} \underline{c}^{2}(z)\underline{\gamma}(z)d\mu(z)\frac{1}{2}\,|\xi|^{2} + |\xi|\,te^{Ct} \int\limits_{B_{M}^{c}} \overline{c}(z)\overline{\gamma}(z)d\mu(z) \\ &+ \frac{C_{q}(1+\mu(B_{M+1})^{q})}{|\xi|^{q}}. \end{split}$$

We can remark that if $\theta = +\infty$ then the result holds $\forall q \ge 1$ and $\forall t > 0$.

By choosing *M* judiciously as a function of ξ in the inequalities given in Proposition 3, we obtain $|\hat{p}_{X_t}(\xi)| \leq C/|\xi|^p$ for some p > 0 and this permits us to deduce some regularity for the density of X_t . The next theorem precise the optimal choice of *M* with respect to ξ and permits us to derive the regularity of the law of the process X_t .

Theorem 4 We assume that Hypotheses 3.0., 3.1., 3.2 and 3.3. hold.

- (a) Assuming 3.3.*a*, the law of X_t admits a density C^k if $t > (3k + 3d 1)\frac{4d}{\theta}$. In the case $\theta = \infty$, the law of X_t admits a density C^{∞} .
- (b) Assuming 3.3.b and the two following hypotheses
 A1: ∃p₁, p₂ > 0 such that :

$$\limsup_{M} M^{p_1} \int_{B^c_M} \overline{c}(z)\overline{\gamma}(z)d\mu(z) < +\infty;$$
$$\limsup_{M} M^{p_2} \int_{B^c_M} \underline{c}^2(z)\underline{\gamma}(z)d\mu(z) < +\infty;$$

A2:
$$\exists \rho > 0$$
 such that $\mu(B_M) \leq CM^{\rho}$ where $B_M = \{z \in \mathbb{R}^d; |z| < M\};$

Case 1 if $\theta = +\infty$ then the law of X_t admits a density \mathcal{C}^k with $k < \min(p_1/\rho - 1 - d, p_2/\rho - 2 - d)$ if $\min(p_1/\rho - 1 - d, p_2/\rho - 2 - d) \ge 1$.

Case 2 if $0 < \theta < \infty$ let $q^*(t, \theta) = \left[\frac{1}{3}\left(\frac{t\theta}{4d} + 1\right)\right]$; then the law of X_t admits a density C^k for $k < \sup_{0 < r < 1/\rho} \min(rp_1 - 1 - d, rp_2 - 2 - d, q^*(t, \theta)(1 - r\rho) - d)$, if for some $0 < r < 1/\rho$, $\min(rp_1 - 1 - d, rp_2 - 2 - d, q^*(t, \theta)(1 - r\rho) - d) \ge 1$.

Proof (a) Assuming 3.3.*a* and letting M go to infinity in the right-hand side of the inequality given in Proposition 3, we deduce

$$|\hat{p}_{X_t}(\xi)| \le C/|\xi|^q,$$

and the result follows.

(b) From A1, for M large enough, we have

$$\int\limits_{B_M^c} \overline{c}(z)\overline{\gamma}(z)d\mu(z) \le C/M^p$$

and

$$\int_{B_{M-1}^c} \underline{c}^2(z) \underline{\gamma}(z) d\mu(z) \le C/M^{p_2}.$$

Now assuming 3.3.*b* and A2 and choosing $M = |\xi|^r$, for $0 < r < 1/\rho$, we obtain from Proposition 3

$$|\hat{p}_{X_t}(\xi)| \le C\left(\frac{1}{|\xi|^{rp_1-1}} + \frac{1}{|\xi|^{rp_2-2}} + \frac{1}{|\xi|^{q(1-r\rho)}}\right),$$

for q and t such that $4d(3q-1)/t < \theta$. Now if $\theta = \infty$, we obtain for q large enough

$$|\hat{p}_{X_t}(\xi)| \le C\left(\frac{1}{|\xi|^{rp_1-1}} + \frac{1}{|\xi|^{rp_2-2}}\right).$$

In the case $\theta < \infty$, the best choice of q is $q^*(t, \theta)$. This achieves the proof of theorem 4.

We end this section with some examples in order to illustrate the results of Theorem 4.

Example 1 In this example we assume that h = 1 so $\mu(dz) = dz$ and that $\gamma(z)$ is equal to a constant $\gamma > 0$. We also assume that Hypothesis 3.3.b holds. We have $\mu(B_M) = r_d M^d$ where r_d is the volume of the unit ball in \mathbb{R}^d so $\rho = d$. We will consider two types of behaviour for c.

(i) Exponential decay: we assume that $\overline{c}(z) = e^{-b|z|^c}$ and $\underline{c}(z) = e^{-a|z|^c}$ for some constants $0 < b \le a$ and c > 0. We have

$$\int_{\underline{|c^2|>1/u|}} \underline{\gamma}(z) d\mu(z) = \frac{\underline{\gamma}r_d}{(2a)^{d/c}} \times (\ln u)^{d/c}.$$

We deduce then

$$\theta = 0$$
 if $c > d$, $\theta = \infty$ if $0 < c < d$ and $\theta = \frac{\gamma r_d}{2a}$ if $c = d$. (32)

If c > d, Hypothesis 3.2.iii fails, this is coherent with the result of [6]. Now observe that

$$\int_{B_M^c} \underline{c}^2(z) \underline{\gamma}(z) d\mu(z) + \int_{B_M^c} \overline{c}(z) \overline{\gamma}(z) d\mu(z) \le e^{-\eta |z|^c}$$

for some $\eta > 0$ so $p_1 = p_2 = \infty$. In the case 0 < c < d we obtain a density C^{∞} for every t > 0. In the case c = d we have $q^*(t, \theta) = \left[\frac{1}{3}\left(1 + \frac{\gamma r_d}{8da} \times t\right)\right]$. If $t < 8da(3d + 2)/(\gamma r_d)$ we obtain nothing and if $t \ge 8da(3d + 2)/(\gamma r_d)$ we obtain a density C^k where k is the largest integer less than $\left[\frac{1}{3}\left(1 + \frac{\gamma r_d}{8da} \times t\right)\right] - d$. (ii) Polynomial decay. We assume that $\overline{c}(z) = b/(1 + |z|^p)$ and $\underline{c}(z) = a/(1 + |z|^p)$ for some constants $0 < a \le b$ and p > d. We have

$$\int_{\{\underline{c}^2 > 1/u\}} \underline{\gamma}(z) d\mu(z) = \underline{\gamma} r_d \times (a\sqrt{u} - 1)^{d/p}$$

so $\theta = \infty$ and our result works for every t > 0. Hence a simple computation gives

$$\int\limits_{B_{M}^{c}} \underline{c}^{2}(z)\underline{\gamma}(z)d\mu(z) \leq \frac{C}{M^{2p-d}}, \quad \int\limits_{B_{M}^{c}} \overline{c}(z)\overline{\gamma}(z)d\mu(z) \leq \frac{C}{M^{p-d}}$$

and then $p_1 = p - d$ and $p_2 = 2p - d$. If $p \ge d(d + 3)$ then $\min(p/d - 2 - d, 2p/d - 3 - d) \ge 1$ and we obtain a density C^k with $k < \frac{p}{d} - d - 2$. Conversely if

p < d(d + 3), we can say nothing about the regularity of the density of X_t . We give now an example where the function γ satisfies Hypothesis 3.3.a.

Example 2 As in the preceding example, we assume h = 1. We consider the function $\gamma(z, x) = \exp(-\alpha(x)/(1 + |z|^q))$ for some q > d. We assume that α is a smooth function which is bounded and has bounded derivatives and moreover there exists two constants such that $\overline{\alpha} \ge \alpha(x) \ge \underline{\alpha} > 0$. Notice that the derivatives with respect to x of $\ln \gamma(z, x)$ are bounded by $C/(1 + |z|^q)$ which is integrable with respect to the Lebesgue measure if q > d. So Hypothesis 3.3.a is true. Moreover we check that $\gamma(z) = \exp(-\overline{\alpha}/(1 + |z|^q))$.

(i) Exponential decay. We take c as in Example 1.i). It follows that

$$\int_{\underline{\{c^2>1/u\}}} \underline{\gamma}(z) d\mu(z) \ge \exp(-\overline{\alpha}) \frac{r_d}{(2a)^{d/c}} \times (\ln u)^{d/c}.$$

So we obtain once again θ as in (32). In the case c > d we can say nothing, in the case c < d we obtain a density C^{∞} and in the case c = d we have $\theta = \frac{r_d}{2a}$ and we obtain a density C^k if $t > \frac{8ad(3k+3d-1)}{r_d}$. In particular we have no results if $t \leq \frac{8ad(3d-1)}{r_d}$. Notice that the only difference with respect to the previous example concerns the case c = d when we have a slight gain.

(ii) Polynomial decay. At last we take c as in the example 1.ii). We check that $\theta = \infty$ so we obtain a density C^{∞} , which is a better result than the one of the previous example.

Example 3 We consider the process (Y_t) solution of the stochastic equation

$$dY_t = f(Y_t)dL_t,$$

where L_t is a Lévy process with intensity measure $|y|^{-(1+\rho)} \mathbb{1}_{\{|y| \le 1\}} dy$, with $0 < \rho < 1$. The infinitesimal generator of *Y* is given by

$$L\psi(x) = \int_{\{|y| \le 1\}} (\psi(x + f(x)y) - \psi(x)) \frac{dy}{|y|^{1+\rho}}.$$

If we introduce some function g(x) in this operator we obtain

$$L\psi(x) = \int_{\{|y| \le 1\}} (\psi(x + f(x)y) - \psi(x))g(x)\frac{dy}{|y|^{1+\rho}}.$$

We are interested to represent this operator through a stochastic equation. In order to come back in our framework, we translate the integrability problem from 0 to ∞ by the change of variables $z = y^{-1}$ and we obtain

$$L\psi(x) = \int_{\{|z| \ge 1\}} (\psi(x + f(x)z^{-1}) - \psi(x))g(x)\frac{dz}{|z|^{1-\rho}}.$$

$$X_t = x + \int_0^t \int_{\mathbb{R} \times \mathbb{R}_+} f(X_{s-}) z^{-1} \mathbf{1}_{\{u < g(X_{s-})\}} N(ds, dz, du).$$

We have $E = \mathbb{R} \times \mathbb{R}_+$, $d\mu(z) = \frac{1}{|z|^{1-\rho}} \mathbb{1}_{\{|z| \ge 1\}} dz$, $c(z, x) = f(x)z^{-1}$ and $\gamma(z, x) = g(x)$. We make the following assumptions. There exist two constants \underline{f} and \overline{f} such that $\forall x \ \underline{f} \le f(x) \le \overline{f}$ and we suppose that all derivatives of f are bounded by \overline{f} . Moreover we assume that there exist two constants \underline{g} and \overline{g} such that g and its derivative are bounded by \overline{g} and $0 < \underline{g} \le g(x)$, $\forall x$. Consequently it is easy to check that Hypotheses 3.0., 3.1., 3.2. and 3.3.*b* are satisfied, with $\theta = +\infty$. Moreover we have $\mu(B_M) \le CM^{\rho}$ and A2 holds with $p_1 = 1-\rho$ and $p_2 = 2-\rho$. Consequently we deduce that the law of X_t admits a density \mathcal{C}^k with $k < 1/\rho - 3$ if $1/\rho - 3 \ge 1$.

The next sections are the successive steps to prove Proposition 3.

3.3 Approximation of X_t

In order to prove that the process X_t , solution of (31), has a smooth density, we will apply the differential calculus and the integration by parts formula of Sect. 2. But since the random variable X_t can not be viewed as a simple functional, the first step consists in approximate it. We describe in this section our approximation procedure. We consider a non-negative and smooth function $\varphi : \mathbb{R}^d \to \mathbb{R}_+$ such that $\varphi(z) = 0$ for |z| > 1 and $\int_{\mathbb{R}^d} \varphi(z) dz = 1$. And for $M \in \mathbb{N}$ we denote $\Phi_M(z) = \varphi * 1_{B_M}$ with $B_M = \{z \in \mathbb{R}^d : |z| < M\}$. Then $\Phi_M \in C_b^\infty$ and we have $1_{B_{M-1}} \le \Phi_M \le 1_{B_{M+1}}$. We denote by X_t^M the solution of the equation

$$X_{t}^{M} = x + \int_{0}^{t} \int_{E} c_{M}(z, X_{s-}^{M}) \mathbb{1}_{\{u < \gamma(z, X_{s-}^{M})\}} N(ds, dz, du) + \int_{0}^{t} g(X_{s}^{M}) ds.$$
(33)

where $c_M(z, x) := c(z, x)\Phi_M(z)$. Observe that (33) is obtained from (31) replacing the coefficient *c* by the truncating one c_M . Let $N_M(ds, dz, du) := 1_{B_{M+1}}(z) \times 1_{[0,2\overline{C}]}(u)N(ds, dz, du)$. Since $\{u < \gamma(z, X_{s-}^M)\} \subset \{u < 2\overline{C}\}$ and $\Phi_M(z) = 0$ for |z| > M + 1, we may replace *N* by N_M in the above equation and consequently X_t^M is solution of the equation

$$X_t^M = x + \int_0^t \int_E c_M(z, X_{s-}^M) \mathbf{1}_{\{u < \gamma(z, X_{s-}^M)\}} N_M(ds, dz, du) + \int_0^t g(X_s^M) ds.$$

Since the intensity measure \widehat{N}_M is finite we may represent the random measure N_M by a compound Poisson process. Let $\lambda_M = 2\overline{C} \times \mu(B_{M+1}) = t^{-1}E(N_M(t, E))$ and let J_t^M a Poisson process of parameter λ_M . We denote by T_k^M , $k \in \mathbb{N}$ the jump times of J_t^M . We also consider two sequences of independent random variables $(Z_k^M)_{k \in \mathbb{N}}$ and $(U_k)_{k \in \mathbb{N}}$ respectively in \mathbb{R}^d and \mathbb{R}_+ which are independent of J^M and such that

$$Z_k \sim \frac{1}{\mu(B_{M+1})} \mathbb{1}_{B_{M+1}}(z) d\mu(z), \text{ and } U_k \sim \frac{1}{2\overline{C}} \mathbb{1}_{[0,2\overline{C}]}(u) du.$$

To simplify the notation, we omit the dependence on M for the variables (T_k^M) and (Z_k^M) . Then (33) may be written as

$$X_{t}^{M} = x + \sum_{k=1}^{J_{t}^{M}} c_{M}(Z_{k}, X_{T_{k}-}^{M}) \mathbf{1}_{(U_{k},\infty)}(\gamma(Z_{k}, X_{T_{k}-}^{M})) + \int_{0}^{t} g(X_{s}^{M}) ds.$$
(34)

Lemma 4 Assume that Hypotheses 3.0., 3.1., 3.2 and 3.3. hold true then we have

$$E\left|X_{t}^{M}-X_{t}\right| \leq \varepsilon_{M} := te^{Ct} \int_{\{|z|>M\}} \overline{c}(z)\overline{\gamma}(z)d\mu(z),$$
(35)

for some constant C.

Proof We have $E |X_t^M - X_t| \le I_M^1 + I_M^2$ with

$$I_{M}^{1} = E \int_{0}^{t} \int_{\mathbb{R}^{d}} \int_{0}^{\overline{C}} \left| c(z, X_{s}) 1_{\{u < \gamma(z, X_{s})\}} - c_{M}(z, X_{s}^{M}) 1_{\{u < \gamma(z, X_{s}^{M})\}} \right| du d\mu(z) ds$$

$$I_{M}^{2} = E \int_{0}^{t} \left| g(X_{s}) - g(X_{s}^{M}) \right| ds.$$

Since $|\nabla_x c(z, x)| \le \overline{c}(z)$ we have $I_M^1 \le I_M^{1,1} + I_M^{1,2}$ with

$$I_M^{1,1} = E \int_0^t \int_{\mathbb{R}^d} \int_0^{\overline{C}} \left| c(z, X_s) - c_M(z, X_s^M) \right| 1_{\{u < \overline{\gamma}(z)\}} du d\mu(z) ds$$

$$\leq t \int_{\mathbb{R}^d} \overline{c}(z) \overline{\gamma}(z) (1 - \Phi_M(z)) d\mu(z) + \int_{\mathbb{R}^d} \overline{c}(z) \overline{\gamma}(z) dz \times E \int_0^t \left| X_s - X_s^M \right| ds$$

and, since $|\nabla_x \gamma(z, x)| \leq \overline{\gamma}^{x, 1}$

$$\begin{split} I_M^{1,2} &= E \int_0^t \int_{\mathbb{R}^d} \int_0^{\overline{C}} \overline{c}(z) \left| \mathbf{1}_{\{u < \gamma(z, X_s)\}} - \mathbf{1}_{\{u < \gamma(z, X_s^M)\}} \right| du d\mu(z) ds \\ &= E \int_0^t \int_{\mathbb{R}^d} \overline{c}(z) \left| \gamma(z, X_s) - \gamma(z, X_s^M) \right| d\mu(z) ds \\ &\leq \int_{\mathbb{R}^d} \overline{c}(z) \overline{\gamma}^{x,1} d\mu(z) \times E \int_0^t \left| X_s - X_s^M \right| ds. \end{split}$$

A similar inequality holds for I_M^2 so we obtain

$$E\left|X_{t}^{M}-X_{t}\right| \leq t \times \int_{R^{d}} \overline{\gamma}(z)\overline{c}(z)(1-\Phi_{M}(z))d\mu(z) + C\int_{0}^{t} E\left|X_{s}-X_{s}^{M}\right|ds.$$

We conclude by using Gronwall's lemma.

The random variable X_t^M solution of (34) is a function of $(Z_1 \dots, Z_{J_t^M})$ but it is not a simple functional, as defined in Sect. 2 because the coefficient $c_M(z, x)1_{(u,\infty)}$ $(\gamma(z, x))$ is not differentiable with respect to z. In order to avoid this difficulty we use the following alternative representation. Let $z_M^* \in \mathbb{R}^d$ such that $|z_M^*| = M + 3$. We define

$$q_{M}(z,x) := \varphi(z - z_{M}^{*})\theta_{M,\gamma}(x) + \frac{1}{2\overline{C}\mu(B_{M+1})} \mathbf{1}_{B_{M+1}}(z)\gamma(z,x)h(z)$$

$$\theta_{M,\gamma}(x) := \frac{1}{\mu(B_{M+1})} \int_{\{|z| \le M+1\}} \left(1 - \frac{1}{2\overline{C}}\gamma(z,x)\right)\mu(dz).$$
 (36)

We recall that φ is the function defined at the beginning of this subsection : a nonnegative and smooth function with $\int \varphi = 1$ and which is null outside the unit ball. Moreover from Hypothesis 3.1, $0 \le \gamma(z, x) \le \overline{C}$ and then $1 \ge \theta_{M,\gamma}(x) \ge 1/2$. By construction the function q_M satisfies $\int q_M(x, z)dz = 1$. Hence we can check that

$$E(f(X_{T_k}^M) \mid X_{T_k-}^M = x) = \int_{R^d} f(x + c_M(z, x))q_M(z, x)dz.$$
 (37)

In fact the left-hand side term of (37) is equal to I + J with

$$I = E(f(X_{T_k}^M) 1_{\{U_k \ge \gamma(Z_k, X_{T_k-}^M)\}} | X_{T_k-}^M = x) \text{ and}$$

$$J = E(f(X_{T_k}^M) 1_{\{U_k < \gamma(Z_k, X_{T_k-}^M)\}} | X_{T_k-}^M = x).$$

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A simple calculation leads to

$$I = f(x)P(U_k \ge \gamma(Z_k, x)) = f(x)\theta_{M,\gamma}(x) = \int_{|z| > M+1} f(x + c_M(z, x))q_M(z, x)dz$$

where the last equality results from the fact that $c_M(z, x) = 0$ for |z| > M + 1. Moreover one can easily see that $J = \int_{|z| \le M+1} f(x + c_M(z, x))q_M(z, x)dz$ and (37) is proved.

From the relation (37) we construct a process (\overline{X}_t^M) equal in law to (X_t^M) on the following way.

We denote by $\Psi_t(x)$ the solution of $\Psi_t(x) = x + \int_0^t g(\Psi_s(x)) ds$. We assume that the times $T_k, k \in \mathbb{N}$ are fixed and we consider a sequence $(z_k)_{k \in \mathbb{N}}$ with $z_k \in \mathbb{R}^d$. Then we define $x_t, t \ge 0$ by $x_0 = x$ and, if x_{T_k} is given, then

$$x_t = \Psi_{t-T_k}(x_{T_k}) \quad T_k \le t < T_{k+1},$$

$$x_{T_{k+1}} = x_{T_{k+1}^-} + c_M(z_{k+1}, x_{T_{k+1}^-}).$$

We remark that for $T_k \le t < T_{k+1}$, x_t is a function of z_1, \ldots, z_k . Notice also that x_t solves the equation

$$x_t = x + \sum_{k=1}^{J_t^M} c_M(z_k, x_{T_k^-}) + \int_0^t g(x_s) ds.$$

We consider now a sequence of random variables $(\overline{Z}_k), k \in \mathbb{N}^*$ and we denote $\mathcal{G}_k = \sigma(T_p, p \in \mathbb{N}) \lor \sigma(\overline{Z}_p, p \leq k)$ and $\overline{X}_t^M = x_t(\overline{Z}_1, \dots, \overline{Z}_{J_t^M})$. We assume that the law of \overline{Z}_{k+1} conditionally on \mathcal{G}_k is given by

$$P(\overline{Z}_{k+1} \in dz \mid \mathcal{G}_k) = q_M(x_{T_{k+1}^-}(\overline{Z}_1, \dots, \overline{Z}_k), z)dz = q_M(\overline{X}_{T_{k+1}^-}^M, z)dz.$$

Clearly \overline{X}_t^M satisfies the equation

$$\overline{X}_{t}^{M} = x + \sum_{k=1}^{J_{t}^{M}} c_{M}(\overline{Z}_{k}, \overline{X}_{T_{k}-}^{M}) + \int_{0}^{t} g(\overline{X}_{s}^{M}) ds$$
(38)

and \overline{X}_t^M has the same law as X_t^M . Moreover we can prove a little bit more. **Lemma 5** For a locally bounded and measurable function $\psi : \mathbb{R}^d \to \mathbb{R}$ let

$$\overline{S}_t(\psi) = \sum_{k=1}^{J_t^M} (\Phi_M \psi)(\overline{Z}_k), \quad S_t(\psi) = \sum_{k=1}^{J_t^M} (\Phi_M \psi)(Z_k) \mathbb{1}_{\{\gamma(Z_k, X^M(T_k-)) > U_k\}},$$

then $(\overline{X}_t^M, \overline{S}_t(\psi))_{t\geq 0}$ has the same law as $(X_t^M, S_t(\psi))_{t\geq 0}$.

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Proof Observing that $(\overline{X}_t^M, \overline{S}_t(\psi))_{t\geq 0}$ solves a system of equations similar to (38) but in dimension d + 1, it suffices to prove that $(\overline{X}_t^M)_{t\geq 0}$ has the same law as $(X_t^M)_{t\geq 0}$. This readily follows from

$$E\left(f(X_{T_{k+1}}^M)\left|X_{T_{k+1}}^M=x\right)=E\left(f\left(\overline{X}_{T_{k+1}}^M\right)\right|\overline{X}_{T_{k+1}}^M=x\right)$$

which is a consequence of (37).

Remark Looking at the infinitesimal generator L of X it is clear that the natural approximation of X_t is \overline{X}_t^M instead of X_t^M . But we use the representation given by X_t^M for two reasons. First it is easier to obtain estimates for this process because we have a stochastic equation and so we may use the stochastic calculus associated to a Poisson point measure. Moreover, having this equation in mind, gives a clear idea about the link with other approaches by Malliavin calculus to the solution of a stochastic equation with jumps: we mainly think to [6]. Remark that X_t is solution of an equation with discontinuous coefficients so the approach developed by Bichteler et al. [6] does not work. And if we consider the equation so it is no more a Poisson point measure.

3.4 The integration by parts formula

The random variable \overline{X}_{t}^{M} constructed previously is a simple functional but unfortunately its Malliavin covariance matrix is degenerated. To avoid this problem we use a classical regularization procedure. Instead of the variable \overline{X}_{t}^{M} , we consider the regularized one F_{M} defined by

$$F_M = \overline{X}_t^M + \sqrt{U_M(t)} \times \Delta, \qquad (39)$$

where Δ is a *d*-dimensional standard Gaussian variable independent of the variables $(\overline{Z}_k)_{k\geq 1}$ and $(T_k)_{k\geq 1}$ and $U_M(t)$ is defined by

$$U_M(t) = t \int_{\substack{B_{M-1}^c \\ B_{M-1}^c}} \underline{c}^2(z) \underline{\gamma}(z) d\mu(z).$$
(40)

We observe that $F_M \in S^d$ where S is the space of simple functionals for the differential calculus based on the variables $(\overline{Z}_k)_{k\in\mathbb{N}}$ with $\overline{Z}_0 = (\Delta^r)_{1\leq r\leq d}$ and $\overline{Z}_k = (\overline{Z}_k^r)_{1\leq r\leq d}$ and we are now in the framework of Sect. 2 by taking $\mathcal{G} = \sigma(T_k, k \in \mathbb{N})$ and defining the weights (π_k) by $\pi_0^r = 1$ and $\pi_k^r = \Phi_M(\overline{Z}_k)$ for $1 \leq r \leq d$. Conditionally on \mathcal{G} ,

the density of the law of $(\overline{Z}_1, \ldots, \overline{Z}_{L^M})$ is given by

$$p_M(\omega, z_1, \dots, z_{J_t^M}) = \prod_{j=1}^{J_t^M} q_M\left(z_j, \Psi_{T_j - T_{j-1}}\left(\overline{X}_{T_{j-1}}^M\right)\right)$$

where $\overline{X}_{T_{j-1}}^{M}$ is a function of z_i , $1 \le i \le j-1$. We can check that p_M satisfies the hypothesis H1 of Sect. 2.

To clarify the notation, the derivative operator can be written in this framework for $F \in S$ by $DF = (D_{k,r}F)$ where $D_{k,r} = \pi_k^r \partial_{\overline{Z}_k^r}$ for $k \ge 0$ and $1 \le r \le d$. Consequently we deduce that $D_{k,r}F_M^{r'} = D_{k,r}\overline{X}_t^{M,r'}$, for $k \ge 1$ and $D_{0,r}F_M^{r'} = \sqrt{U_M(t)}\delta_{r,r'}$ with $\delta_{r,r'} = 0$ if $r \ne r'$, $\delta_{r,r'} = 1$ otherwise.

The Malliavin covariance matrix of \overline{X}_{t}^{M} is equal to

$$\sigma(\overline{X}_t^M)^{i,j} = \sum_{k=1}^{J_t^M} \sum_{r=1}^d D_{k,r} \overline{X}_t^{M,i} D_{k,r} \overline{X}_t^{M,j}$$

for $1 \le i, j \le d$ and finally the Malliavin covariance matrix of F_M is given by

$$\sigma(F_M) = \sigma(\overline{X}_t^M) + U_M(t) \times Id.$$

Using the results of Sect. 2, we can state an integration by part formula and give a bound for the weight $H^q(F_M, 1)$ in terms of the Sobolev norms of F_M , the divergence LF_M and the determinant of the inverse of the Malliavin covariance matrix det $\sigma(F_M)$. The control of these last three quantities is rather technical and is studied in detail in Sect. 4.

Proposition 4 Assume Hypotheses 3.0. 3.1. 3.2. and let $\phi : \mathbb{R}^d \to \mathbb{R}$ be a bounded smooth function with bounded derivatives. For every multi-index $\beta = (\beta_1, \dots, \beta_q) \in \{1, \dots, d\}^q$ such that $4d(3q-1)/t < \theta$

(a) *if 3.3.a holds then*

$$\left| E(\partial_{\beta}\phi(F_M)) \right| \le C_q \, \|\phi\|_{\infty} \,. \tag{41}$$

(b) *if 3.3.b holds then*

$$|E(\partial_{\beta}\phi(F_{M}))| \le C_{q} \|\phi\|_{\infty} (1 + \mu(B_{M+1})^{q}), \tag{42}$$

Remark If $\theta = \infty$ then $\forall t > 0$, we have an integration by parts formula for any order of derivation q. Conversely if θ is finite, we need to have t large enough to integrate q times by part.

Proof The integration by parts formula (11) gives, for every smooth $\phi : \mathbb{R}^d \to \mathbb{R}$ and every multi-index $\beta = (\beta_1, \dots, \beta_q)$

$$E(\partial_{\beta}\phi(F_M)) = E(\phi(F_M)H^q_{\beta}(F_M, 1)),$$

and consequently

$$\left| E(\partial_{\beta}\phi(F_M)) \right| \le \|\phi\|_{\infty} E(|H^q_{\beta}(F_M, 1)|).$$

So we just have to bound $|H_{\beta}^{q}(F_{M}, 1)|$. From the second part of Theorem 3 we have

$$|H^{q}(F_{M},1)| \leq C_{q} \frac{1}{|\det \sigma(F_{M})|^{3q-1}} (1+|F_{M}|_{q+1}^{(6d+1)q}) (1+|LF_{M}|_{q-1}^{q}).$$

Now from Lemma 13 (see Sect. 4), we have:

(a) assuming 3.3.*a*, for $l, p \ge 1$,

$$E|LF_M|_l^p \leq C_{l,p};$$

(b) assuming 3.3.*b*, for $l, p \ge 1$,

$$E|LF_M|_l^p \le C_{l,p}(1+\mu(B_{M+1})^p).$$

Hence from Lemma 9, for $l, p \ge 1$

$$E|F_M|_l^p \le C_{l,p};$$

and from Lemma 16, we have for $p \ge 1, t > 0$ such that $2dp/t < \theta$

$$E\frac{1}{\det\sigma(F_M))^p} \le C_p.$$

The final result is then a straightforward consequence of Cauchy-Schwarz inequality.

3.5 Estimates for the Fourier transform of X_t

In this section, we prove Proposition 3.

Proof The proof consists first to approximate X_t by \overline{X}_t^M and then to apply the integration by parts formula.

Approximation. We have

$$\left| E(e^{i\langle \xi, X_t \rangle}) \right| \leq |\xi| E \left| X_t - \overline{X}_t^M \right| + \left| E(e^{i\left\langle \xi, \overline{X}_t^M \right\rangle} - e^{i\langle \xi, F_M \rangle}) \right| + \left| E(e^{i\langle \xi, F_M \rangle}) \right|.$$

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From (35) we deduce

$$E\left(\left|X_t-\overline{X}_t^M\right|\right) \leq \varepsilon_M = te^{Ct}\int\limits_{B_M^c} \overline{c}(z)\overline{\gamma}(z)d\mu(z).$$

Moreover

$$E(e^{i\left\langle\xi,\overline{X}_{t}^{M}\right\rangle} - e^{i\left\langle\xi,F_{M}\right\rangle}) = E(e^{i\left\langle\xi,\overline{X}_{t}^{M}\right\rangle}(1 - e^{i\left\langle\xi,\sqrt{U_{M}(t)}\Delta\right\rangle}))$$
$$= E(e^{i\left\langle\xi,\overline{X}_{t}^{M}\right\rangle})(1 - e^{-\frac{1}{2}|\xi|^{2}U_{M}(t)}),$$

so that

$$\left| E(e^{i\left\langle \xi, \overline{X}_t^M \right\rangle} - e^{i\left\langle \xi, F_M \right\rangle}) \right| \le U_M(t) \frac{1}{2} |\xi|^2.$$

We conclude that

$$\left| E(e^{i\langle \xi, X_t \rangle}) \right| \le U_M(t) \frac{1}{2} \left| \xi \right|^2 + \left| \xi \right| t e^{Ct} \int_{B_M^c} \overline{c}(z) d\mu(z) + \left| E(e^{i\langle \xi, F_M \rangle}) \right|.$$

Integration by parts. We denote $e_{\xi}(x) = \exp(i \langle \xi, x \rangle)$ and we have $\partial_{\beta} e_{\xi}(x) = i^{|\beta|} \xi_{\beta_1} \dots \xi_{\beta_q} e_{\xi}(x)$. Consequently

(a) assuming 3.3.*a* and applying (41) for β such that $|\beta| = q$ we obtain

$$\left| E(e^{i \langle \xi, F_M \rangle}) \right| \leq \frac{C_q}{|\xi|^q},$$

(b) assuming 3.3.b, we obtain similarly from (42)

$$\left|\xi_{\beta_1}\ldots\xi_{\beta_q}\right|\left|E(e^{i\langle\xi,F_M\rangle})\right| = \left|E(\partial_\beta e_{\xi}(F_M))\right| \le C_q(1+\mu(B_{M+1})^q),$$

and then

$$\left| E(e^{i \langle \xi, F_M \rangle}) \right| \leq \frac{C_q}{|\xi|^q} (1 + \mu(B_{M+1})^q),$$

and the proposition is proved.

4 Sobolev norms—Divergence–Covariance matrix

4.1 Sobolev norms

We prove in this section that $\forall l \ge 1$ and $\forall p \ge 1 E |F_M|_l^p \le C_{l,p}$. We begin this section with a preliminary lemma which will be also useful to control the covariance matrix.

4.1.1 Preliminary

We consider a Poisson point measure N(ds, dz, du) on $\mathbb{R}^d \times \mathbb{R}_+$ with compensator $\mu(dz) \times 1_{(0,\infty)}(u)du$ and two non negative measurable functions $f, g : \mathbb{R}^d \to \mathbb{R}_+$. For a measurable set $B \subset \mathbb{R}^d$ we denote $B_g = \{(z, u) : z \in B, u < g(z)\} \subset \mathbb{R}^d \times \mathbb{R}_+$ and we consider the process

$$N_t(1_{B_g}f) := \int_0^t \int_{B_g} f(z)N(ds, dz, du).$$

Moreover we note $v_g(dz) = g(z)d\mu(z)$ and

$$\alpha_{g,f}(s) = \int_{\mathbb{R}^d} (1 - e^{-sf(z)}) d\nu_g(dz), \quad \beta_{B,g,f}(s) = \int_{B^c} (1 - e^{-sf(z)}) d\nu_g(dz).$$

We have the following result.

Lemma 6 Let $\phi(s) = Ee^{-sN_t(f_{1B_g})}$ the Laplace transform of the random variable $N_t(f_{1B_g})$ then we have

$$\phi(s) = e^{-t(\alpha_{g,f}(s) - \beta_{B,g,f}(s))}.$$

Proof From Itô's formula we have

$$\exp(-sN_t(f1_{B_g})) = 1 - \int_0^t \int_{\mathbb{R}^d \times \mathbb{R}_+} \exp(-s(N_{r-1}(f1_{B_g})))$$
$$\times (1 - \exp(-sf(z)1_{B_g}(z, u)))dN(r, z, u)$$

and consequently

$$E(\exp(-sN_t(f1_{B_g}))) = 1 - \int_0^t E(\exp(-s(N_{r-1}(f1_{B_g})))$$
$$\times \int_{\mathbb{R}^d \times \mathbb{R}_+} (1 - \exp(-sf(z)1_{B_g}(z, u)))d\mu(z)dudr.$$

But

$$\int_{\mathbb{R}^d \times \mathbb{R}_+} (1 - \exp(-sf(z)1_{B_g}(z, u)))d\mu(z)du$$

$$= \int_{\mathbb{R}^d \times \mathbb{R}_+} 1_{B_g}(z, u)(1 - \exp(-sf(z)))d\mu(z)du$$

$$= \int_{\mathbb{R}^d} 1_B(z)(1 - \exp(-sf(z))) \int_{\mathbb{R}_+} 1_{\{u < g(z)\}} dud\mu(z)$$

$$= \int_B (1 - \exp(-sf(z)))g(z)d\mu(z) = \alpha_{g,f}(s) - \beta_{B,g,f}(s).$$

It follows that

$$E(\exp(-sN_t(f1_{B_g}))) = \exp(-t(\alpha_{g,f}(s) - \beta_{B,g,f}(s))).$$

4.1.2 Bound for $|\overline{X}_t^M|_l$

In this section, we use the notation $\overline{c}_1(z) = \sup_x |\nabla_x c(z, x)|$. Under Hypothesis 3.3.i we have $\overline{c}_1(z) \leq \overline{c}(z)$, but we introduce this notation to highlight the dependence on the first derivative of the function *c*.

Lemma 7 Let (\overline{X}_t^M) the process solution of (38) then under Hypotheses 3.0., 3.1. and 3.2. we have $\forall l \ge 1$,

$$\sup_{s \le t} |\overline{X}_s^M|_{1,l} \le C_l \left(1 + \sum_{k=1}^{J_l^M} \overline{c}(\overline{Z}_k) \right)^{l \times l!} \sup_{s \le t} (\mathcal{E}_s^M)^{l \times l!}$$

where C_l is an universal constant and where \mathcal{E}_t^M is solution of the linear equation

$$\mathcal{E}_t^M = 1 + C_l \sum_{k=1}^{J_t^M} \overline{c}_1(\overline{Z}_k) \mathcal{E}_{T_k-}^M + C_l \int_0^t \mathcal{E}_s^M ds.$$
(43)

Consequently $\forall l, p \ge 1$

$$\sup_{M} E \sup_{s \le t} |\overline{X}_{s}^{M}|_{1,l}^{p} < \infty$$

. .

Before proving this lemma we first give a result which is a straightforward consequence of Lemma 1 and formula (21).

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Lemma 8 Let $\phi : \mathbb{R}^d \mapsto \mathbb{R} \ a \ \mathcal{C}^{\infty}$ function and $F \in S^d$ then $\forall l \ge 1$ we have

$$|\phi(F)|_{1,l} \le |\nabla \phi(F)||F|_{1,l} + C_l \sup_{2 \le |\beta| \le l} |\partial_{\beta} \phi(F)||F|_{1,l-1}^l.$$

We proceed now to the proof of Lemma 7.

Proof We first recall that from Hypothesis 3.0.g and its derivatives are bounded and from Hypothesis 3.2.i) the coefficient c as well as its derivatives are bounded by the function \overline{c} . Now the truncated coefficient c_M of (38) is equal to $c_M = c \times \phi_M$ where ϕ_M is a C^{∞} bounded function with derivatives uniformly bounded with respect to M. Consequently using Lemma 8 we obtain for $l \ge 1$

$$|\overline{X}_t^M|_{1,l} \le C_l \left(A_{t,l-1} + \sum_{k=1}^{J_t^M} \overline{c}_1(\overline{Z}_k) |\overline{X}_{T_k-}^M|_{1,l} + \int_0^t |\overline{X}_s^M|_{1,l} ds \right),$$

with

$$A_{t,l-1} = \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k)(|\overline{Z}_k|_{1,l} + |\overline{Z}_k|_{1,l-1}^l + |\overline{X}_{T_k}^M|_{1,l-1}^l) + \int_0^t |\overline{X}_s^M|_{1,l-1}^l ds.$$

. .

This gives

$$\forall s \le t \quad |\overline{X}_{s}^{M}|_{1,l} \le A_{t,l-1} \mathcal{E}_{s}^{M}, \tag{44}$$

Under Hypotheses 3.0. 3.1. and 3.2. we have

$$\forall p \ge 1 \quad E\left(\sup_{s \le t} |\mathcal{E}_t^M|^p\right) \le C_p.$$

Now one can easily check that for $l \ge 1$

$$|\overline{Z}_k|_{1,l} \le |\pi_k|_{l-1},$$

but since $\pi_k = \phi_M(\overline{Z}_k)$ we deduce from Lemma 8 that

$$|\overline{Z}_k|_{1,l} \le 1 + C_l(|\overline{Z}_k|_{1,l-1} + |\overline{Z}_k|_{1,l-2}^{l-1}).$$

Observing that $|\overline{Z}_k|_{1,1} = |D\overline{Z}_k| = |\pi_k| \le 1$ we conclude that $\forall l \ge 1$

$$|\overline{Z}_k|_{1,l} \le C_l.$$

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This gives

$$A_{t,l-1} \le tC_l \left(1 + \sup_{s \le t} |\overline{X}_s^M|_{1,l-1} \right)^l \left(1 + \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k) \right).$$

$$(45)$$

From this inequality we can prove easily Lemma 7 by induction. For l = 1 we remark that

$$\forall s \leq t \quad |\overline{X}_s^M|_{1,1} \leq A_{t,0} \mathcal{E}_s^M, \quad \text{with} \quad A_{t,0} = \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k),$$

and the result is true. To complete the proof of Lemma 7, we prove that $\forall p \ge 1$

$$E\left(\sum_{k=1}^{J_l^M} \overline{c}(\overline{Z}_k)\right)^p \le C_p.$$

We have the equality in law

$$\sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k) \simeq \int_0^t \int_E \overline{c}(z) \mathbf{1}_{u < \gamma(z, X_{T_k-}^M)} \mathbf{1}_{B_{M+1}}(z) \mathbf{1}_{[0, 2\overline{C}]}(u) N(ds, dz, du),$$

moreover using the notations of Sect. 4.1.1. we have

$$\int_{0}^{t} \int_{E} \overline{c}(z) \mathbf{1}_{u < \gamma(z, X_{T_{k}^{-}}^{M})} \mathbf{1}_{B_{M+1}}(z) \mathbf{1}_{[0, 2\overline{C}]}(u) N(ds, dz, du) \leq N_{t}(\mathbf{1}_{B_{\overline{\gamma}}}\overline{c})$$

with $B_{\overline{\gamma}} = \{(z, u); z \in B_{M+1}; 0 < u < \overline{\gamma}(z)\}$. From Lemma 6 it follows that

$$Ee^{-sN_t(1_{B_{\overline{\gamma}}}\overline{c})} = \exp\left(-t\int\limits_{B_{M+1}} (1-e^{-s\overline{c}(z)})\overline{\gamma}(z)d\mu(z)\right)$$

and since from Hypotheses 3.1. and 3.2., $\int_{\mathbb{R}^d} |\overline{c}(z)\overline{\gamma}(z)|d\mu(z) < \infty$ we deduce that $\forall p \ge 1$, $EN_t(1_{B_{\overline{\gamma}}}\overline{c})^p = t^p(\int_{B_{M+1}} |\overline{c}(z)\overline{\gamma}(z)|d\mu(z))^p \le C_p$ where the constant C_p does not depend on M. This achieves the proof of Lemma 7.

4.1.3 Bound for $|F_M|_l$

Lemma 9 Under Hypotheses 3.0., 3.1. and 3.2. we have

$$\forall l, p \ge 1 \quad E|F_M|_l^p \le C_{l,p}.$$

We have $F_M = \overline{X}_t^M + \sqrt{U_M(t)}\Delta$ and then $|F_M|_l \leq |\overline{X}_t^M|_l + \sqrt{U_M(t)}|\Delta|_l$. But $|\Delta|_l \leq |\Delta| + d$ and $U_M(t) \leq t \int_{\mathbb{R}^d} \overline{c}^2(z)\overline{\gamma}(z)d\mu(z) < \infty$. So the conclusion of Lemma 9 follows from Lemma 7.

4.2 Divergence

In this section our goal is to bound $|LF_M|_l$ for $l \ge 0$. From the definition of the divergence operator L we have $LF_M^r = L\overline{X}_t^{M,r} - \Delta^r$ and then

$$|LF_M|_l \le |L\overline{X}_t^M|_l + |\Delta| + d,$$

so we just have to bound $|L\overline{X}_{t}^{M}|_{l}$. We proceed as in the previous section and we first state a lemma similar to Lemma 8.

Lemma 10 Let $\phi : \mathbb{R}^d \mapsto \mathbb{R} \ a \ \mathcal{C}^{\infty}$ function and $F \in S^d$ then $\forall l \ge 1$ we have

$$\begin{split} |L\phi(F)|_{l} &\leq |\nabla\phi(F)||LF|_{l} + C_{l} \sup_{2 \leq |\beta| \leq l+2} |\partial_{\beta}\phi(F)|(1+|F|_{l}^{l})(|LF|_{l-1}+|F|_{1,l+1}^{2}), \\ &\leq |\nabla\phi(F)||LF|_{l} + C_{l} \sup_{2 \leq |\beta| \leq l+2} |\partial_{\beta}\phi(F)|(1+|F|_{l+1}^{l+2})(1+|LF|_{l-1}). \end{split}$$

For l = 0, we have

$$|L\phi(F)| \le \nabla\phi(F)||LF| + \sup_{\beta=2} |\partial_{\beta}\phi F||F|_{1,1}^2.$$

The proof follows from (7) and Lemma 8 and we omit it.

Next we give a bound for $|L\overline{Z}_k|_l$. We recall the notation

$$\overline{\gamma}_{\ln}^{z,l}(z) = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,z} \ln \gamma(z, x)|, \quad \overline{h}_{\ln}^{l}(z) = \sup_{1 \le |\beta| \le l} |\partial_{\beta} \ln h(z)|,$$
$$\overline{\theta}_{\ln}^{l} = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta} \ln \theta_{M,\gamma}(x)|,$$
$$\overline{\gamma}_{\ln}^{x,l}(z) = \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,x} \ln \gamma(z, x)|, \quad \overline{\gamma}^{x,l} = \sup_{z} \sup_{x} \sup_{1 \le |\beta| \le l} |\partial_{\beta,x} \gamma(z, x)|$$

Lemma 11 Assuming Hypotheses 3.0., 3.1., 3.2 and 3.3., we have $\forall l \ge 0$ and $\forall k \le J_t^M$

$$|L\overline{Z}_{k}|_{l} \leq C_{l} \left(\overline{\gamma}_{\ln}^{z,l+1}(\overline{Z}_{k}) + \overline{h}_{\ln}^{z,l+1}(\overline{Z}_{k}) + \sup_{s \leq t} |\overline{X}_{s}^{M}|_{l+1}^{l+1} \right)$$
$$\times \sum_{j=k+1}^{J_{l}^{M}} \overline{\theta}_{\ln}^{l+1} \mathbf{1}_{B(z_{M}^{*},1)}(\overline{Z}_{j}) + \overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_{j})) ,$$

with $\overline{\theta}_{\ln}^l \leq C_l(\overline{\gamma}^{x,l})^l$.

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In addition, if we assume 3.3.a., we obtain $\forall p \ge 1$

$$E \sup_{k \le J_t^M} |L\overline{Z}_k|_l^p \le C_{p,l}.$$

On the other hand, assuming 3.3*.b, we have* $\forall p \geq 1$

$$E \sup_{k \le J_l^M} |L\overline{Z}_k|_l^p \le C_{p,l} (1 + \mu (B_{M+1})^p)$$

Proof We first recall that we have proved in the preceding section that $\forall l \ge 1, |\overline{Z}_k|_l \le C_l$. Now $L\overline{Z}_k^r = \delta(D\overline{Z}_k^r)$ and since $D_{k,r}\overline{Z}_k^r = \pi_k$ we obtain

$$L\overline{Z}_k^r = -\partial_{k,r}(\pi_k^2) - \pi_k D_{k,r} \ln p_M,$$

this leads to

$$|L\overline{Z}'_k|_l \le C_l(1+|D_{k,r}\ln p_M|_l).$$

Recalling that $\ln p_M = \sum_{j=1}^{J_t^M} \ln q_M(\overline{Z}_j, \overline{X}_{T_j-}^M)$ and that $\overline{X}_{T_j-}^M$ depends on \overline{Z}_k for $k \leq j-1$ we obtain

$$D_{k,r} \ln p_M = D_{k,r} \ln q_M(\overline{Z}_k, \overline{X}_{T_k-}^M) + \sum_{j=k+1}^{J_t^M} D_{k,r} \ln q_M(\overline{Z}_j, \overline{X}_{T_j-}^M)$$

But on $\{\pi_k > 0\}$, we have $q_M(\overline{Z}_k, \overline{X}_{T_k-}^M) = C\gamma(\overline{Z}_k, \overline{X}_{T_k-}^M)h(\overline{Z}_k)$, and then

$$D_{k,r} \ln q_M(\overline{Z}_k, \overline{X}_{T_k-}^M) = D_{k,r} \ln \gamma(\overline{Z}_k, \overline{X}_{T_k-}^M) + D_{k,r} \ln h(\overline{Z}_k).$$

Now for $j \ge k + 1$, if $|\overline{Z}_j - z_M^*| < 1$ then

$$\ln q_M(\overline{Z}_j, \overline{X}_{T_j-}^M) = \ln \varphi(\overline{Z}_j - z_M^*) + \ln \theta_{M,\gamma}(\overline{X}_{T_j-}^M)$$

consequently

$$D_{k,r}\ln q_M(\overline{Z}_j,\overline{X}^M_{T_j-}) = D_{k,r}\ln \theta_{M,\gamma}(\overline{X}^M_{T_j-}),$$

and if $\overline{Z}_j \in B_{M+1}$ then

$$D_{k,r} \ln q_M(\overline{Z}_j, \overline{X}_{T_j-}^M) = D_{k,r} \ln \gamma(\overline{Z}_j, \overline{X}_{T_j-}^M)$$

and finally

$$D_{k,r} \ln q_M(\overline{Z}_j, \overline{X}_{T_j-}^M) = D_{k,r} \ln \theta_{M,\gamma}(\overline{X}_{T_j-}^M) \mathbf{1}_{B(z_M^*, 1)}(\overline{Z}_j) + D_{k,r} \ln \gamma(\overline{Z}_j, \overline{X}_{T_j-}^M) \mathbf{1}_{B_{M+1}}(\overline{Z}_j).$$

It is worth to note that this random variable is a simple variable as defined in section 2. Putting this together, it yields

$$\begin{aligned} |D_{k,r} \ln p_M|_l &\leq |D_{k,r} \ln \gamma(\overline{Z}_k, \overline{X}_{T_k-}^M)|_l + |D_{k,r} \ln h(\overline{Z}_k)|_l \\ &+ \sum_{j=k+1}^{J_l^M} \left(\left| D_{k,r} \ln \theta_{M,\gamma}(\overline{X}_{T_j-}^M) \mathbf{1}_{B(z_M^*,1)}(\overline{Z}_j) \right|_l \right. \\ &+ |D_{k,r} \ln \gamma(\overline{Z}_j, \overline{X}_{T_j-}^M)|_l \right). \end{aligned}$$

Applying Lemma 8, this gives

$$\begin{split} |D_{k,r}\ln p_{M}|_{l} &\leq (\overline{\gamma}_{\ln}^{z,l+1}(\overline{Z}_{k}) + \overline{h}_{\ln}^{z,l+1}(\overline{Z}_{k}))|\overline{Z}_{k}|_{1,l+1}^{l+1} + \sum_{j=k+1}^{J_{l}^{M}} \left(\overline{\theta}_{\ln}^{l+1} \mathbb{1}_{B(z_{M}^{*},1)}(\overline{Z}_{j}) + \overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_{j})\right)|\overline{X}_{T_{j}}^{M})|_{1,l+1}^{l+1}. \end{split}$$

We obtain then, for $k \leq J_t^M$

$$\begin{split} |L\overline{Z}_k|_l &\leq C_l(\overline{\gamma}_{\ln}^{z,l+1}(\overline{Z}_k) + \overline{h}_{\ln}^{z,l+1}(\overline{Z}_k) + \sup_{s \leq t} |\overline{X}_s^M|_{l+1}^{l+1} \sum_{j=k+1}^{J_l^M} (\overline{\theta}_{\ln}^{l+1} \mathbf{1}_{B(z_M^*,1)}(\overline{Z}_j) \\ &+ \overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_j))). \end{split}$$

Now from the definition of $\theta_{M,\gamma}$, we have

$$\partial_{\beta}\theta_{M,\gamma}(x) = -\frac{1}{2\overline{C}\mu(B_{M+1})} \int\limits_{B_{M+1}} \partial_{\beta,x}\gamma(z,x)d\mu(z).$$

Then assuming 3.3. and recalling that $1/2 \le \theta_{M,\gamma}(x) \le 1$, we obtain

$$\overline{\theta}_{\ln}^{l} \leq C_{l} (\overline{\gamma}^{x,l})^{l}$$

this finally gives

$$|L\overline{Z}_{k}|_{l} \leq C_{l}(\overline{\gamma}_{\ln}^{z,l+1}(\overline{Z}_{k}) + \overline{h}_{\ln}^{z,l+1}(\overline{Z}_{k}) + \sup_{s \leq t} |\overline{X}_{s}^{M}|_{l+1}^{l+1} \sum_{j=k+1}^{J_{t}^{M}} ((\overline{\gamma}^{x,l+1})^{l+1} \mathbb{1}_{B(z_{M}^{*},1)}(\overline{Z}_{j}) + \overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_{j}))).$$

The first part of Lemma 11 is proved. Moreover, we can check that from 3.3, we have $\forall p \ge 1$

$$E\left(\sum_{j=1}^{J_t^M} \mathbb{1}_{B(z_M^*,1)}(\overline{Z}_j)\right)^p \le t^p \sup_{z^*} \left(\int_{B(z^*,1)} \overline{\gamma}(z) d\mu(z)\right)^p < \infty.$$

Now assuming 3.3.*a*, we have $\forall p \ge 1$

$$E\left(\sum_{j=1}^{J_i^M}\overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_j)\right)^p \le t^p \left(\int\overline{\gamma}_{\ln}^{x,l+1}(z)\overline{\gamma}(z)d\mu(z)\right)^p < \infty,$$

then the second part of Lemma 11 follows from Lemma 7 and Cauchy–Schwarz inequality. At last, assuming 3.3.*b*, we check that $\sum_{j=1}^{J_t^M} \overline{\gamma}_{\ln}^{x,l+1}(\overline{Z}_j) \leq \overline{\gamma}_{\ln}^{x,l+1}J_t^M$, and the third part follows easily.

We can now state the main lemma of this section.

Lemma 12 Assuming Hypotheses 3.0., 3.1. and 3.2., we have $\forall l \geq 0$

$$\sup_{s\leq t}|L\overline{X}_{s}^{M}|_{l}\leq B_{t,l}^{M}\left(1+\sup_{k\leq J_{t}^{M}}|L\overline{Z}_{k}|_{l}\right),$$

where $B_{t,l}^M$ is a random variable such that $\forall p \ge 1$, $E(B_{t,l}^M)^p \le C_p$ for a constant C_p independent on M. More precisely we have

$$B_{t,l}^{M} \leq C_l \left(1 + \sum_{k=1}^{J_t^{M}} \overline{c}(\overline{Z}_k)\right)^{l+1} \left(1 + \sup_{s \leq t} |\overline{X}_s^{M}|_{l+1}^{l+2}\right)^{l+1} \sup_{s \leq t} (\mathcal{E}_s^{M})^{l+1}$$

where \mathcal{E}_s is solution of (43).

Proof We proceed by induction. From (38) we have

$$L\overline{X}_{t}^{M} = \sum_{k=1}^{J_{t}^{M}} Lc_{M}(\overline{Z}_{k}, \overline{X}_{T_{k}-}^{M}) + \int_{0}^{t} Lg(\overline{X}_{s}^{M}) ds.$$

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For l = 0, the second part of Lemma 10 gives

$$|L\overline{X}_{t}^{M}| \leq B_{t,0} + C\left(\sum_{k=1}^{J_{t}^{M}} \overline{c}_{1}(\overline{Z}_{k})|L\overline{X}_{T_{k}-}^{M}| + \int_{0}^{t} |L\overline{X}_{s}^{M}|ds\right)$$

with

$$B_{t,0} = C\left(\sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k)(|L\overline{Z}_k| + |\overline{Z}_k|_{1,1}^2 + |\overline{X}_{T_k}^M|_{1,1}^2) + \int_0^t |\overline{X}_s^M|_{1,1}^2 ds\right).$$

This gives

$$\forall s \leq t, \quad |L\overline{X}_s^M| \leq B_{t,0}\mathcal{E}_s^M,$$

where \mathcal{E}_{s}^{M} is solution of (43) and

$$B_{t,0} \leq C\left(1 + \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k)\right) \left(1 + \sup_{s \leq t} |\overline{X}_s^M|_1^2\right) \left(1 + \sup_{k \leq J_t^M} |L\overline{Z}_k|\right).$$

Consequently Lemma 12 is proved for l = 0.

For l > 0, we obtain similarly from Lemma 10

$$|L\overline{X}_{t}^{M}|_{l} \leq B_{t,l-1} + C_{l} \left(\sum_{k=1}^{J_{t}^{M}} \overline{c}_{1}(\overline{Z}_{k}) |L\overline{X}_{T_{k}-}^{M}|_{l} + \int_{0}^{t} |L\overline{X}_{s}^{M}|_{l} ds \right)$$

with

$$B_{t,l-1} = C_l \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k) (|L\overline{Z}_k|_l + 1 + |L\overline{X}_{T_k-}^M|_{l-1}) (1 + |\overline{Z}_k|_{l+1}^{l+2} + |\overline{X}_{T_k-}^M|_{l+1}^{l+2}) + C_l \int_0^t (1 + |L\overline{X}_{T_k-}^M|_{l-1}) (1 + |\overline{X}_s^M|_{l+1}^{l+2}) ds.$$

We deduce then that

$$B_{t,l-1} \leq C_l \left(1 + \sup_{s \leq t} |L\overline{X}_s^M|_{l-1} \right) \left(1 + \sup_{s \leq t} |\overline{X}_s^M|_{l+1}^{l+2} \right) \left(1 + \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k) \right)$$
$$+ C_l \sup_{k \leq J_t^M} |L\overline{Z}_k|_l \left(1 + \sup_{s \leq t} |\overline{X}_s^M|_{l+1}^{l+2} \right) \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k),$$

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now from the induction hypothesis, we have

$$B_{t,l-1} \leq C_l \left(1 + \sup_{s \leq t} |\overline{X}_s^M|_{l+1}^{l+2} \right)^{l+1} \left(1 + \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k) \right)^{l+1} \sup_{s \leq t} (\mathcal{E}_s^M)^l \left(1 + \sup_{k \leq J_t^M} |L\overline{Z}_k|_{l-1} \right) \\ + C_l \sup_{k \leq J_t^M} |L\overline{Z}_k|_l \left(1 + \sup_{s \leq t} |\overline{X}_s^M|_{l+1}^{l+2} \right) \sum_{k=1}^{J_t^M} \overline{c}(\overline{Z}_k),$$

this leads to

$$\forall s \leq t \quad |L\overline{X}_s^M|_l \leq B_{t,l}^M \left(1 + \sup_{k \leq J_t^M} |L\overline{Z}_k|_l\right),$$

with

$$B_{t,l}^{M} \leq C_{l}(1 + \sup_{s \leq t} |\overline{X}_{s}^{M}|_{l+1}^{l+2})^{l+1} \left(1 + \sum_{k=1}^{J_{t}^{M}} \overline{c}(\overline{Z}_{k})\right)^{l+1} \sup_{s \leq t} (\mathcal{E}_{s}^{M})^{l+1}.$$

From Lemma 7, we observe that $E(B_{t,l}^M)^p < C_p$.

Finally recalling that

$$|LF_M|_l \le |L\overline{X}_t^M|_l + |\Delta| + d$$

and combining Lemmas 7, 11 and 12 we deduce easily the following lemma.

Lemma 13 Assuming Hypotheses 3.0., 3.1. and 3.2., we have $\forall l, p \ge 1$

- (a) *if* 3.3.*a* holds, $E|LF_M|_l^p \le C_{l,p}$; (b) *if* 3.3.*b* holds, $E|LF_M|_l^p \le C_{l,p}(1 + \mu(B_{M+1})^p)$.

4.3 The covariance matrix

4.3.1 Preliminaries

We consider an abstract measurable space E, a measure ν on this space and a non negative measurable function $f: E \to \mathbb{R}_+$ such that $\int f d\nu < \infty$. For t > 0 and $p \ge 1$ we note

$$\alpha_f(t) = \int_E (1 - e^{-tf(a)}) d\nu(a)$$
 and $I_t^p(f) = \int_0^\infty s^{p-1} e^{-t\alpha_f(s)} ds.$

Lemma 14 (i) Suppose that for $p \ge 1$ and t > 0

$$\underline{\lim}_{u \to \infty} \frac{1}{\ln u} \alpha_f(u) > p/t \tag{46}$$

then $I_t^p(f) < \infty$. (ii) A sufficient condition for (46) is

$$\underline{\lim}_{u \to \infty} \frac{1}{\ln u} \nu \left(f \ge \frac{1}{u} \right) > p/t.$$
(47)

In particular, if $\underline{\lim}_{u\to\infty} \frac{1}{\ln u} \nu(f \ge \frac{1}{u}) = \infty$ then $\forall p \ge 1$ and $\forall t > 0$, $I_t^p(f) < +\infty$.

We remark that if v is finite then (47) can not be satisfied.

Proof (i) From (46) one can find $\varepsilon > 0$ such that as *s* goes to infinity $s^{p-1}e^{-t\alpha_f(s)} \le 1/s^{1+\varepsilon}$ and consequently $I_t^p(f) < \infty$.

(ii) With the notation $n(dz) = v \circ f^{-1}(dz)$ we have

$$\alpha_f(u) = \int_0^\infty (1 - e^{-uz}) dn(z) = \int_0^\infty e^{-y} n\left(\frac{y}{u}, \infty\right) dy.$$

Using Fatou's lemma and (47) we obtain

$$\underline{\lim}_{u\to\infty}\frac{1}{\ln u}\int_{0}^{\infty}e^{-y}n\left(\frac{y}{u},\infty\right)dy\geq\int_{0}^{\infty}e^{-y}\underline{\lim}_{u\to\infty}\frac{1}{\ln u}n\left(\frac{y}{u},\infty\right)dy>p/t.$$

We come now back to the framework of Sect. 4.1.1 and we consider the Poisson point measure N(ds, dz, du) on $\mathbb{R}^d \times \mathbb{R}_+$ with compensator $\mu(dz) \times 1_{(0,\infty)}(u)du$. We recall that

$$N_t(1_{B_g}f) := \int_0^t \int_{B_g} f(z)N(ds, dz, du),$$

for $f, g: \mathbb{R}^d \to \mathbb{R}_+$ and $B_g = \{(z, u) : z \in B, u < g(z)\} \subset \mathbb{R}^d \times \mathbb{R}_+$ and that

$$\alpha_{g,f}(s) = \int_{\mathbb{R}^d} (1 - e^{-sf(z)}) d\nu_g(dz), \quad \beta_{B,g,f}(s) = \int_{B^c} (1 - e^{-sf(z)}) d\nu_g(dz).$$

We have the following result.

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Lemma 15 Let $U_t = t \int_{B^c} f(z) dv_g(z)$, then $\forall p \ge 1$

$$E\left(\frac{1}{(N_t(1_{B_g}f)+U_t)^p}\right) \le \frac{1}{\Gamma(p)} \int_0^\infty s^{p-1} \exp(-t\alpha_{g,f}(s)) ds = \frac{1}{\Gamma(p)} I_t^p(f).$$
(48)

Suppose moreover that for some $0 < \theta \leq \infty$

$$\underline{\lim}_{a \to \infty} \frac{1}{\ln a} \nu_g \left(f \ge \frac{1}{a} \right) = \theta, \tag{49}$$

then for every t > 0 and $p \ge 1$ such that $p/t < \theta$

$$E\left(\frac{1}{(N_t(1_{B_g}f)+U_t)^p}\right)<\infty.$$

Observe that if $\nu(B) < \infty$ then $E \frac{1}{(N_t (1_{B_g} f)^p)} = \infty$

Proof By a change of variables we obtain for every $\lambda > 0$

$$\lambda^{-p}\Gamma(p) = \int_{0}^{\infty} s^{p-1} e^{-\lambda s} ds.$$

Taking the expectation in the previous equality with $\lambda = N_t(f \mathbf{1}_{B_g}) + U_t$ we obtain

$$E(\frac{1}{(N_t(f1_{B_g})+U_t)^p}) = \frac{1}{\Gamma(p)} \int_0^\infty s^{p-1} E(\exp(-s(N_t(f1_{B_g})+U_t))ds.$$

Now from Lemma 6 we have

$$E(\exp(-sN_t(f1_{B_g}))) = \exp(-t(\alpha_{g,f}(s) - \beta_{B,g,f}(s))).$$

Moreover, from the definition of U_t one can easily check that $\exp(-sU_t) \leq \exp(-t\beta_{B,g,f}(s))$ and then

$$E(\exp(-s(N_t(f1_{B_q}) + U_t)) \le \exp(-t\alpha_{g,f}(s))$$

this achieves the proof of (48). The second part of the lemma follows directly from Lemma 14. $\hfill \Box$

4.3.2 The Malliavin covariance matrix

In this section, we prove that under some additional assumptions on p and t, $E(\det \sigma (F_M))^{-p} \leq C_p$, for the Malliavin covariance matrix $\sigma(F_M)$ defined in Sect. 3.4.

We first remark that from Hypothesis 3.2 ii) the tangent flow of (38) is invertible and that the moments of all order of this inverse are finite. More precisely we define Y_t^M , $t \ge 0$ and \hat{Y}_t^M , $t \ge 0$ as the matrix solutions of the equations

$$Y_t^M = I + \sum_{k=1}^{J_t^M} \nabla_x c_M(\overline{Z}_k, \overline{X}_{T_k-}^M) Y_{T_k-}^M + \int_0^t \nabla_x g(\overline{X}_s^M) Y_s^M ds,$$
(50)

$$\widehat{Y}_t^M = I - \sum_{k=1}^{J_t^M} \nabla_x c_M (I + \nabla_x c_M)^{-1} (\overline{Z}_k, \overline{X}_{T_k-}^M) \widehat{Y}_{T_k-}^M - \int_0^t \nabla_x g(\overline{X}_s^M) \widehat{Y}_s^M ds.$$
(51)

Then $\widehat{Y}_t^M \times Y_t^M = I$, $\forall t \ge 0$. Moreover we can prove under Hypotheses 3.0, 3.1 and 3.2. that $\forall p \ge 1$

$$E\left(\sup_{s\leq t}\left(\left\|\widehat{Y}_{s}^{M}\right\|^{p}+\left\|Y_{s}^{M}\right\|^{p}\right)\right)\leq K_{p}<\infty$$
(52)

where K_p is a constant.

Lemma 16 Assuming Hypothesis 3.0, 3.1, 3.2 we have for $p \ge 1$, t > 0 such that $2dp/t < \theta$

$$E\left(\frac{1}{(\det\sigma(F_M))^p}\right) \le C_p,\tag{53}$$

where the constant C_p does not depend on M.

Proof We first give a lower bound for the lowest eigenvalue of the matrix $\sigma(\overline{X}_t^M)$.

$$\rho_t := \inf_{|\xi|=1} \left\langle \sigma(\overline{X}_t^M) \xi, \xi \right\rangle = \inf_{|\xi|=1} \sum_{k=1}^{J_t^M} \sum_{r=1}^d \left\langle D_{k,r} \overline{X}_t^M, \xi \right\rangle^2.$$

But from (38) we have

$$D_{k,r}\overline{X}_{t}^{M} = \sum_{k'=1}^{J_{t}^{M}} \nabla_{z}c_{M}\left(\overline{Z}_{k'}, \overline{X}_{T_{k'}}^{M}\right) D_{k,r}\overline{Z}_{k'} + \sum_{k'=1}^{J_{t}^{M}} \nabla_{x}c_{M}\left(\overline{Z}_{k'}, \overline{X}_{T_{k'}}^{M}\right) D_{k,r}\overline{X}_{T_{k'}}^{M} + \int_{0}^{t} \nabla_{x}g\left(\overline{X}_{s}^{M}\right) D_{k,r}\overline{X}_{s}^{M} ds$$

where $\nabla_z c_M = (\partial_{z_r} c_M^{r'})_{r',r}$ and $\nabla_x c_M = (\partial_{x_r} c_M^{r'})_{r',r}$. Since $D_{k,r} \overline{Z}_{k'} = 0$ for $k \neq k'$ we obtain

$$D_{k,r}\overline{X}_{t}^{M,r'} = \left(Y_{t}^{M}\nabla_{z}c_{M}\left(\overline{Z}_{k},\overline{X}_{T_{k}}^{M}\right)D_{k,r}\overline{Z}_{k}\right)_{r',r} = \pi_{k}\left(Y_{t}^{M}\nabla_{z}c_{M}\left(\overline{Z}_{k},\overline{X}_{T_{k}}^{M}\right)\right)_{r',r}$$

We deduce that

$$\sum_{r=1}^{d} \left\langle D_{k,r} \overline{X}_{t}^{M}, \xi \right\rangle^{2} = \sum_{r=1}^{d} \pi_{k}^{2} \left\langle \partial_{z^{r}} c_{M} (\overline{Z}_{k}, \overline{X}_{T_{k}}^{M}), (Y_{t}^{M})^{*} \xi \right\rangle^{2},$$

but recalling that $\pi_k \ge 1_{B_{M-1}}(\overline{Z}_k)$ and $c_M = c$ on B_{M-1} we obtain

$$\sum_{r=1}^{d} \left\langle D_{k,r} \overline{X}_{t}^{M}, \xi \right\rangle^{2} \geq \sum_{r=1}^{d} \mathbb{1}_{B_{M-1}}(\overline{Z}_{k}) \left\langle \partial_{z^{r}} c(\overline{Z}_{k}, \overline{X}_{T_{k}}^{M}), (Y_{t}^{M})^{*} \xi \right\rangle^{2},$$

and consequently using Hypothesis 3.2. (iii)

$$\rho_t \geq \inf_{|\xi|=1} \sum_{k=1}^{J_t^M} \mathbf{1}_{B_{M-1}}(\overline{Z}_k) \underline{c}^2(\overline{Z}_k) |(Y_t^M)^* \xi|^2 \geq \left\| \widehat{Y}_t^M \right\|^{-2} \sum_{k=1}^{J_t^M} \mathbf{1}_{B_{M-1}}(\overline{Z}_k) \underline{c}^2(\overline{Z}_k).$$

Now since $\sigma(F_M) = \sigma(\overline{X}_t^M) + U_M(t)$ we have

$$E\left|\frac{1}{\det\sigma(F_M)}\right|^p \le E\left|\frac{1}{\rho_t + U_M(t)}\right|^{dp} \le E\left(\frac{1 + \|\widehat{Y}_t^M\|^2}{\sum_{k=1}^{J_t^M} \mathbb{1}_{B_{M-1}}(\overline{Z}_k)\underline{c}^2(\overline{Z}_k) + U_M(t)}\right)^{dp}$$

Now observe that the denominator of the last fraction is equal in law to

$$\sum_{k=1}^{J_t^M} \mathbf{1}_{B_{M-1}}(Z_k)\underline{c}^2(Z_k)\mathbf{1}_{U_k < \gamma(Z_k, X_{T_k^{-}}^M)} + U_M(t) \ge N_t(\mathbf{1}_{B_{\underline{\gamma}}^M}\underline{c}^2) + U_M(t),$$

with $B_{\underline{\gamma}}^{M} = \{(z, u); z \in B_{M-1}; 0 < u < \underline{\gamma}(z)\}$. Assuming Hypothesis 3.2.iii, we can apply Lemma 15 with $f = \underline{c}^{2}$ and $d\nu(z) = \underline{\gamma}(z)d\mu(z)$. This gives for $p' \ge 1$ such that $p'/t < \theta$

$$E\left(\frac{1}{N_t(1_{B_{\underline{\gamma}}^M}\underline{c}^2)+U_M(t)}\right)^{p'} \le C_{p'}.$$

Finally since the moments of $\|\widehat{Y}_t^M\|$ are bounded uniformly on *M* the result of Lemma 16 follows from Cauchy–Schwarz inequality.

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