

Dirichlet Forms associated to Poisson Measures and Lévy Processes: The Lent Particle Method

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University of Osaka, January 2010

Based on joint works with N. Bouleau.

PART III: Application of the lent particle to prove existence of density of Poisson functionals.

In this part, we shall:

- ▶ Give some first examples;
- ▶ Apply the lent particle method to the case of sde's driven by Lévy measure.

Adaptation of the lent particle method to the case of Lévy processes

If (Y_t) is a \mathbb{R}^d -valued centered Lévy process without gaussian part, we have the representation

$$Y_t = \int_0^t \int_{\mathbb{R}^d} x \tilde{N}(ds, dx).$$

\Leftrightarrow we have to consider time-space Poisson measures.

So, we consider:

- N : a Poisson random measure on $[0, +\infty[\times X$ with intensity $dt \times \nu(du)$ defined on the probability space $(\Omega_1, \mathcal{A}_1, \mathbb{P}_1)$ where Ω_1 is the **configuration space**, \mathcal{A}_1 the σ -field generated by N and \mathbb{P}_1 the law of N .
- $(X, \mathcal{X}, \nu, \mathbf{d}, \gamma)$: a local symmetric Dirichlet structure which admits a carré du champ operator and satisfies property (EID).

The lent particle method in this setting

The Dirichlet form we consider on the bottom space $[0, +\infty[\times X$ is the product form:

$$(L^2(dt), 0) \otimes (\mathbf{d}, e)$$

which satisfies (EID).

Operators:

$$\begin{aligned}\varepsilon_{(t,u)}^+(w_1) &= w_1 \mathbf{1}_{\{(t,u) \in \text{supp } w_1\}} + (w_1 + \varepsilon_{(t,u)}) \mathbf{1}_{\{(t,u) \notin \text{supp } w_1\}} \\ \varepsilon_{(t,u)}^-(w_1) &= w_1 \mathbf{1}_{\{(t,u) \notin \text{supp } w_1\}} + (w_1 - \varepsilon_{(t,u)}) \mathbf{1}_{\{(t,u) \in \text{supp } w_1\}}.\end{aligned}$$

$$\varepsilon^+ H(w_1, t, u) = H(\varepsilon_{(t,u)}^+ w_1, t, u) \quad \varepsilon^- H(w_1, t, u) = H(\varepsilon_{(t,u)}^- w_1, t, u).$$

Upper Dirichlet structure

Following our construction we obtain a Dirichlet form $(\mathbb{D}, \mathcal{E})$ on $L^2(\Omega_1)$ which carré du champ Γ , gradient operator that we \sharp and we know that for $F \in \mathbb{D}$:

$$\blacktriangleright F^\sharp = \int_0^{+\infty} \int_{X \times R} \varepsilon^-((\varepsilon^+ F(t, \cdot))^b(u, r)) dN \odot \rho(dt, du, dr) \in L^2(\mathbb{P}_1 \times \hat{\mathbb{P}}).$$

$$\blacktriangleright \forall F \in \mathbb{D}, \quad \Gamma[F] = \hat{\mathbb{E}}(F^\sharp)^2 = \int_0^{+\infty} \int_X \varepsilon^-(\gamma[\varepsilon^+ F]) dN.$$

$\blacktriangleright (\mathbb{D}, \mathcal{E}, \Gamma)$ satisfies (EID).

Let us recall the example

$(X, \mathcal{X}) = (\mathbb{R}^r, \mathcal{B}(\mathbb{R}^r))$, $\nu = kdx$, $\xi = (\xi_{ij})_{1 \leq i, j \leq r}$. We assume that there exists an open set $O \subset \mathbb{R}^r$ and a function ψ continuous on O and null on $\mathbb{R}^r \setminus O$ such that

1. $k > 0$ on O ν -a.e. and is locally bounded on O
2. ξ is locally bounded and locally elliptic on O .
3. $k \geq \psi > 0$ ν -a.e. on O .
4. for all $i, j \in \{1, \dots, r\}$, $\xi_{i,j}\psi$ belongs to $H_{loc}^1(O)$.

(**d**, **e**), the local Dirichlet form which satisfies (EID):

$$\forall f, g \in H, e(f, g) = \sum_{i,j=1}^r \int_O \xi_{i,j}(x) \partial_i f(x) \partial_j g(x) \psi(x) dx$$

$$\forall f \in \mathbf{d}, \gamma(f)(x) = \sum_{i,j=1}^r \xi_{i,j}(x) \partial_i f(x) \partial_j f(x) \frac{\psi(x)}{k(x)}.$$



Application to Lévy's stochastic area

We consider $X(t) = (X_1(t), X_2(t))$ a Lévy process with values in \mathbb{R}^2 and Lévy measure σ .

\hookrightarrow The bottom structure is $(\mathbb{R}^2, \mathcal{B}(\mathbb{R}^2), \nu, \mathbf{d}, e)$.

Assume that the gradient is given by the general formula:

$$\gamma[f](x) = \alpha_{11}(x)f_1'^2(x) + 2\alpha_{12}(x)f_1'(x)f_2'(x) + \alpha_{22}(x)f_2'^2(x),$$

where $f_i' = \frac{\partial f}{\partial x_i}$.

Let be

$$V(t) = \left(X_1(t), X_2(t), \int_0^t X_1(s_-)dX_2(s) - \int_0^t X_2(s_-)dX_1(s) \right).$$

Let us apply the lent particle method...

We have for $0 < \alpha < t$ and $x = (x_1, x_2) \in \mathbb{R}^2$:

$$\varepsilon_{(\alpha, x)}^+ V = V +$$

$$(x_1, x_2, X_1(\alpha_-)x_2 + x_1(X_2(t) - X_2(\alpha)) - X_2(\alpha_-)x_1 - x_2(X_1(t) - X_1(\alpha)))$$

$$= V + (x_1, x_2, x_1(X_2(t) - 2X_2(\alpha)) - x_2(X_1(t) - 2X_1(\alpha)))$$

because $\varepsilon^+ V$ is defined $\mathbb{P} \times \nu \times d\alpha$ -a.e. and $\nu \times d\alpha$ is diffuse, so

$$\gamma[\varepsilon^+ V] = \begin{pmatrix} \alpha_{11} & \alpha_{12} & A\alpha_{11} - B\alpha_{12} \\ \alpha_{12} & \alpha_{22} & A\alpha_{12} - B\alpha_{22} \\ A\alpha_{11} - B\alpha_{12} & A\alpha_{12} - B\alpha_{22} & A^2\alpha_{11} - 2AB\alpha_{12} + B^2\alpha_{22} \end{pmatrix}$$

denoting $A = (X_2(t) - 2X_2(\alpha))$ and $B = (X_1(t) - 2X_1(\alpha))$.

The operator ε^- gives a functional defined \mathbb{P}_N -a.e. so that for example

$$\varepsilon_{(\alpha, x_1, x_2)}^-(X(t)) = X(t) - \Delta X_\alpha \quad \mathbb{P}_N(d\alpha dx_1 dx_2)\text{-a.e.}$$

This yields

$$\varepsilon^- A = X_2(t) - \Delta X_2(\alpha) - 2X_2(\alpha_-) \quad \text{let us denote it } \tilde{A}$$


$$\varepsilon^- B = X_1(t) - \Delta X_1(\alpha) - 2X_1(\alpha_-) \quad \text{let us denote it } \tilde{B}$$

and eventually $\Gamma[V] = \sum_{\alpha \leq t}$

$$\begin{pmatrix} \alpha_{11}(\Delta X_\alpha) & \alpha_{12}(\Delta X_\alpha) & \tilde{A}\alpha_{11}(\Delta X_\alpha) - \tilde{B}\alpha_{12}(\Delta X_\alpha) \\ \sim & \alpha_{22}(\Delta X_\alpha) & \tilde{A}\alpha_{12}(\Delta X_\alpha) - \tilde{B}\alpha_{22}(\Delta X_\alpha) \\ \sim & \sim & \tilde{A}^2\alpha_{11}(\Delta X_\alpha) - 2\tilde{A}\tilde{B}\alpha_{12}(\Delta X_\alpha) + \tilde{B}^2\alpha_{22}(\Delta X_\alpha) \end{pmatrix}$$

the symbol \sim denoting the symmetry of the matrix.

First case: $\sigma = \nu = k(x)dx$

Let us consider the case $\alpha_{12} = 0$, that ν possesses a density k and that we assume hypotheses of Lemma , and under these assumptions, with same notation, we may choose $\alpha_{11}(x) = \alpha_{22}(x) = (x_1^2 + x_2^2) \frac{\psi(x)}{k(x)}$. We have

$$\Gamma[V] = \sum_{\alpha \leq t} |\Delta X_\alpha|^2 \frac{\psi(\Delta X_\alpha)}{k(\Delta X_\alpha)} \begin{pmatrix} 1 & 0 & \tilde{A} \\ 0 & 0 & 0 \\ \tilde{A} & 0 & \tilde{A}^2 \end{pmatrix} +$$
$$|\Delta X_\alpha|^2 \frac{\psi(\Delta X_\alpha)}{k(\Delta X_\alpha)} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & \tilde{B} \\ 0 & \tilde{B} & \tilde{B}^2 \end{pmatrix}$$

Hence X has a density if the dimension of the vector space spanned by


$$\left(\begin{pmatrix} 1 \\ 0 \\ X_2(t) - \Delta X_2(\alpha) - 2X_2(\alpha-) \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ X_1(t) - \Delta X_1(\alpha) - 2X_1(\alpha-) \end{pmatrix} \right)$$

is equal to 3, where $JT = \{\alpha \in [0, t], \Delta X_\alpha \in O\}$.

Proposition

Let us suppose that moreover $\nu(O) = +\infty$, then V has a density.

A case where the Lévy measure is carried by a graph

Assume that the Lévy measure of X_1 is $k(x_1)dx_1$ and that it satisfies hypotheses of Lemma .

Take $X_2 = [X_1]$.

\hookrightarrow The Lévy measure of $(X_1, [X_1])$ is carried by the curve $x_2 = x_1^2$.

We put $\lambda(x_1, x_2) = 2x_1$, $\alpha_{11} = X_1^2 \frac{\psi(x)}{k(x)}$, $\alpha_{12} = \lambda\alpha_{11}$ and

$\alpha_{22} = \lambda^2\alpha_{11}$. We have

$$\Gamma[V] = \sum_{\alpha \leq t} \alpha_{11}(\Delta X_\alpha) \begin{pmatrix} 1 & \lambda & \tilde{A} - \lambda\tilde{B} \\ \lambda & \lambda^2 & \lambda\tilde{A} - \lambda^2\tilde{B} \\ \tilde{A} - \lambda\tilde{B} & \lambda\tilde{A} - \lambda^2\tilde{B} & (\tilde{A} - \lambda\tilde{B})^2 \end{pmatrix}.$$

V has a density as soon as

$$\dim \mathcal{L} \left(\begin{pmatrix} 1 \\ \lambda \\ \tilde{A} - \lambda \tilde{B} \end{pmatrix}, \alpha \in JT \right) = 3 \quad (1)$$

with $\tilde{A} - \lambda \tilde{B} = -X_2(\alpha_-) + \lambda(\Delta X(\alpha))X_1(\alpha_-) + X_2(t) - X_2(\alpha) - \lambda(\Delta X(\alpha))(X_1(t) - X_1(\alpha))$.

Proposition

$V = \left(X_1(t), [X_1]_t, \int_0^t X_1(s_-)d[X_1](s) - \int_0^t [X_1](s_-)dX_1(s) \right)$ has a density as soon as the Lévy measure of X_1 satisfies hypotheses of Lemma given at the beginning with $\nu(O) = +\infty$.

The SDE we consider

We consider another probability space $(\Omega_2, \mathcal{A}_2, \mathbb{P}_2)$ on which an \mathbb{R}^n -valued semimartingale $Z = (Z^1, \dots, Z^n)$ is defined, $n \in \mathbb{N}^*$.

Assumption on Z : There exists a positive constant C such that for any square integrable \mathbb{R}^n -valued predictable process h :

$$\forall t \geq 0, \mathbb{E}\left[\left(\int_0^t h_s dZ_s\right)^2\right] \leq C^2 \mathbb{E}\left[\int_0^t |h_s|^2 ds\right]. \quad (2)$$

We shall work on the product probability space:

$$(\Omega, \mathcal{A}, \mathbb{P}) = (\Omega_1 \times \Omega_2, \mathcal{A}_1 \otimes \mathcal{A}_2, \mathbb{P}_1 \times \mathbb{P}_2).$$

Let $d \in \mathbb{N}^*$, we consider the following SDE :

$$X_t = x + \int_0^t \int_X c(s, X_{s-}, u) \tilde{N}(ds, du) + \int_0^t \sigma(s, X_{s-}) dZ_s \quad (3)$$

where $x \in \mathbb{R}^d$, $c : \mathbb{R}^+ \times \mathbb{R}^d \times X \rightarrow \mathbb{R}^d$ and $\sigma : \mathbb{R}^+ \times \mathbb{R}^d \rightarrow \mathbb{R}^{d \times n}$ satisfy the set of hypotheses below denoted (R).

Hypotheses (R)

For simplicity, we fix a finite terminal time $T > 0$.

1. There exists $\eta \in L^2(X, \nu)$ such that:

a) for all $t \in [0, T]$ and $u \in X$, $c(t, \cdot, u)$ is differentiable with continuous derivative and

$$\forall u \in X, \quad \sup_{t \in [0, T], x \in \mathbb{R}^d} |D_x c(t, x, u)| \leq \eta(u),$$

b) $\forall (t, u) \in [0, T] \times U$, $|c(t, 0, u)| \leq \eta(u)$,

c) for all $t \in [0, T]$ and $x \in \mathbb{R}^d$, $c(t, x, \cdot) \in \mathbf{d}$ and

$$\sup_{t \in [0, T], x \in \mathbb{R}^d} \gamma[c(t, x, \cdot)](u) \leq \eta(u),$$

d) for all $t \in [0, T]$, all $x \in \mathbb{R}^d$ and $u \in X$, the matrix $I + D_x c(t, x, u)$ is invertible and

$$\sup_{t \in [0, T], x \in \mathbb{R}^d} \left| (I + D_x c(t, x, u))^{-1} \right| \leq \eta(u).$$

2. For all $t \in [0, T]$, $\sigma(t, \cdot)$ is differentiable with continuous derivative and

$$\sup_{t \in [0, T], x \in \mathbb{R}^d} |D_x \sigma(t, x)| < +\infty.$$

3. As a consequence of hypotheses 1. and 2. above, it is well known that equation (3) admits a unique solution X such that $\mathbb{E}[\sup_{t \in [0, T]} |X_t|^2] < +\infty$. We suppose that for all $t \in [0, T]$, the matrix $(I + \sum_{j=1}^n D_x \sigma_{\cdot, j}(t, X_{t-}) \Delta Z_t^j)$ is invertible and its inverse is bounded by a deterministic constant uniformly with respect to $t \in [0, T]$.

Derivation of the equation

$\mathcal{H}_{\mathbb{D}}$: the set of real valued processes $(H_t)_{t \in [0, T]}$, which belong to $L^2([0, T]; \mathbb{D})$ i.e. such that

$$\|H\|_{\mathcal{H}_{\mathbb{D}}}^2 = \mathbb{E}\left[\int_0^T |H_t|^2 dt\right] + \int_0^T \mathcal{E}(H_t) dt < +\infty.$$

Proposition

The equation (3) admits a unique solution X in $\mathcal{H}_{\mathbb{D}}^d$. Moreover, the gradient of X satisfies:

$$\begin{aligned} X_t^\# &= \int_0^t \int_U D_x c(s, X_{s-}, u) \cdot X_{s-}^\# \tilde{N}(ds, du) \\ &+ \int_0^t \int_{X \times R} c^b(s, X_{s-}, u, r) N \odot \rho(ds, du, dr) \\ &+ \int_0^t D_x \sigma(s, X_{s-}) \cdot X_{s-}^\# dZ_s \end{aligned}$$

Derivative of the flow generated by X

Let us define the $\mathbb{R}^{d \times d}$ -valued processes U and K by

$$dU_s = \sum_{j=1}^n D_x \sigma_{\cdot j}(s, X_{s-}) dZ_s^j.$$

$$K_t = I + \int_0^t \int_X D_x c(s, X_{s-}, u) K_{s-} \tilde{N}(ds, du) + \int_0^t dU_s K_{s-}$$

Under our hypotheses, for all $t \geq 0$, the matrix K_t is invertible and its inverse $\bar{K}_t = (K_t)^{-1}$ satisfies:

$$\begin{aligned} \bar{K}_t &= I - \int_0^t \int_X \bar{K}_{s-} (I + D_x c(s, X_{s-}, u))^{-1} D_x c(s, X_{s-}, u) \tilde{N}(ds, du) \\ &\quad - \int_0^t \bar{K}_{s-} dU_s + \sum_{s \leq t} \bar{K}_{s-} (\Delta U_s)^2 (I + \Delta U_s)^{-1} \\ &\quad + \int_0^t \bar{K}_s d \langle U^c, U^c \rangle_s. \end{aligned}$$

Obtaining the carré du champ matrix

Theorem

For all $t \in [0, T]$,

$$\Gamma[X_t] = K_t \int_0^t \int_X \bar{K}_s \gamma[c(s, X_{s-}, \cdot)] \bar{K}_s^* N(ds, du) K_t^*.$$

Proof. Let $(\alpha, u) \in [0, T] \times X$. We put $X_t^{(\alpha, u)} = \varepsilon_{(\alpha, u)}^+ X_t$.

$$\begin{aligned} X_t^{(\alpha, u)} &= x + \int_0^\alpha \int_X c(s, X_{s-}^{(\alpha, u)}, u') \tilde{N}(ds, du') \\ &\quad + \int_0^\alpha \sigma(s, X_{s-}^{(\alpha, u)}) dZ_s + c(\alpha, X_{\alpha-}^{(\alpha, u)}, u) \\ &\quad + \int_{] \alpha, t]} \int_X c(s, X_{s-}^{(\alpha, u)}, u') \tilde{N}(ds, du') + \int_{] \alpha, t]} \sigma(s, X_{s-}^{(\alpha, u)}) dZ_s. \end{aligned}$$

Let us remark that $X_t^{(\alpha,u)} = X_t$ if $t < \alpha$ so that, taking the gradient with respect to the variable u , we obtain:

$$\begin{aligned} (X_t^{(\alpha,u)})^b &= (c(\alpha, X_{\alpha^-}^{(\alpha,u)}, u))^b \\ &+ \int_{] \alpha, t]} \int_X D_x c(s, X_{s^-}^{(\alpha,u)}, u') \cdot (X_{s^-}^{(\alpha,u)})^b \tilde{N}(ds, du') \\ &+ \int_{] \alpha, t]} D_x \sigma(s, X_{s^-}^{(\alpha,u)}) \cdot (X_{s^-}^{(\alpha,u)})^b dZ_s. \end{aligned}$$

Let us now introduce the process $K_t^{(\alpha,u)} = \varepsilon_{(\alpha,u)}^+(K_t)$ which satisfies the following SDE:

$$K_t^{(\alpha,u)} = I + \int_0^t \int_X D_x c(s, X_{s^-}^{(\alpha,u)}, u') K_{s^-}^{(\alpha,u)} \tilde{N}(ds, du') + \int_0^t dU_s^{(\alpha,u)} K_{s^-}^{(\alpha,u)}$$

and its inverse $\bar{K}_t^{(\alpha,u)} = (K_t^{(\alpha,u)})^{-1}$. Then, using the flow property, we have:

$$\forall t \geq 0, (X_t^{(\alpha,u)})^b = K_t^{(\alpha,u)} \bar{K}_\alpha^{(\alpha,u)} (c(\alpha, X_{\alpha^-}, u))^b.$$

Now, we calculate the carré du champ and then we take back the particle:

$$\forall t \geq 0, \varepsilon_{(\alpha, u)}^- \gamma[(X_t^{(\alpha, u)})] = K_t \bar{K}_\alpha \gamma[c(\alpha, X_{\alpha^-}, \cdot)] \bar{K}_\alpha^* K_t^*$$

Finally integrating with respect to N we get

$$\forall t \geq 0, \Gamma[X_t] = K_t \int_0^t \int_X \bar{K}_s \gamma[c(s, X_{s^-}, \cdot)](u) \bar{K}_s^* N(ds, du) K_t^*.$$

First application

Proposition

Assume that X is a topological space, that the intensity measure $ds \times \nu$ of N is such that ν has an infinite mass near some point u_0 in X . If the matrix $(s, y, u) \rightarrow \gamma[c(s, y, \cdot)](u)$ is continuous on a neighborhood of $(0, x, u_0)$ and invertible at $(0, x, u_0)$, then the solution X_t of (3) has a density for all $t \in]0, T]$.

Proof.

Let us fix $t \in]0, T]$. As ν has infinite mass near u_0 , as X is right continuous and $\gamma[c]$ continuous, N admits almost surely a jump at time $s \in]0, t]$ with size $u \in X$ such that $\gamma[c(s, X_{s-}, \cdot)](u)$ is invertible. As a consequence,

$$\Gamma[X_t] \geq K_t \bar{K}_s \gamma[c(s, X_{s-}, \cdot)](u) \bar{K}_s^* K_t^*.$$

As $\Gamma[X_t]$ dominates an invertible matrix, it is also invertible and this permits to conclude.

Equation driven by a Lévy process

- Y : Lévy process with values in \mathbb{R}^d , independent of X_0 .
- $a: \mathbb{R}^+ \times \mathbb{R}^k \mapsto \mathbb{R}^{k \times d}$.


We consider the equation:

$$X_t = X_0 + \int_0^t a(s, X_{s-}) dY_s.$$

A condition which ensures that $X_t * \mathbb{P} \ll \lambda^k$

Proposition

We assume that:

1. The Lévy measure, ν , of Y satisfies hypotheses of the example  given at the beginning with $\nu(O) = +\infty$ and $\xi_{i,j}(x) = x_i^2 \delta_{i,j}$. Then we may choose the operator γ to be

$$\gamma[f] = \frac{\psi(x)}{k(x)} \sum_{i=1}^d x_i^2 \sum_{i=1}^d (\partial_i f)^2 \quad \text{for } f \in C_0^\infty(\mathbb{R}^d)$$

2. a is $C^1 \cap \text{Lip}$ with respect to the first variable uniformly in s and $\sup_{t,x} |(I + D_x a)^{-1}(x, t) \cdot u| \leq \eta(u)$
3. a is continuous with respect to the second variable at 0, and such that the matrix $aa^*(X_0, 0)$ is invertible;

then for all $t > 0$ the law of X_t is absolutely continuous w.r.t. the Lebesgue measure.

Idea of the proof ($d = 1$)

$$\gamma[f] = \frac{\psi(x)}{k(x)} x^2 f'^2(x).$$

We have the representation: $Y_t = \int_0^t \int_{\mathbb{R}} u \tilde{N}(ds, du)$, so that

$$X_t = X_0 + \int_0^t \int_{\mathbb{R}} a(s, X_{s-}) u \tilde{N}(ds, du).$$

The lent particle method yields:

$$\Gamma[X_t] = K_t^2 \int_0^t \int_X \bar{K}_s^2 a^2(s, X_{s-}) \gamma[j](u) N(ds, du)$$

where j is the identity application: $\gamma[j](u) = \frac{\psi(u)}{k(u)} u^2$.

$$\begin{aligned} \Gamma[X_t] &= K_t^2 \int_0^t \int_X \bar{K}_s^2 a^2(s, X_{s-}) \frac{\psi(u)}{k(u)} u^2 N(ds, du) \\ &= K_t^2 \sum_{\alpha < t} \bar{K}_s^2 a^2(s, X_{s-}) \frac{\psi(\Delta Y_s)}{k(\Delta Y_s)} \Delta Y_s^2. \end{aligned}$$