

# Continuity of stationary solutions of delay differential equations driven by Lévy processes

(joint work with Oleg Butkovsky)

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Statistique Asymptotique des Processus Stochastiques X

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Let  $\lambda$  be a finite signed measure on a finite interval  $J = [-r, 0]$ ,  $r \geq 0$ . Consider the equation

$$\begin{aligned} X(t) &= X(0) + \int_0^t \int_J X(s+u) \lambda(du) ds + Z(t), & t \geq 0, \\ X(t) &= X_0(t), & t \in J. \end{aligned} \quad (1)$$

Here  $Z = (Z(t), t \geq 0)$  is a real-valued process with independent stationary increments starting from 0 and having càdlàg trajectories, i.e.  $Z$  is a Lévy process, and  $X_0 = (X_0(t), t \in J)$  is an initial process with càdlàg trajectories, **independent** of  $Z$ .

Let  $\mathbb{D}_0 = \mathbb{D}([-r, 0], \mathbb{R})$  be the Skorokhod space of càdlàg real-valued functions on  $[-r, 0]$  with the usual Skorokhod topology. Denote by  $\mathbb{M}$  the set of all finite signed measures on  $[-r, 0]$ .

Define as usual,  $X_t := \{X(t+s), -r \leq s \leq 0\}$ ,  $t \geq 0$ .

Equation (1) has a unique strong solution and the process  $X := (X_t)_{t \geq 0}$  is a homogeneous Markov process taking values in  $(\mathbb{D}_0, \mathcal{B}(\mathbb{D}_0))$ .

To emphasize that  $X$  is a solution to (1) with the delay measure  $\lambda \in \mathbb{M}$  and the initial distribution  $\alpha := \text{Law}(X_0)$  we write  $X = X^{(\lambda, \alpha)}$ .

If the initial condition is deterministic, i.e.,  $\alpha = \delta_y$ ,  $y \in \mathbb{D}$ , then we simplify the notation and write  $X = X^{(\lambda, y)}$ .

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**Assumption A1.** The characteristic equation

$$h_\lambda(z) := z - \int_{[-r,0]} e^{zu} \lambda(du) = 0 \quad (2)$$

has no complex solutions  $z$  with  $\operatorname{Re} z \geq 0$ .

**Assumption A2.** We have

$$\int_{|y|>1} \log |y| F(dy) < \infty,$$

where  $F$  is the Lévy measure of  $Z$ .

It was shown by **G. and Küchler (2000)** that the process  $X$  has an invariant measure if and only if **A1** and **A2** are satisfied. Moreover, in this case the invariant measure is unique. We assume that these conditions hold and denote the invariant measure of  $X^{(\lambda)}$  by  $\pi^\lambda$ . We denote by  $\mathbb{M}^-$  the set of all finite signed measures on  $[-r, 0]$  that satisfy **A1**.

We study the relationship between invariant measures of this SDDE under an additional condition on the Lévy process  $Z$ .

**Assumption A3.** The Lévy process  $Z$  has a non-zero Gaussian component.

This assumption implies that a Gaussian component  $X^c$  (which is also the continuous martingale component of  $X$  and  $Z$ ) has mean zero and variance  $\sigma^2 t$ ,  $\sigma > 0$ .

## Formula for the likelihood ratio

For any interval  $I \subset \mathbb{R}$  we denote

$P_I^{(\lambda, \alpha)} := \text{Law}(\{X^{(\lambda, \alpha)}(s), s \in I\})$ , where  $\lambda \in \mathbb{M}$ .

Let  $X = X^{(\lambda, \alpha)}$ .

### Proposition

**Assume A3.** Let  $\alpha$  and  $\beta$  be two probability measures on  $\mathbb{D}_0$ . If  $\beta \ll \alpha$  (resp.  $\beta \sim \alpha$ ) then, for any  $T > 0$  and any delay measures  $\lambda, \mu \in \mathbb{M}$  we have  $P_{[-r, T]}^{(\mu, \beta)} \ll P_{[-r, T]}^{(\lambda, \alpha)}$  (resp.  $P_{[-r, T]}^{(\mu, \beta)} \sim P_{[-r, T]}^{(\lambda, \alpha)}$ ) and

$$\frac{dP_{[-r, T]}^{(\mu, \beta)}}{dP_{[-r, T]}^{(\lambda, \alpha)}}(X) = \frac{d\beta}{d\alpha}(X_0) \exp \left\{ \frac{1}{\sigma^2} \int_0^T A(t) dX^c(t) - \frac{1}{2\sigma^2} \int_0^T A^2(t) dt \right\},$$

where

$$A(t) = \int_J X(t+u) (\mu - \lambda)(du), \quad \lambda, \mu \in \mathbb{M}.$$



## Proposition

Assume **A2** and **A3**. Then for any delay measures  $\lambda, \mu \in \mathbb{M}^-$ , the corresponding invariant measures  $\pi^\lambda$  and  $\pi^\mu$  are equivalent.

**Proof:** By the previous proposition,

$$\mathbf{P}_{[-r,r]}^{(\lambda, \pi(\lambda))} \sim \mathbf{P}_{[-r,r]}^{(-\delta_0, \pi(\lambda))},$$

therefore,

$$\mathbf{P}_{[0,r]}^{(\lambda, \pi(\lambda))} \sim \mathbf{P}_{[0,r]}^{(-\delta_0, \pi(\lambda))}.$$

However,  $\text{Law}(X^{(-\delta_0, \pi(\lambda))}(0)) = \text{Law}(X^{(\lambda)}(0))$  is a convolution of a nondegenerate Gaussian distribution and another one; hence it is equivalent to a nondegenerate Gaussian distribution. Therefore,

$$\mathbf{P}_{[0,r]}^{(-\delta_0, \pi(\lambda))} \sim \mathbf{P}_{[0,r]}^{(-\delta_0, \pi(-\delta_0))}.$$

Combining. we get

$$\pi(\lambda) = \mathbf{P}_{[0,r]}^{(\lambda, \pi(\lambda))} \sim \mathbf{P}_{[0,r]}^{(-\delta_0, \pi(-\delta_0))} = \pi(-\delta_0).$$

Our goal now is to show that invariant measures are close when delay measure are close. This was done in [G. & Küchler \(2003, 2011\)](#) in the case where  $Z$  is a Brownian motion.

$\|\cdot\|_{\text{TV}}$  stands for the total variation norm.

We shall also need a Wasserstein-type distance between probability measures on  $\mathbb{D}_0$ . Namely,

$$W_d(\alpha, \beta) := \inf E d(X, Y),$$

where

$$d(x, y) := \sup_{t \in J} |x(t) - y(t)| \wedge 1$$

and the infimum is taken over all random elements  $X$  and  $Y$  distributed as  $\alpha$  and  $\beta$ , correspondingly.

It is known that  $W_d(\alpha, \beta) \leq \|\beta - \alpha\|_{\text{TV}}$ .

## Theorem (Butkovsky & G. (2014))

Let  $\lambda, \lambda_1, \lambda_2, \dots$  be a sequence of finite signed measures on  $[-r, 0]$ , satisfying **A1**. Suppose that the Lévy process  $Z$  satisfies **A2–A3**. If  $\lambda_n$  converges weakly to  $\lambda$  as  $n \rightarrow \infty$  then

$$\lim_{n \rightarrow \infty} d_{TV}(\pi^{\lambda_n}, \pi^\lambda) = 0.$$

Since the equation (1) involves no stochastic integrals and is treated pathwise, we will formulate a number of results for solutions of the equation (1) with deterministic  $Z$  and  $X_0$ .

Given a measure  $\lambda$ , we call a function  $x_0: [-r, \infty[ \rightarrow \mathbb{R}$  the fundamental solution of the homogeneous equation

$$\begin{aligned} X(t) &= X(0) + \int_0^t \int_J X(s+u) \lambda(du) ds, & t \geq 0, \\ X(t) &= X_0(t), & t \in J, \end{aligned} \quad (3)$$

if it is the solution of (3) corresponding to the initial condition

$$X_0(t) = \begin{cases} 1, & t = 0, \\ 0, & -r \leq t < 0. \end{cases}$$

In other words, a function  $x_0(t)$ ,  $t \geq -r$ , is the fundamental solution of (3) if it is absolutely continuous on  $\mathbb{R}_+$ ,  $x_0(t) = 0$  for  $t < 0$ ,  $x_0(0) = 1$ , and

$$\dot{x}_0(t) = \int_J x_0(t+u) \lambda(du) \quad (4)$$

for Lebesgue-almost all  $t > 0$ .

## The variation-of-constants formula

The solution of (1) can be represented via the fundamental solution  $x_0$  of (3):

$$X(t) = \begin{cases} x_0(t)X_0(0) + \int_J \int_u^0 X_0(s)x_0(t+u-s) ds \lambda(du) \\ \quad + \int_{[0,t]} Z(t-s) dx_0(s), & t \geq 0, \\ X_0(t), & t \in J. \end{cases} \quad (5)$$

### Remark

*The domain of integration in the last integral in (5) includes zero:*

$$\int_{[0,t]} Z(t-s) dx_0(s) = Z(t) + \int_{]0,t]} Z(t-s) dx_0(s).$$



The asymptotic behaviour of solutions of the equations (1) and (3) for  $t \rightarrow \infty$  is connected with the set of complex solutions of the characteristic equation

$$h(z) := z - \int_J e^{zu} \lambda(du) = 0. \quad (6)$$

Note that a complex number  $z$  solves (6) if and only if  $(e^{zt}, t \geq -r)$  solves (3) for the initial condition  $X_0(t) = e^{zt}, t \in J$ .

The set  $\Lambda := \{z \in \mathbb{C} \mid h(z) = 0\}$  is not empty; moreover, it is infinite except the case where  $\lambda$  is concentrated at 0. Since  $h(\cdot)$  is an entire function,  $\Lambda$  consists of isolated points only. It is easy to check that  $z_n \in \Lambda$  and  $|z_n| \rightarrow \infty$  imply  $\operatorname{Re} z_n \rightarrow -\infty$ , thus the set  $\{z \in \Lambda \mid \operatorname{Re} z \geq c\}$  is finite for every  $c \in \mathbb{R}$ . In particular, it holds

$$v_0 = v_0(\lambda) := \max \{ \operatorname{Re} z \mid z \in \Lambda \} < \infty. \quad (7)$$

It is easy to check from (4) that  $1/h(z)$  is the Laplace transform of  $(x_0(t), t \geq 0)$  at least if  $\operatorname{Re} z$  is large enough. (In fact,

$$1/h(z) = \int_0^{\infty} e^{-zt} x_0(t) dt$$

if  $\operatorname{Re} z > \nu_0$ .) Applying a standard method based on the inverse Laplace transform and Cauchy's residue theorem, it is easy to show that for any  $\gamma > \nu_0$

$$x_0(t) = o(e^{\gamma t}) \quad \text{as } t \rightarrow \infty.$$

## Formula for the solution of the equation (1)

Return to the case where  $Z$  is a Lévy process. Integration by parts gives

$$\int_{[0,t]} Z(t-s) dx_0(s) = \int_0^t x_0(t-s) dZ(s).$$

Thus, using (5), any solution of the equation (1) can be written in the form

$$X(t) = x_0(t)X_0(0) + \int_J \int_u^0 X_0(s)x_0(t+u-s) ds \lambda(du) + \int_0^t x_0(t-s) dZ(s), \quad (8)$$

Note also that

$$\int_0^t x_0(t-s) dZ(s) \stackrel{d}{=} \int_0^t x_0(s) dZ(s). \quad (9)$$

## Representation of the stationary solution

Let  $Z = (Z(t), t \geq 0)$  and  $\tilde{Z} = (\tilde{Z}(t), t \geq 0)$  be two independent Lévy processes with the same characteristics  $(b, c, F)$ . We define a two-sided Lévy process  $(Z(t), t \in \mathbb{R})$  by

$$Z(t) = \begin{cases} Z(t), & t \geq 0, \\ -\tilde{Z}(-t - 0), & t < 0, \end{cases}$$

and put

$$X(t) = \int_{-\infty}^t x_0(t-s) dZ(s).$$

The process  $X = (X(t), t \geq -r)$  is well defined up to a modification under Assumptions **A1** and **A2** and is a stationary solution to (1).

The first part of the proof consists of a series of (more or less standard) lemmas having the aim to show that

$$\lim_{n \rightarrow \infty} W_d(\pi(\lambda_n), \pi(\lambda)) = 0.$$

The second part uses a coupling-type argument for stochastic functional equations due to [Es-Sarhir, von Renesse, Scheutzow \(2009\)](#).

Notation:  $\pi_n := \pi(\lambda_n)$ ,  $\pi := \pi(\lambda)$ .

For  $M \geq 0$  consider the “ball” in  $\mathbb{D}_0$

$$B_M := \{x \in \mathbb{D}_0 : \sup_{t \in J} |x(t)| \leq M\}.$$

### Lemma

*Assume that a sequence of measures  $\{\lambda_n\}$ ,  $\lambda_n \in \mathbb{M}^-$ , converges weakly to a measure  $\lambda \in \mathbb{M}^-$  as  $n \rightarrow \infty$ . Then for any  $\varepsilon > 0$  there exists  $M > 0$  such that*

$$\pi_n(\mathbb{D} \setminus B_M) < \varepsilon \quad \text{for all } n.$$

### Lemma

*Under the assumptions of Theorem, we have, for any  $M, N > 0$ ,  $t \geq 0$ ,*

$$\begin{aligned} \left\| P_{[-r,t]}^{(\lambda)} - P_{[-r,t]}^{(\lambda_n, \pi)} \right\|_{\text{TV}} &\leq C(n; N, t) + 2\pi(\mathbb{D} \setminus B_M) \\ &\quad + 2 \sup_{x \in B_M} P\left( \sup_{s \in [-r,t]} |X^{(\lambda, x)}(s)| \geq N \right), \end{aligned}$$

*where, for every  $N$  and  $t$ ,*

$$C(n; N, t) \rightarrow 0, \quad n \rightarrow \infty.$$



### Lemma

*Under the assumptions of Theorem, we have, for any  $M > 0$  and  $t \geq 2r$ ,*

$$W_d(\mathbb{P}_{[t-r,t]}^{(\lambda)}, \mathbb{P}_{[t-r,t]}^{(\lambda, \lambda_n)}) \leq (1 + \|\lambda\|_{\text{TV}} r) (M w^{(\lambda)}(t) + \pi(\mathbb{D} \setminus B_M) + \pi_n(\mathbb{D} \setminus B_M)),$$

*where  $w^{(\lambda)}(t) \rightarrow 0$  as  $t \rightarrow \infty$ .*

Combining the statements of previous lemmas and taking

$$\limsup_{M \rightarrow \infty} \limsup_{t \rightarrow \infty} \limsup_{N \rightarrow \infty} \limsup_{n \rightarrow \infty},$$

we get

$$\lim_{n \rightarrow \infty} W_d(\pi(\lambda_n), \pi(\lambda)) = 0.$$

### Lemma

For any  $x, y \in \mathbb{D}_0$  and  $t > t_0 + r$ ,  $t_0 > 0$  being fixed,

$$\begin{aligned} \|\mathbb{P}_t^{(\lambda, x)} - \mathbb{P}_t^{(\lambda, y)}\|_{\text{TV}} &\leq \frac{\sqrt{2}|x(0) - y(0)|}{\sigma t_0^{1/2}} \\ &\quad + \sqrt{2}\|\lambda\|_{\text{TV}}(t_0 + r)^{1/2} \sup_{u \in J} |x(u) - y(u)| \sigma^{-1}. \end{aligned}$$

## Sketch of the proof: coupling

Let  $X = X^{(\lambda, x)}$ ,

$$R(t) = \begin{cases} x(t) - y(t), & \text{if } t \in J; \\ R(0)\chi(t), & \text{if } t \in [0, t_0]; \\ 0, & \text{if } t \geq t_0, \end{cases}$$

where  $\chi(t)$ ,  $0 \leq t \leq t_0$ , is an absolutely continuous decreasing function,  $\chi(0) = 1$ ,  $\chi(t_0) = 0$ ,  $Y(t) := X(t) - R(t)$ ,  $Q(t) := R'(t) - \int_J R(t+u)\lambda(du)$ . Note that  $Q(t) = 0$  for  $t \geq t_0 + r$ . Then we can rewrite  $\tilde{Y}$  as the solution of

$$Y(t) = Y(0) + \int_0^t \int_J Y(s+u) \lambda(du) ds + \tilde{Z}(t), \quad t \geq 0,$$
$$Y(t) = y(t), \quad t \in J,$$

where  $\tilde{Z}(t) = Z(t) - \int_0^t Q(s) ds$ .

By Girsanov's theorem,  $\text{Law}(\tilde{Z}|Q) = \text{Law}(Z|P)$ , where  $Q$  is a new measure with the density process

$$\exp\left(\sigma^{-2} \int_0^t Q(s) dZ^c(s) - (2\sigma^2)^{-1} \int_0^t Q^2(s) ds\right)$$

with respect to  $P$ . This means that the law of  $Y$  under  $Q$  coincides with the law of  $X^{(\lambda,y)}$ .

Now it follows, for  $t \geq t_0 + r$ ,

$$\begin{aligned} \|\mathbf{P}_t^{(\lambda,x)} - \mathbf{P}_t^{(\lambda,y)}\|_{\text{TV}} &= 2 \sup_A |\mathbf{P}(X_t \in A) - \mathbf{Q}(Y_t \in A)| \\ &\leq \|\mathbf{P} - \mathbf{Q}\|_{\text{TV}} \\ &\leq \left(2\sigma^{-2} \int_0^{t_0+r} Q^2(s) ds\right)^{1/2} \\ &\leq \dots \end{aligned}$$

## Rest of the proof of Theorem

It follows from the definition of  $W_d$  and Chebyshev's inequality that there exist random elements  $Y^n$  and  $Y$  distributed as  $\pi_n$  and  $\pi$ , respectively, such that

$$P(\sup_{u \in J} |Y^n(u) - Y(u)| > \gamma) \leq W_d(\pi_n, \pi) / \gamma,$$

for any  $\gamma \in (0, 1)$ . The fourth lemma now yields

$$\|P_t^{(\lambda, \lambda_n)} - P_t^{(\lambda)}\|_{TV} \leq \frac{W_d(\pi_n, \pi)}{\gamma} + \frac{\sqrt{2}\gamma}{\sigma(t-r)^{1/2}} + \sqrt{2}\|\lambda\|_{TV} t^{1/2} \gamma \sigma^{-1}.$$

Using the first part of the proof and taking consequently limits in the inequality above as  $n \rightarrow \infty$  and  $\gamma \rightarrow 0$ , we get

$$\lim_{n \rightarrow \infty} \|P_t^{(\lambda, \lambda_n)} - P_t^{(\lambda)}\|_{TV} = 0.$$

Finally, the same argument as in the second lemma shows that

$$\lim_{n \rightarrow \infty} \|P_t^{(\lambda, \lambda_n)} - P_t^{(\lambda_n)}\|_{TV} = 0.$$

Thank you for your attention!