

Exponential functionals of Markov additive processes



Anita Behme

joint work in progress with
Apostolos Sideris

May 24th 2019
Probability and Analysis

Overview

- ▶ Exponential functionals of Lévy processes
- ▶ From Lévy processes to MAPs
- ▶ Exponential functionals of MAPs:
 - ▶ Definition
 - ▶ An example
 - ▶ Main results and methodology
 - ▶ Discussion and more results
 - ▶ Open questions

Generalized Ornstein-Uhlenbeck processes

Let $(\xi_t, \eta_t)_{t \geq 0}$ be a bivariate Lévy process.

The **generalized Ornstein-Uhlenbeck (GOU) process** $(V_t)_{t \geq 0}$ driven by (ξ, η) is given by

$$V_t = e^{-\xi t} \left(V_0 + \int_{(0,t]} e^{\xi_s} d\eta_s \right), \quad t \geq 0,$$

where V_0 is a finite random variable, independent of (ξ, η) .

Lévy processes

Definition: A Lévy process in \mathbb{R}^d on a probability space (Ω, \mathcal{F}, P) is a stochastic process $X = (X_t)_{t \geq 0}$, $X_t : \Omega \rightarrow \mathbb{R}^d$ satisfying the following properties:

- ▶ $X_0 = 0$ a.s.
- ▶ X has **independent increments**, i.e. for all $0 \leq t_0 \leq t_1 \leq \dots \leq t_n$ the random variables $X_{t_0}, X_{t_1} - X_{t_0}, \dots, X_{t_n} - X_{t_{n-1}}$ are independent.
- ▶ X has **stationary increments**, i.e. for all $s, t \geq 0$ it holds $X_{s+t} - X_s \stackrel{d}{=} X_t$.
- ▶ X has a.s. **càdlàg paths**, i.e. for P -a.e. $\omega \in \Omega$ the path $t \mapsto X_t(\omega)$ is right-continuous in $t \geq 0$ and has left limits in $t > 0$.

Lévy processes

Definition: A Lévy process in \mathbb{R}^d on a probability space (Ω, \mathcal{F}, P) is a stochastic process $X = (X_t)_{t \geq 0}$, $X_t : \Omega \rightarrow \mathbb{R}^d$ satisfying the following properties:

- ▶ $X_0 = 0$ a.s.
- ▶ X has **independent increments**, i.e. for all $0 \leq t_0 \leq t_1 \leq \dots \leq t_n$ the random variables $X_{t_0}, X_{t_1} - X_{t_0}, \dots, X_{t_n} - X_{t_{n-1}}$ are independent.
- ▶ X has **stationary increments**, i.e. for all $s, t \geq 0$ it holds $X_{s+t} - X_s \stackrel{d}{=} X_t$.
- ▶ X has a.s. **càdlàg paths**, i.e. for P -a.e. $\omega \in \Omega$ the path $t \mapsto X_t(\omega)$ is right-continuous in $t \geq 0$ and has left limits in $t > 0$.

A Lévy process X is uniquely determined by its characteristic triplet $(\gamma_X, \sigma_X, \nu_X)$.

Exponential functionals

Theorem (Lindner, Maller '05): The GOU process

$$V_t = e^{-\xi_t} \left(V_0 + \int_0^t e^{\xi_s} d\eta_s \right), \quad t \geq 0,$$

solving

$$dV_t = V_{t-} dU_t + dL_t,$$

with

- ▶ $\xi_t = -\log(\mathcal{E}(U)_t)$
- ▶ (U, L) (or similarly (ξ, η)) bivariate Lévy processes
- ▶ V_0 starting random variable independent of (ξ, η)

has a (nontrivial) **stationary distribution** if and only if the integral

$$V_\infty := \int_0^\infty e^{-\xi_{t-}} dL_t$$

converges a.s.

Exponential functionals

Theorem (Lindner, Maller '05): The GOU process

$$V_t = e^{-\xi t} \left(V_0 + \int_0^t e^{\xi s} d\eta_s \right), \quad t \geq 0,$$

solving

$$dV_t = V_{t-} dU_t + dL_t,$$

with

- ▶ $\xi_t = -\log(\mathcal{E}(U)_t)$
- ▶ (U, L) (or similarly (ξ, η)) bivariate Lévy processes
- ▶ V_0 starting random variable independent of (ξ, η)

has a (nontrivial) **stationary distribution** if and only if the integral

$$V_\infty := \int_0^\infty e^{-\xi t} dL_t$$

converges a.s.

The stationary distribution is given by the law of the so-called **exponential functional** V_∞ .

Exponential functionals

Erickson and Maller (2005) proposed **necessary and sufficient conditions for convergence** of

$$V_\infty := \int_0^\infty e^{-\xi_t} d\eta_t.$$

Mainly one needs:

- ▶ ξ tends to infinity
- ▶ η has a finite \log^+ -moment

Convergence of exponential functionals

Precisely, they stated

Theorem (Erickson, Maller '05):

V_∞ exists as a.s. limit as $t \rightarrow \infty$ of $\int_0^t e^{-\xi_s} d\eta_s$ if and only if

$$\lim_{t \rightarrow \infty} \xi_t = \infty \text{ a.s. and } I_{\xi, \eta} = \int_{(e^a, \infty)} \left(\frac{\log y}{A_\xi(\log y)} \right) |\bar{\nu}_\eta(dy)| < \infty,$$

where

$$A_\xi(x) = \gamma_\xi + \bar{\nu}_\xi^+(1) + \int_1^x \bar{\nu}_\xi^+(y) dy,$$

with

$$\bar{\nu}_\xi^+(x) = \nu_\xi((x, \infty)), \quad \bar{\nu}_\xi^-(x) = \nu_\xi((-\infty, -x)), \quad \bar{\nu}_\xi(x) = \bar{\nu}_\xi^+(x) + \bar{\nu}_\xi^-(x),$$

and $\bar{\nu}_\eta^+$, $\bar{\nu}_\eta^-$ and $\bar{\nu}_\eta$ defined likewise.

Hereby $a > 0$ is chosen such that $A_\xi(x) > 0$ for all $x > 0$ and its existence is guaranteed whenever $\lim_{t \rightarrow \infty} \xi_t = \infty$ a.s.

Convergence of exponential functionals

Further:

Theorem (Erickson, Maller '05, continued):

If $\lim_{t \rightarrow \infty} \xi_t = \infty$ a.s. but $I_{\xi, \eta} = \infty$, then

$$\left| \int_0^t e^{-\xi_s} d\eta_s \right| \xrightarrow{\mathbb{P}} \infty, \quad (1)$$

while for $\lim_{t \rightarrow \infty} \xi_t = -\infty$ or oscillating ξ either (1) holds, or there exists some $k \in \mathbb{R} \setminus \{0\}$ such that

$$\int_0^t e^{-\xi_s} d\eta_s = k(1 - e^{-\xi_t}) \quad \text{for all } t > 0 \quad \text{a.s.}$$

Markov additive processes (MAPs)

Markov additive processes (MAPs)

$(\xi_t, \eta_t, J_t)_{t \geq 0}$ is a (bivariate) MAP with

- ▶ **Markovian component** $(J_t)_{t \geq 0}$: Right-continuous, ergodic, continuous time Markov chain with countable state space \mathcal{S} , intensity matrix Q and stationary law π .

Markov additive processes (MAPs)

$(\xi_t, \eta_t, J_t)_{t \geq 0}$ is a (bivariate) MAP with

- ▶ **Markovian component** $(J_t)_{t \geq 0}$: Right-continuous, ergodic, continuous time Markov chain with countable state space \mathcal{S} , intensity matrix Q and stationary law π .
- ▶ **Additive component** $(\xi_t, \eta_t)_{t \geq 0}$:

$$(\xi_t, \eta_t) := (X_t^{(1)}, Y_t^{(1)}) + (X_t^{(2)}, Y_t^{(2)}), \quad t \geq 0.$$

- ▶ $(X_t^{(1)}, Y_t^{(1)})$ behaves in law like a bivariate Lévy process $(\xi_t^{(j)}, \eta_t^{(j)})$ whenever $J_t = j$,
- ▶ $(X_t^{(2)}, Y_t^{(2)})$ is a pure jump process given by

$$(X_t^{(2)}, Y_t^{(2)}) = \sum_{n \geq 1} \sum_{i, j \in \mathcal{S}} Z_n^{(i, j)} \mathbb{1}_{\{J_{T_n^-} = i, J_{T_n} = j, T_n \leq t\}},$$

for i.i.d. random variables $Z_n^{(i, j)}$ in \mathbb{R}^2 .

Markov additive processes (MAPs)

$(\xi_t, \eta_t, J_t)_{t \geq 0}$ is a (bivariate) MAP with

- ▶ **Markovian component** $(J_t)_{t \geq 0}$: Right-continuous, ergodic, continuous time Markov chain with countable state space \mathcal{S} , intensity matrix Q and stationary law π .
- ▶ **Additive component** $(\xi_t, \eta_t)_{t \geq 0}$:

$$(\xi_t, \eta_t) := (X_t^{(1)}, Y_t^{(1)}) + (X_t^{(2)}, Y_t^{(2)}), \quad t \geq 0.$$

- ▶ $(X_t^{(1)}, Y_t^{(1)})$ behaves in law like a bivariate Lévy process $(\xi_t^{(j)}, \eta_t^{(j)})$ whenever $J_t = j$,
- ▶ $(X_t^{(2)}, Y_t^{(2)})$ is a pure jump process given by

$$(X_t^{(2)}, Y_t^{(2)}) = \sum_{n \geq 1} \sum_{i, j \in \mathcal{S}} Z_n^{(i, j)} \mathbb{1}_{\{J_{T_n^-} = i, J_{T_n} = j, T_n \leq t\}},$$

for i.i.d. random variables $Z_n^{(i, j)}$ in \mathbb{R}^2 .

As starting value under \mathbb{P}_j we use $(\xi_0, \eta_0, J_0) = (0, 0, j)$.

Throughout neither ξ nor η is degenerate constantly equal to 0.

Special cases of MAPs

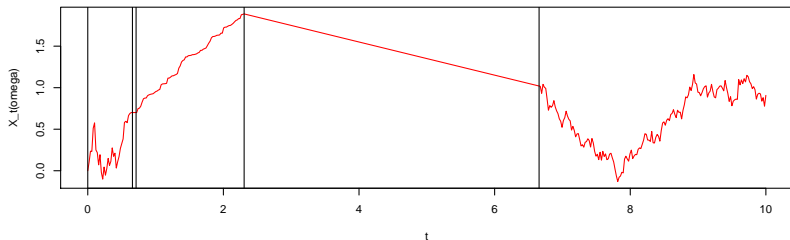
- ▶ $\mathcal{S} = \{0\}$: (bivariate) Lévy process

Special cases of MAPs

- ▶ $\mathcal{S} = \{0\}$: (bivariate) Lévy process
- ▶ $(X_t^{(1)}, Y_t^{(1)}) \equiv 0$: (bivariate) continuous-time Markov chain in \mathbb{R}^2

Special cases of MAPs

- ▶ $\mathcal{S} = \{0\}$: (bivariate) Lévy process
- ▶ $(X_t^{(1)}, Y_t^{(1)}) \equiv 0$: (bivariate) continuous-time Markov chain in \mathbb{R}^2
- ▶ $(X_t^{(2)}, Y_t^{(2)}) \equiv 0$: no common jumps of $(J_t)_{t \geq 0}$ and $(\xi_t, \eta_t)_{t \geq 0}$:



Exponential functionals of MAPs

Exponential functionals of MAPs

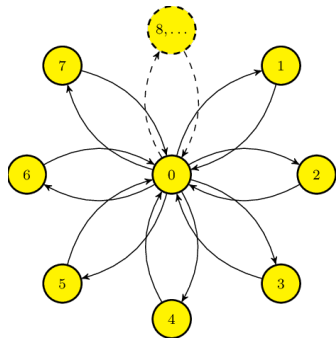
Given a bivariate Markov additive process $(\xi_t, \eta_t, J_t)_{t \geq 0}$ with Markovian component $(J_t)_{t \geq 0}$, we denote

$$\mathfrak{E}(t) := \mathfrak{E}_{(\xi, \eta)}(t) := \int_{(0, t]} e^{-\xi_s} d\eta_s, \quad 0 < t < \infty.$$

Literature:

Some recent results on \mathfrak{E} for $\eta_t = t$ on arXiv
(Salminen et al./Stephenson)

An example¹



▶ $S = \mathbb{N}_0$

▶ $(J_t)_{t \geq 0}$ continuous time petal flower Markov chain with intensity matrix

$$Q = (q_{i,j})_{i,j \in \mathbb{N}_0} = \begin{pmatrix} -q & q_{0,1} & q_{0,2} & \cdots \\ q & -q & 0 & \cdots \\ q & 0 & -q & \cdots \\ \vdots & \vdots & & \ddots \end{pmatrix}$$

for $q > 0$ fixed and $q_{0,j} = qp_{0,j}$
 $j \in \mathbb{N}$.

$(J_t)_{t \geq 0}$ is irreducible, recurrent with stationary distribution

$$\pi_0 = \frac{1}{2}, \quad \text{and} \quad \pi_j = \frac{p_{0,j}}{2} = \frac{q_{0,j}}{2q}, \quad j \in \mathbb{N}$$

¹Thanks to Gerold Alsmeyer for the picture!

An example

Now choose ξ and η to be conditionally independent with

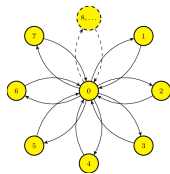
$$\xi_t = X_t^{(2)} = \sum_{n \geq 1} \sum_{i,j \in \mathbb{N}_0} Z_n^{(i,j)} \mathbb{1}_{\{J_{T_n-} = i, J_{T_n} = j, T_n \leq t\}}, \quad Z_n^{(i,j)} := \begin{cases} -p_{0,j}^{-1}, & i = 0, \\ 2 + p_{0,i}^{-1}, & j = 0, \\ 0, & \text{else.} \end{cases}$$

Then

$$\xi_{\tau_n(0)} = 2n \rightarrow \infty \quad \mathbb{P}_0\text{-a.s.}$$

but $\liminf_{t \rightarrow \infty} \xi_t = -\infty$.

Thus ξ is oscillating.



An example

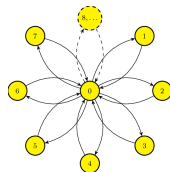
Now choose ξ and η to be conditionally independent with

$$\xi_t = X_t^{(2)} = \sum_{n \geq 1} \sum_{i,j \in \mathbb{N}_0} Z_n^{(i,j)} \mathbb{1}_{\{J_{T_n^-} = i, J_{T_n} = j, T_n \leq t\}}, \quad Z_n^{(i,j)} := \begin{cases} -p_{0,j}^{-1}, & i = 0, \\ 2 + p_{0,i}^{-1}, & j = 0, \\ 0, & \text{else.} \end{cases}$$

Then

$$\xi_{\tau_n(0)} = 2n \rightarrow \infty \quad \mathbb{P}_0\text{-a.s.}$$

but $\liminf_{t \rightarrow \infty} \xi_t = -\infty.$



Thus ξ is oscillating.

Choosing $\eta_t = \int_{(0,t]} \gamma_{\eta(s)} ds$ with $\gamma_{\eta(s)} = \begin{cases} 1, & j = 0, \\ 0, & \text{otherwise,} \end{cases}$

we observe under \mathbb{P}_0

$$\int_{(0,t]} e^{-\xi_{s-}} d\eta_s = \int_{(0,t]} e^{-\xi_{s-} - \gamma_{\eta(s)}} ds = \int_{(0,t]} e^{-N_s - \mathbb{1}_{\{J_s=0\}}} ds.$$

An example

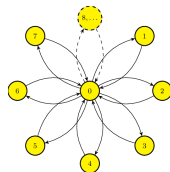
Now choose ξ and η to be conditionally independent with

$$\xi_t = X_t^{(2)} = \sum_{n \geq 1} \sum_{i,j \in \mathbb{N}_0} Z_n^{(i,j)} \mathbb{1}_{\{J_{T_n-}=i, J_{T_n}=j, T_n \leq t\}}, \quad Z_n^{(i,j)} := \begin{cases} -p_{0,j}^{-1}, & i = 0, \\ 2 + p_{0,i}^{-1}, & j = 0, \\ 0, & \text{else.} \end{cases}$$

Then

$$\xi_{\tau_n(0)} = 2n \rightarrow \infty \quad \mathbb{P}_0\text{-a.s.}$$

but $\liminf_{t \rightarrow \infty} \xi_t = -\infty.$



Thus ξ is oscillating.

Choosing $\eta_t = \int_{(0,t]} \gamma_{\eta(s)} ds$ with $\gamma_{\eta(s)} = \begin{cases} 1, & j = 0, \\ 0, & \text{otherwise,} \end{cases}$

we observe under \mathbb{P}_0

$$\int_{(0,t]} e^{-\xi_s} d\eta_s = \int_{(0,t]} e^{-\xi_s - \gamma_{\eta(s)}} ds = \int_{(0,t]} e^{-N_s} \mathbb{1}_{\{J_s=0\}} ds.$$

I.e. the exponential functional converges \mathbb{P}_0 -a.s. as $t \rightarrow \infty$.

An example

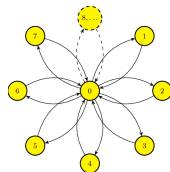
Now choose ξ and η to be conditionally independent with

$$\xi_t = X_t^{(2)} = \sum_{n \geq 1} \sum_{i,j \in \mathbb{N}_0} Z_n^{(i,j)} \mathbb{1}_{\{J_{T_n-}=i, J_{T_n}=j, T_n \leq t\}}, \quad Z_n^{(i,j)} := \begin{cases} -p_{0,j}^{-1}, & i = 0, \\ 2 + p_{0,i}^{-1}, & j = 0, \\ 0, & \text{else.} \end{cases}$$

Then

$$\xi_{\tau_n(0)} = 2n \rightarrow \infty \quad \mathbb{P}_0\text{-a.s.}$$

but $\liminf_{t \rightarrow \infty} \xi_t = -\infty$.



Thus ξ is oscillating.

Choosing $\eta_t = \int_{(0,t]} \gamma_{\eta(s)} ds$ with $\gamma_{\eta(s)} = \begin{cases} 1, & j = 0, \\ 0, & \text{otherwise,} \end{cases}$

we observe under \mathbb{P}_0

$$\int_{(0,t]} e^{-\xi_{s-}} d\eta_s = \int_{(0,t]} e^{-\xi_{s-} - \gamma_{\eta(s)}} ds = \int_{(0,t]} e^{-N_s - \mathbb{1}_{\{J_s=0\}}} ds.$$

I.e. the exponential functional converges \mathbb{P}_0 -a.s. as $t \rightarrow \infty$.

Methodology

Methodology

Lévy case (Erickson & Maller '05): The exponential functional

$$\int_{(0,\infty)} e^{-\xi_t} d\eta_t,$$

can be discretized for any $h > 0$ as:

$$\begin{aligned} \int_{(0,nh]} e^{-\xi_s} d\eta_s &= \sum_{i=0}^{n-1} \int_{(ih,(i+1)h]} e^{-\xi_s} d\eta_s \\ &= \sum_{i=0}^{n-1} \left(\prod_{j=0}^{i-1} e^{-(\xi_{(j+1)h} - \xi_{jh})} \right) \int_{(ih,(i+1)h]} e^{-(\xi_s - \xi_{ih})} d\eta_s, \end{aligned}$$

Thus: Convergence of the integral is strongly connected to convergence of [discrete-time perpetuities](#) as studied by Goldie & Maller (2000).

Methodology II

Alsmeyer & Buckmann (2017) generalized the results from Goldie & Maller (2000) to a Markovian environment:

Methodology II

Alsmeyer & Buckmann (2017) generalized the results from Goldie & Maller (2000) to a Markovian environment:

They provide necessary and sufficient conditions for convergence as $n \rightarrow \infty$ of

$$Z_n := \sum_{i=1}^n \left(\prod_{j=1}^{i-1} A_j \right) B_i,$$

where

- ▶ $(A_n, B_n)_{n \in \mathbb{N}}$, sequence of random vectors in \mathbb{R}^2 , modulated by
- ▶ $(M_n)_{n \in \mathbb{N}_0}$, ergodic Markov chain with countable state space \mathcal{S} and stationary law π .

Methodology II

Alsmeyer & Buckmann (2017) generalized the results from Goldie & Maller (2000) to a Markovian environment:

They provide necessary and sufficient conditions for convergence as $n \rightarrow \infty$ of

$$Z_n := \sum_{i=1}^n \left(\prod_{j=1}^{i-1} A_j \right) B_i,$$

where

- ▶ $(A_n, B_n)_{n \in \mathbb{N}}$, sequence of random vectors in \mathbb{R}^2 , modulated by
- ▶ $(M_n)_{n \in \mathbb{N}_0}$, ergodic Markov chain with countable state space \mathcal{S} and stationary law π .

"Modulated" in the sense that

- ▶ conditionally on $M_n = i_n \in \mathcal{S}$, $n = 0, 1, 2, \dots$ the random vectors $(A_1, B_1), (A_2, B_2), \dots$ are independent, and
- ▶ for all $n \in \mathbb{N}$ the conditional law of (A_n, B_n) is temporally homogeneous and depends only on $(i_{n-1}, i_n) \in \mathcal{S}^2$.

Methodology III

We shall

- ▶ find "good" discretizations of $\mathfrak{E}_{\xi,\eta}(t)$ and
- ▶ apply/extend Alsmeyer & Buckmanns results/techniques,

to obtain **necessary and/or sufficient conditions** for convergence of $\mathfrak{E}_{\xi,\eta}(t)$ as $t \rightarrow \infty$.

Results

Define

$$\mathbb{T}_j := \{t \geq 0 : J_t = t\},$$

and

$$\tau_1(j) := \text{first return time to } j, \quad \tau_1^-(j) := \text{first exit time of } j.$$

Main Theorem (B&Sideris, '19):

Results

Define

$$\mathbb{T}_j := \{t \geq 0 : J_t = t\},$$

and

$$\tau_1(j) := \text{first return time to } j, \quad \tau_1^-(j) := \text{first exit time of } j.$$

Main Theorem (B&Sideris, '19):

(i) Assume that $\lim_{t \in \mathbb{T}_j, t \rightarrow \infty} \xi_t = \infty$ for some/all $j \in \mathcal{S}$ and that

$$\int_{(1, \infty)} \frac{\log q}{\int_{(0, \log q]} \mathbb{P}_j(\xi_{\tau_1(j)} > u) du} \mathbb{P}_j \left(\sup_{0 \leq t \leq \tau_1(j)} \left| \int_0^t e^{-\xi_s} d\eta_s \right| \in dq \right) < \infty,$$

for some/all $j \in \mathcal{S}$, then $\mathfrak{E}_{(\xi, \eta)}(t) \rightarrow \mathfrak{E}_{(\xi, \eta)}^\infty \mathbb{P}_\pi$ -a.s. as $t \rightarrow \infty$ for some random variable $\mathfrak{E}_{(\xi, \eta)}^\infty$.

Results

Define

$$\mathbb{T}_j := \{t \geq 0 : J_t = t\},$$

and

$$\tau_1(j) := \text{first return time to } j, \quad \tau_1^-(j) := \text{first exit time of } j.$$

Main Theorem (B&Sideris, '19):

(ii) If $\lim_{t \in \mathbb{T}_j, t \rightarrow \infty} \xi_t = \infty$ for some $j \in \mathcal{S}$ and

$$\int_{(1, \infty)} \frac{\log q}{\int_{(0, \log q]} \mathbb{P}_j(\xi_{\tau_1(j)} > u) du} \mathbb{P}_j \left(\left| \int_{(\tau_1^-(j), \tau_1(j)]} e^{-\xi_s} d\eta_s + Y_{\tau_1^-(j)}^{b, \eta} \right| \in dq \right) < \infty, \quad (2)$$

then there exists a probability measure $Q_j = \mathbb{P}_j(\mathfrak{E}_{(\xi, \eta)}^\infty \in \cdot)$ on \mathbb{R} , such that $\mathfrak{E}_{(\xi, \eta)}(t) \rightarrow \mathfrak{E}_{(\xi, \eta)}^\infty$ in \mathbb{P}_j -probability.

Results

Define

$$\mathbb{T}_j := \{t \geq 0 : J_t = j\},$$

and

$$\tau_1(j) := \text{first return time to } j, \quad \tau_1^-(j) := \text{first exit time of } j.$$

Main Theorem (B&Sideris, '19):

(iii) If $\liminf_{t \in \mathbb{T}_j, t \rightarrow \infty} \xi_t < \infty$ for some $j \in \mathcal{S}$, then either there exists a (unique) sequence $\{c_i, i \in \mathcal{S}\}$ in \mathbb{R} such that

$$\mathfrak{E}_{(\xi, \eta)}(t) = \int_{(0, t]} e^{-\xi_s} d\eta_s = c_{J_0} - c_{J_t} e^{-\xi_t} \quad \mathbb{P}_\pi\text{-a.s.}$$

for all $t \geq 0$, or

$$|\mathfrak{E}_{(\xi, \eta)}(t)| \rightarrow \infty \quad \text{in } \mathbb{P}_\pi\text{-probability.}$$

Results

Define

$$\mathbb{T}_j := \{t \geq 0 : J_t = j\},$$

and

$$\tau_1(j) := \text{first return time to } j, \quad \tau_1^-(j) := \text{first exit time of } j.$$

Main Theorem (B&Sideris, '19):

(iii) If $\liminf_{t \in \mathbb{T}_j, t \rightarrow \infty} \xi_t < \infty$ for some $j \in \mathcal{S}$ (or if (2) fails(?)), then either there exists a (unique) sequence $\{c_i, i \in \mathcal{S}\}$ in \mathbb{R} such that

$$\mathfrak{E}_{(\xi, \eta)}(t) = \int_{(0, t]} e^{-\xi_s} d\eta_s = c_{J_0} - c_{J_t} e^{-\xi_t} \quad \mathbb{P}_\pi\text{-a.s.}$$

for all $t \geq 0$, or

$$|\mathfrak{E}_{(\xi, \eta)}(t)| \rightarrow \infty \quad \text{in } \mathbb{P}_\pi\text{-probability.}$$

Degeneracy

Degeneracy

Recall: Degeneracy in the Lévy case $\mathcal{S} = \{1\}$ is characterized by

$$\mathbb{P} \left(\int_0^t e^{-\xi_s} d\eta_s = k(1 - e^{-\xi_t}) \quad \text{for all } t > 0 \right) = 1$$

Degeneracy

Recall: Degeneracy in the Lévy case $\mathcal{S} = \{1\}$ is characterized by

$$\mathbb{P} \left(\int_0^t e^{-\xi_s} d\eta_s = k(1 - e^{-\xi_t}) \quad \text{for all } t > 0 \right) = 1$$

For $\mathfrak{E}_{\xi, \eta}$ we observe: $\mathfrak{E}_{\xi, \eta}$ is degenerate iff there exists a (unique) sequence $\{c_i, i \in \mathcal{S}\}$ in \mathbb{R} such that

$$\mathfrak{E}_{(\xi, \eta)}(t) = \int_{(0, t]} e^{-\xi_s} d\eta_s = c_{J_0} - c_{J_t} e^{-\xi_t} \quad \mathbb{P}_\pi\text{-a.s. for all } t \geq 0. \quad (3)$$

Degeneracy

Recall: Degeneracy in the Lévy case $\mathcal{S} = \{1\}$ is characterized by

$$\mathbb{P} \left(\int_0^t e^{-\xi_{s-}} d\eta_s = k(1 - e^{-\xi t}) \quad \text{for all } t > 0 \right) = 1$$

For $\mathfrak{E}_{\xi, \eta}$ we observe: $\mathfrak{E}_{\xi, \eta}$ is degenerate iff there exists a (unique) sequence $\{c_i, i \in \mathcal{S}\}$ in \mathbb{R} such that

$$\mathfrak{E}_{(\xi, \eta)}(t) = \int_{(0, t]} e^{-\xi_{s-}} d\eta_s = c_{J_0} - c_{J_t} e^{-\xi t} \quad \mathbb{P}_\pi\text{-a.s. for all } t \geq 0. \quad (3)$$

Further:

Proposition (B&Sideris, '19+) Assume (3) holds for all $t \geq 0$ and some sequence $\{c_i, i \in \mathcal{S}\}$. Then

$$\eta_t = - \int_{(0, t]} c_{J_{s-}} dU_s - \int_{(0, t]} dc_{J_s}, \quad t \geq 0, \quad \mathbb{P}_\pi\text{-a.s.} \quad (4)$$

where $(U_t)_{t \geq 0} = (\text{Log}(e^{-\xi t}))_{t \geq 0}$.

Conversely, if (4) holds for some sequence $\{c_i, i \in \mathcal{S}\}$, then (3) is fulfilled for all $t \geq 0$.

Degeneracy, continued

$$\eta_t = - \int_{(0,t]} c_{J_s-} dU_s - \int_{(0,t]} dc_{J_s}, \quad t \geq 0, \mathbb{P}_\pi\text{-a.s.}$$

implies

Degeneracy, continued

$$\eta_t = - \int_{(0,t]} c_{J_s^-} dU_s - \int_{(0,t]} dc_{J_s}, \quad t \geq 0, \mathbb{P}_\pi\text{-a.s.}$$

implies

- ▶ If $(X_t^{(2)}, Y_t^{(2)}) \equiv 0$ (that is $\Delta\xi_{T_n} = 0 = \Delta\eta_{T_n}$ \mathbb{P}_π -a.s. for all n), then $c_i \equiv c$, i.e. $\eta_t = -cU_t$ and

$$\mathfrak{E}_{(\xi,\eta)}(t) = \int_{(0,t]} e^{-\xi_{s-}} d\eta_s = c(1 - e^{-\xi_t}) \quad \mathbb{P}_\pi\text{-a.s. for all } t \geq 0.$$

Degeneracy, continued

$$\eta_t = - \int_{(0,t]} c_{J_s^-} dU_s - \int_{(0,t]} dc_{J_s}, \quad t \geq 0, \mathbb{P}_\pi\text{-a.s.}$$

implies

- ▶ If $(X_t^{(2)}, Y_t^{(2)}) \equiv 0$ (that is $\Delta\xi_{T_n} = 0 = \Delta\eta_{T_n}$ \mathbb{P}_π -a.s. for all n), then $c_i \equiv c$, i.e. $\eta_t = -cU_t$ and

$$\mathfrak{E}_{(\xi,\eta)}(t) = \int_{(0,t]} e^{-\xi_{s-}} d\eta_s = c(1 - e^{-\xi t}) \quad \mathbb{P}_\pi\text{-a.s. for all } t \geq 0.$$

- ▶ If $\Delta\xi_{T_n} = 0$ \mathbb{P}_π -a.s. for all n , then $Z_\eta^{(i,j)} = c_i - c_j$.

Degeneracy, continued

$$\eta_t = - \int_{(0,t]} c_{J_s^-} dU_s - \int_{(0,t]} dc_{J_s}, \quad t \geq 0, \mathbb{P}_\pi\text{-a.s.}$$

implies

- ▶ If $(X_t^{(2)}, Y_t^{(2)}) \equiv 0$ (that is $\Delta\xi_{T_n} = 0 = \Delta\eta_{T_n}$ \mathbb{P}_π -a.s. for all n), then $c_i \equiv c$, i.e. $\eta_t = -cU_t$ and

$$\mathfrak{E}_{(\xi,\eta)}(t) = \int_{(0,t]} e^{-\xi_{s-}} d\eta_s = c(1 - e^{-\xi t}) \quad \mathbb{P}_\pi\text{-a.s. for all } t \geq 0.$$

- ▶ If $\Delta\xi_{T_n} = 0$ \mathbb{P}_π -a.s. for all n , then $Z_\eta^{(i,j)} = c_i - c_j$.
- ▶ If $\Delta\eta_{T_n} = 0$ \mathbb{P}_π -a.s. for all n , then $Z_\xi^{(i,j)} = \log \frac{c_j}{c_i}$.

Although we can state

necessary and sufficient conditions

for convergence of $\mathfrak{E}_{(\xi,\eta)}(t)$, these are hardly applicable.

Splitting

To obtain easier (sufficient) conditions, we split the exponential functional in two pieces:

$$\begin{aligned}\mathfrak{E}_{(\xi,\eta)}(t) &= \int_{(0,t]} e^{-\xi_s} d(\gamma_s^\eta + W_s^\eta + Y_s^{b,\eta} + Y_s^{s,\eta} + Y_s^{(2)}) \\ &= \int_{(0,t]} e^{-\xi_s} d(\underbrace{\gamma_s^\eta}_{\text{drift}} + \underbrace{W_s^\eta + Y_s^{s,\eta}}_{\text{martingale part}}) + \int_{(0,t]} e^{-\xi_s} d(\underbrace{Y_s^{b,\eta} + Y_s^{(2)}}_{\text{"big" jumps}}) \\ &=: \mathfrak{E}^{(1)}(t) + \mathfrak{E}^{(2)}(t).\end{aligned}$$

Splitting

To obtain easier (sufficient) conditions, we split the exponential functional in two pieces:

$$\begin{aligned}\mathfrak{E}_{(\xi, \eta)}(t) &= \int_{(0, t]} e^{-\xi_s} d(\gamma_s^\eta + W_s^\eta + Y_s^{b, \eta} + Y_s^{s, \eta} + Y_s^{(2)}) \\ &= \int_{(0, t]} e^{-\xi_s} d(\underbrace{\gamma_s^\eta}_{\text{drift}} + \underbrace{W_s^\eta + Y_s^{s, \eta}}_{\text{martingale part}}) + \int_{(0, t]} e^{-\xi_s} d(\underbrace{Y_s^{b, \eta} + Y_s^{(2)}}_{\text{"big" jumps}}) \\ &=: \mathfrak{E}^{(1)}(t) + \mathfrak{E}^{(2)}(t).\end{aligned}$$

- ▶ $\mathfrak{E}^{(1)}(t)$ converges a.s. under rather weak conditions.

Splitting

To obtain easier (sufficient) conditions, we split the exponential functional in two pieces:

$$\begin{aligned}\mathfrak{E}_{(\xi,\eta)}(t) &= \int_{(0,t]} e^{-\xi_s} d(\gamma_s^\eta + W_s^\eta + Y_s^{b,\eta} + Y_s^{s,\eta} + Y_s^{(2)}) \\ &= \int_{(0,t]} e^{-\xi_s} d(\underbrace{\gamma_s^\eta}_{\text{drift}} + \underbrace{W_s^\eta + Y_s^{s,\eta}}_{\text{martingale part}}) + \int_{(0,t]} e^{-\xi_s} d(\underbrace{Y_s^{b,\eta} + Y_s^{(2)}}_{\text{"big" jumps}}) \\ &=: \mathfrak{E}^{(1)}(t) + \mathfrak{E}^{(2)}(t).\end{aligned}$$

- ▶ $\mathfrak{E}^{(1)}(t)$ converges a.s. under rather weak conditions.
- ▶ $\mathfrak{E}^{(2)}(t)$ can be embedded in a discrete time setting and directly corresponds to a Markov modulated perpetuity.

The characteristic κ_ξ

Recall: Lévy processes in \mathbb{R} either drift to $\pm\infty$ or oscillate.

The characteristic κ_ξ

Recall: Lévy processes in \mathbb{R} either drift to $\pm\infty$ or oscillate.

MAPs often show a similar behaviour:

Define the **long-term mean** (here for the MAP $(\xi_t, J_t)_{t \geq 0}$)

$$\kappa_\xi := \sum_{j \in \mathcal{S}} \pi_j \left(\gamma_{\xi^{(j)}} + \int_{|x| \geq 1} x \nu_{\xi^{(j)}}(dx) \right) + \sum_{\substack{(i,j) \in \mathcal{S} \times \mathcal{S} \\ i \neq j}} \pi_i q_{i,j} \int_{\mathbb{R}} x F_\xi^{(i,j)}(dx).$$

Whenever \mathcal{S} is finite, κ_ξ fully determines the long-term behaviour of ξ :

$$\kappa_\xi > 0 \quad \Rightarrow \quad \lim_{t \rightarrow \infty} \xi_t = \infty \quad \mathbb{P}_\pi\text{-a.s.},$$

$$\kappa_\xi < 0 \quad \Rightarrow \quad \lim_{t \rightarrow \infty} \xi_t = -\infty \quad \mathbb{P}_\pi\text{-a.s.}, \quad \text{while}$$

$$\mathbb{P}_\pi \left(\left. \begin{array}{l} \kappa_\xi = 0 \text{ and} \\ \sup_{t \geq 0} |\xi_t| < \infty \end{array} \right\} < 1 \right) \Rightarrow \left\{ \begin{array}{l} \limsup_{t \rightarrow \infty} \xi_t = \infty \text{ and} \\ \liminf_{t \rightarrow \infty} \xi_t = -\infty \end{array} \right\} \quad \mathbb{P}_\pi\text{-a.s.}$$

Proposition (B&Sideris, '19): Assume that $0 < \kappa_\xi < \infty$ and

$$\sup_{j \in \mathcal{S}} \left(|\gamma_{\eta^{(j)}}| + \sigma_{\eta^{(j)}}^2 + \int_{(0,1)} x^2 \nu_{\eta^{(j)}}(dx) \right) < \infty. \quad (5)$$

Then $\mathfrak{E}^{(1)}(t)$ converges \mathbb{P}_π -a.s. to a finite random variable as $t \rightarrow \infty$.

Proposition (B&Sideris, '19): Assume that $0 < \kappa_{\xi} < \infty$ and

$$\sup_{j \in \mathcal{S}} \left(|\gamma_{\eta^{(j)}}| + \sigma_{\eta^{(j)}}^2 + \int_{(0,1)} x^2 \nu_{\eta^{(j)}}(dx) \right) < \infty. \quad (5)$$

Then $\mathfrak{E}^{(1)}(t)$ converges \mathbb{P}_{π} -a.s. to a finite random variable as $t \rightarrow \infty$.

Note that

- ▶ For \mathcal{S} finite, (5) is always fulfilled.

Proposition (B&Sideris, '19): Assume that $0 < \kappa_\xi < \infty$ and

$$\sup_{j \in \mathcal{S}} \left(|\gamma_{\eta^{(j)}}| + \sigma_{\eta^{(j)}}^2 + \int_{(0,1)} x^2 \nu_{\eta^{(j)}}(dx) \right) < \infty. \quad (5)$$

Then $\mathfrak{E}^{(1)}(t)$ converges \mathbb{P}_π -a.s. to a finite random variable as $t \rightarrow \infty$.

Note that

- ▶ For \mathcal{S} finite, (5) is always fulfilled.
- ▶ In general (5) is necessary.

A result for $\mathfrak{E}^{(2)}$

Proposition (B&Sideris, '19): Assume $0 < \kappa_\xi < \infty$.

1. $\mathfrak{E}^{(2)}(t)$ converges \mathbb{P}_π -a.s. to a finite random variable as $t \rightarrow \infty$ if and only if for some/all $j \in \mathcal{S}$

$$\int_{(1,\infty)} \log q \mathbb{P}_j \left(\sup_{0 < t \leq \tau_1(j)} e^{-\xi t} |\Delta(Y_t^{b,\eta} + Y_t^{(2)})| \in dq \right) < \infty.$$

2. $\mathfrak{E}^{(2)}(t)$ converges in \mathbb{P}_j -probability to some random variable $\mathfrak{E}_\infty^{(2)}$ for all $j \in \mathcal{S}$, if and only if for some/all $j \in \mathcal{S}$

$$\int_{(1,\infty)} \log q \mathbb{P}_j \left(\left| \int_{(0,\tau_1(i))} e^{-\xi t} d(Y_t^{b,\eta} + Y_t^{(2)}) \right| \in dq \right) < \infty.$$

Open questions/Outlook

Open questions/Outlook

- ▶ Fluctuation theory for MAPs with countable state space?

Open questions/Outlook

- ▶ Fluctuation theory for MAPs with countable state space?
- ▶ \mathcal{E} as stationary distribution of a Markov modulated generalized Ornstein-Uhlenbeck process?

**Thank you
for your attention!**

Main References

- ▶ **K. B. Erickson and R. A. Maller (2005)**
Generalised Ornstein-Uhlenbeck processes and the convergence of Lévy integrals.
Séminaire de Probabilités XXXVIII.
- ▶ **G. Alsmeyer and F. Buckmann (2017)**
Stability of perpetuities in Markovian environment.
J. Difference Equ. Appl.
- ▶ **A. Behme and A. Sideris (2019+)**
Exponential functionals of Markov additive processes (working title).
To be submitted soon.