

# Distributional Representations & Dominance of a Lévy Process over its Maximal Jump Processes

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- joint work with  
Yuguang Fan (School of Mathematics & Statistics, Uni Melbourne)  
Ross.A. Maller (RSFAS, ANU):
- talk surveys  
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Maximal Jump Processes. *Bernoulli* **22**(4), 2325–2371.

## intro

- we study relations between a Lévy process  $X = (X_t)_{t \geq 0}$ , its quadratic variation process  $V = (V_t)_{t \geq 0}$ , and its maximal jump processes, with particular interest in how these processes, and how positive and negative parts of the  $X$  process, interact.
- ratio of the process to its extremes in the random walk situation: Darling (1952), Arov & Bobrov (1960), Maller & Resnick (1984), Kesten & Maller ((1992), (1994)); almost sure versions of sum/max relationships, see Feller (1968), Kesten & Maller (1995), Pruitt (1987).
- trimmed sums concerning heavy tailed distributions: Csörgő, Haeusler & Mason (1988), Berkes & Horváth (2010), Berkes, Horváth & Schauer (2010), and Griffin & Pruitt (2013); Silvestrov & Teugels (2002) concerns sums and maxima of random walks and triangular arrays; Ladoucette & Teugels (2013) for an insurance application; connections to St. Petersburg game: Gut & Martin-Löf (2014) give a “maxtrimmed” version of the game, while Fukker, Györfi & Kevei (2015).

- Relevant to our topic, includes that of Doney (2004), Andrew (2008), Bertoin (1997), Doney (2007)
- identities allow for possible ties in the order statistics of the jumps. point process versions are motivated by LePage (1980, 1981), LePage, Woodroffe & Zinn (1981), Mori (1984) for trimmed sums, Khinthine's inverse Lévy measure method (1937),... Rosiński (2001) summarises alternative series representations for Lévy processes.

## notation

- let  $X = (X_t)_{t \geq 0}$  be a real-valued Lévy process with canonical triplet  $(\gamma, \sigma^2, \Pi)$  and characteristic function  $Ee^{i\theta X_t} = e^{t\Psi(\theta)}$ ,  $t \geq 0$ ,  $\theta \in \mathbb{R}$ , with characteristic exponent

$$\Psi(\theta) := i\theta\gamma - \frac{1}{2}\sigma^2\theta^2 + \int_{\mathbb{R}_*} (e^{i\theta x} - 1 - i\theta x \mathbf{1}_{\{|x| \leq 1\}}) \Pi(dx)$$

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- measures on  $(0, \infty)$ :  $\Pi^{(+)}$  is  $\Pi$  restricted to  $(0, \infty)$ ,  $\Pi^{(-)}$  is  $\Pi(-\cdot)$  restricted to  $(0, \infty)$ , and  $\Pi^{|\cdot|} := \Pi^{(+)} + \Pi^{(-)}$   
 $\Delta\Pi(y) := \Pi\{\{y\}\}$ ,  $y \in \mathbb{R}_*$ , and  $\Delta\bar{\Pi}(y) := \bar{\Pi}(y-) - \bar{\Pi}(y)$ ,  $y > 0$

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- positive, negative and two-sided tails of  $\Pi$  are

$$\bar{\Pi}^+(x) := \Pi\{(x, \infty)\}, \quad \bar{\Pi}^-(x) := \Pi\{(-\infty, -x)\}, \quad \bar{\Pi}(x) := \bar{\Pi}^+(x) + \bar{\Pi}^-(x),$$

$\bar{\Pi}^{\leftarrow}(x) = \inf\{y > 0 : \bar{\Pi}(y) \leq x\}$ ,  $x > 0$ , denotes the right-continuous inverse of the nonincreasing function

- interested in small time behaviour of  $X_t$ , so assume  $\bar{\Pi}(0+) = \infty$   
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- one-sided and modulus trimmed versions of  $X$  are defined as

$${}^{(r)}X_t := X_t - \sum_{i=1}^r \Delta X_t^{(i)} \quad \text{and} \quad {}^{(r)}\widetilde{X}_t := X_t - \sum_{i=1}^r \widetilde{\Delta X}_t^{(i)},$$

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- introduce two families of processes, indexed by  $v > 0$ , truncating jumps from sample paths of  $X_t$  and  $V_t$ , respectively. let  $v, t > 0$ . if  $\bar{\Pi}(0+) = \infty$ , then set

$$\widetilde{X}_t^v := X_t - \sum_{0 < s \leq t} \Delta X_s \mathbf{1}_{\{|\Delta X_s| \geq \bar{\Pi}^{\leftarrow}(v)\}}$$

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- introduce two families of processes, indexed by  $\nu > 0$ , truncating jumps from sample paths of  $X_t$  and  $V_t$ , respectively. let  $\nu, t > 0$ . if  $\bar{\Pi}(0+) = \infty$ , then set

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- if  $\bar{\Pi}^+(0+) = \infty$ , then set

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## representation theorem

assume independent  $X$ ,  $\mathfrak{G}_r$ ,  $Y^+$ ,  $Y^-$ , and  $Y$ ,  
where  $r \in \mathbb{N}$ ,  $\mathfrak{G}_r \sim \text{Gamma}(r, 1)$ ,  $Y^\pm = (Y_t^\pm)_{t \geq 0}$  and  $Y = (Y_t)_{t \geq 0}$   
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(i) if  $\bar{\Pi}^+(0+) = \infty$ , for  $t, v > 0$ , let  $\kappa(v) := \bar{\Pi}^+(\bar{\Pi}^{+, \leftarrow}(v)-) - v$ ,  
and  $G_t^v := \bar{\Pi}^{+, \leftarrow}(v) Y_{t\kappa(v)}$  then, for  $t > 0$ , we have

$$\binom{(r)}{X_t, \Delta X_t^{(r)}} \stackrel{D}{=} (X_t^v + G_t^v, \bar{\Pi}^{+, \leftarrow}(v)) \Big|_{v=\mathfrak{G}_r/t}.$$

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$$\kappa^\pm(v) := (\bar{\Pi}(\bar{\Pi}^{\leftarrow}(v) -) - v) \frac{\Delta \bar{\Pi}(\pm \bar{\Pi}^{\leftarrow}(v))}{\Delta \bar{\Pi}(\bar{\Pi}^{\leftarrow}(v))} \mathbf{1}_{\{\Delta \bar{\Pi}(\bar{\Pi}^{\leftarrow}(v)) \neq 0\}}$$

$$\tilde{G}_t^v := \bar{\Pi}^{\leftarrow}(v) (Y_{t\kappa^+(v)}^+ - Y_{t\kappa^-(v)}^-)$$

then, for each  $t > 0$ , we have

$$\binom{(r)}{\tilde{X}_t, |\widetilde{\Delta X}_t^{(r)}|} \stackrel{D}{=} (\tilde{X}_t^v + \tilde{G}_t^v, \bar{\Pi}^{\leftarrow}(v)) \Big|_{v=\mathfrak{G}_r/t}$$

## order statistics with ties

let  $\mathbb{X}$  be a Poisson point process (PPP) on  $[0, \infty) \times \mathbb{R}_*$  with intensity measure  $ds \otimes \Pi(dx)$

$$\mathbb{X} = \sum_s \delta_{(s, \Delta X_s)}$$



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PPPs of positive and negative jumps and jumps in modulus with intensity measures  $ds \otimes \Pi^{\pm, |\cdot|}(dx)$ , respectively,

$$\mathbb{X}^{\pm} = \sum_s \mathbf{1}_{(0, \infty)}(\pm \Delta X_s) \delta_{(s, \pm \Delta X_s)}$$

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restrict processes to the time interval  $s \in [0, t]$ ,  $t > 0$ ,

$$\mathbb{X}_t(\cdot) := \mathbb{X}([0, t] \times \mathbb{R}_* \cap \cdot) \quad \text{and} \quad \mathbb{X}_t^{\pm, |\cdot|}(\cdot) = \mathbb{X}^{\pm, |\cdot|}([0, t] \times (0, \infty) \cap \cdot)$$

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- let  $\tilde{T}^{(1)}(\mathbb{X}_t)$  be randomly chosen, independently of  $(X_t)_{t \geq 0}$ , according to the discrete uniform distribution,

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- let  $r = 2, 3, \dots$  iteratively, define  $\tilde{T}^{(r)}(\mathbb{X}_t) := \tilde{T}^{(1)}({}^{(r-1)}\widetilde{\mathbb{X}}_t)$  and  $\widetilde{\Delta X}_t^{(r)} := \Delta X_{\tilde{T}^{(r)}(\mathbb{X}_t)}$ , and the  $r$ -fold trimmed extremal process of modulus jumps is then defined by

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## order statistics with ties

similarly,

- assuming  $\bar{\Pi}^+(0+) = \infty$ , introduce in  $[0, t] \times (0, \infty)$ ,

$$(T^{(1)}(\mathbb{X}_t^+), \Delta X_t^{(1)}), (T^{(2)}(\mathbb{X}_t^+), \Delta X_t^{(2)}), (T^{(3)}(\mathbb{X}_t^+), \Delta X_t^{(3)}), \dots$$

such that  $\Delta X_t^{(1)} \geq \dots \geq \Delta X_t^{(r)}$  are the  $r^{\text{th}}$  largest order statistics of positive jumps of  $X$  sampled on time interval  $[0, t]$

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- subtract points corresponding to large jumps, introduce the *r-fold trimmed extremal process of positive jumps* by

$${}^{(r)}\mathbb{X}_t^+ := \mathbb{X}_t^+ - \sum_{1 \leq i \leq r} \delta_{(T^{(i)}(\mathbb{X}_t^+), \Delta X_t^{(i)})}$$

## alternative representation

- let  $(\mathcal{U}_i)$ ,  $(\mathcal{U}'_i)$  and  $(\mathcal{E}_i)$  be independent, where  $(\mathcal{U}_i)$  and  $(\mathcal{U}'_i)$  are iid sequences of uniform rvs in  $(0, 1)$ ,  $(\mathcal{E}_i)$  is i.i.d. sequence of exponential rvs with parameter  $E\mathcal{E}_i = 1$ .

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- $\mathcal{G}_r = \sum_{i=1}^r \mathcal{E}_i$  is standard Gamma random walk,  $r \in \mathbb{N}$ .
- for  $t > 0$

$$\mathbb{V}_t := \sum_{i \geq 1} \delta_{(t\mathcal{U}_i, \mathcal{G}_i/t)} \quad \text{and} \quad \mathbb{V}'_t := \sum_{i \geq 1} \delta_{(t\mathcal{U}_i, \mathcal{U}'_i, \mathcal{G}_i/t)}$$

$\mathbb{V}_t$  and  $\mathbb{V}'_t$  are homogeneous PPPs on  $[0, t] \times (0, \infty)$  and  $[0, t] \times (0, 1) \times (0, \infty)$  with intensity measures  $ds \otimes dv$  and  $ds \otimes du' \otimes dv$ , respectively.

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- for  $r \in \mathbb{N}_0 := \{0, 1, 2, \dots\}$ , introduce  $r$ -fold trimmed counterparts:

$${}^{(r)}\mathbb{V}_t := \sum_{i > r} \delta_{(t\mathcal{U}_i, \mathfrak{S}_i/t)} \quad \text{and} \quad {}^{(r)}\mathbb{V}'_t := \sum_{i > r} \delta_{(t\mathcal{U}_i, \mathcal{U}'_i, \mathfrak{S}_i/t)}$$

# transformations for $\bar{\Pi}^+(0+) = \infty$

- if  $\bar{\Pi}^+(0+) = \infty$  then transform

$$(I, \bar{\Pi}^{+, \leftarrow}) : [0, t] \times (0, \infty) \rightarrow [0, \infty) \times (0, \infty)$$

such that

$$\begin{aligned} \mathbb{V}_t^{(I, \bar{\Pi}^{+, \leftarrow})} &:= \sum_{i \geq 1} \delta_{(t \mathbb{U}_i, \bar{\Pi}^{+, \leftarrow}(\mathfrak{S}_i/t))} \\ {}^{(r)}\mathbb{V}_t^{(I, \bar{\Pi}^{+, \leftarrow})} &:= \sum_{i > r} \delta_{(t \mathbb{U}_i, \bar{\Pi}^{+, \leftarrow}(\mathfrak{S}_i/t))}. \end{aligned}$$

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- with Radon-Nikodym  $d\Pi^{\pm} = g^{\pm}d\Pi^{|\cdot|}$  write

$$g^{\pm} : (0, \infty) \rightarrow (0, \infty) \text{ with } g^{+} + g^{-} \equiv 1 \text{ and } d\Pi^{\pm} = g^{\pm}d\Pi^{|\cdot|}$$

returning signs by  $m : [0, t] \times (0, 1) \times (0, \infty) \rightarrow [0, t] \times \mathbb{R}_{*}$

$$m(s, u', x) := \begin{cases} (s, x), & \text{if } u' < g^{+}(x) \\ (s, -x), & \text{if } u' \geq g^{+}(x) \end{cases}$$

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$$m(s, u', x) := \begin{cases} (s, x), & \text{if } u' < g^+(x) \\ (s, -x), & \text{if } u' \geq g^+(x) \end{cases}$$

- map:

$$\mathbb{V}'_t \xrightarrow{(I, I, \bar{\Pi}^{\leftarrow})} \mathbb{V}'_t(I, I, \bar{\Pi}^{\leftarrow}) := \sum_{i \geq 1} \delta_{(t \mathbb{1}_i, \mathbb{1}'_i, \bar{\Pi}^{\leftarrow}(\mathfrak{S}_i/t))}$$

$$\xrightarrow{m} \mathbb{V}'_t m \circ (I, I, \bar{\Pi}^{\leftarrow}) := \sum_{i \geq 1} \delta_{m(t \mathbb{1}_i, \mathbb{1}'_i, \bar{\Pi}^{\leftarrow}(\mathfrak{S}_i/t))}$$

## lemma

let  $t > 0$  and  $r \in \mathbb{N}$ .

(i) if  $\bar{\Pi}^+(0+) = \infty$ , then

$$\begin{aligned} \mathbb{X}_t^+ &\stackrel{D}{=} \mathbb{V}_t^{(l, \bar{\Pi}^{+, \leftarrow})} \\ (T^{(i)}(\mathbb{X}_t^+), \Delta X_t^{(i)})_{i \geq 1} &\stackrel{D}{=} (t\mathcal{U}_i, \bar{\Pi}^{+, \leftarrow}(\mathfrak{S}_i/t))_{i \geq 1} \\ \{(T^{(i)}(\mathbb{X}_t^+), \Delta X_t^{(i)})_{1 \leq i \leq r}, {}^{(r)}\mathbb{X}_t^+\} &\stackrel{D}{=} \\ &\{(t\mathcal{U}_i, \bar{\Pi}^{+, \leftarrow}(\mathfrak{S}_i/t))_{1 \leq i \leq r}, {}^{(r)}\mathbb{V}_t^{(l, \bar{\Pi}^{+, \leftarrow})}\}. \end{aligned}$$

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(ii) if  $\bar{\Pi}(0+) = \infty$ , then

$$\begin{aligned}\mathbb{X}_t &\stackrel{D}{=} \mathbb{V}_t^{m_0(l, l, \bar{\Pi}^{\leftarrow})} \\ (\tilde{T}^{(i)}(\mathbb{X}_t), \widetilde{\Delta X}_t^{(i)})_{i \geq 1} &\stackrel{D}{=} (m(t\mathfrak{U}_i, \mathfrak{U}'_i, \bar{\Pi}^{\leftarrow}(\mathfrak{G}_i/t)))_{i \geq 1} \\ \{(\tilde{T}^{(i)}(\mathbb{X}_t), \widetilde{\Delta X}_t^{(i)})_{1 \leq i \leq r}, {}^{(r)}\tilde{\mathbb{X}}_t\} &\stackrel{D}{=} \\ &\{(m(t\mathfrak{U}_i, \mathfrak{U}'_i, \bar{\Pi}^{\leftarrow}(\mathfrak{G}_i/t)))_{1 \leq i \leq r}, {}^{(r)}\mathbb{V}_t^{m_0(l, l, \bar{\Pi}^{\leftarrow})}\}.\end{aligned}$$

## representation theorem (PPP)

assume  $\mathbb{X}$ ,  $(\mathcal{U}_i)$ ,  $(\mathcal{U}'_i)$ ,  $\mathfrak{G}_r$ ,  $Y^\pm = (Y^\pm(t))_{t \geq 0}$ ,  $Y = (Y(t))_{t \geq 0}$ , are independent, with  $Y^\pm$  and  $Y$  standard Poisson processes.

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for  $x > 0$  delete points in  $\mathbb{X}_t^+$  and  $\mathbb{X}_t$  not lying in  $[0, t] \times (0, x)$  and  $[0, t] \times (-x, x)_*$ :

$\mathbb{X}_t^{+ \cdot < x} := \mathbb{X}^+([0, t] \times (0, x) \cap \cdot)$  and  $\mathbb{X}_t^{|\cdot| < x} := \mathbb{X}([0, t] \times (-x, x)_* \cap \cdot)$ .

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(i) if  $\bar{\Pi}^+(0+) = \infty$ , then for all  $t > 0$ ,  $r \in \mathbb{N}$ ,

$$\kappa(v) = \bar{\Pi}^+(\bar{\Pi}^{+, \leftarrow}(v) -) - v$$

$$(\Delta X_t^{(r)}, {}^{(r)}\mathbb{X}_t^+) \stackrel{D}{=} (\bar{\Pi}^{+, \leftarrow}(v), \mathbb{X}_t^{+\cdot < \bar{\Pi}^{+, \leftarrow}(v)} + \sum_{i=1}^{Y(t\kappa(v))} \delta_{(t\mathfrak{U}_i, \bar{\Pi}^{+, \leftarrow}(v))})_{v \in \mathfrak{G}_r/t}$$

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(ii) if  $\bar{\Pi}(0+) = \infty$  then, for all  $t > 0$ ,  $r \in \mathbb{N}$ ,

$$(|\widetilde{\Delta X}_t^{(r)}|, {}^{(r)}\widetilde{\mathbb{X}}_t)$$

$$\stackrel{D}{=} (\bar{\Pi}^{\leftarrow}(v), \mathbb{X}_t^{|\cdot| < \bar{\Pi}^{\leftarrow}(v)} + \sum_{i=1}^{Y^+(t\kappa^+(v))} \delta_{(t\mathbb{U}_i, \bar{\Pi}^{\leftarrow}(v))} + \sum_{i=1}^{Y^-(t\kappa^-(v))} \delta_{(t\mathbb{U}'_i, -\bar{\Pi}^{\leftarrow}(v))})_{v \in \mathfrak{G}_r/t}$$



## representation theorem (LP revisited)

assume independent  $X$ ,  $\mathfrak{G}_r$ ,  $Y^+$ ,  $Y^-$ , and  $Y$ ,  
 where  $r \in \mathbb{N}$ ,  $\mathfrak{G}_r \sim \text{Gamma}(r, 1)$ ,  $Y^\pm = (Y_t^\pm)_{t \geq 0}$  and  $Y = (Y_t)_{t \geq 0}$   
 are independent Poisson processes with  $EY_1^\pm = EY_1 = 1$ .

(i) if  $\bar{\Pi}^+(0+) = \infty$ , for  $t, v > 0$ , let  $\kappa(v) := \bar{\Pi}^+(\bar{\Pi}^{+, \leftarrow}(v) -) - v$ ,  
 and  $G_t^v := \bar{\Pi}^{+, \leftarrow}(v) Y_{t\kappa(v)}$  then, for  $t > 0$ , we have

$$\binom{(r)}{X_t, \Delta X_t^{(r)}} \stackrel{D}{=} (X_t^v + G_t^v, \bar{\Pi}^{+, \leftarrow}(v)) \Big|_{v=\mathfrak{G}_r/t}$$

(ii) if  $\bar{\Pi}(0+) = \infty$ , then, for  $t, v > 0$ , let

$$\kappa^\pm(v) := (\bar{\Pi}(\bar{\Pi}^{\leftarrow}(v) -) - v) \frac{\Delta \bar{\Pi}(\pm \bar{\Pi}^{\leftarrow}(v))}{\Delta \bar{\Pi}(\bar{\Pi}^{\leftarrow}(v))} \mathbf{1}_{\{\Delta \bar{\Pi}(\bar{\Pi}^{\leftarrow}(v)) \neq 0\}}$$

$$\tilde{G}_t^v := \bar{\Pi}^{\leftarrow}(v) (Y_{t\kappa^+(v)}^+ - Y_{t\kappa^-(v)}^-)$$

then, for each  $t > 0$ , we have

$$\binom{(r)}{\tilde{X}_t, |\widetilde{\Delta X}_t^{(r)}|} \stackrel{D}{=} (\tilde{X}_t^v + \tilde{G}_t^v, \bar{\Pi}^{\leftarrow}(v)) \Big|_{v=\mathfrak{G}_r/t}$$

## $X$ comparable with its large jump processes

**theorem** suppose  $\sigma^2 = 0$  and  $\bar{\Pi}(0+) = \infty$ . then

$$\frac{X_t}{\widetilde{\Delta X}_t^{(1)}} \xrightarrow{P} 1, \text{ as } t \downarrow 0,$$

iff  $\bar{\Pi}(x) \in SV$  at 0 (so that  $X$  is of bounded variation) and  $X$  has drift 0. These imply

$$(*) \quad \frac{|\widetilde{\Delta X}_t^{(2)}|}{|\widetilde{\Delta X}_t^{(1)}|} \xrightarrow{P} 0, \text{ as } t \downarrow 0;$$

and conversely  $(*)$  implies  $\bar{\Pi}(x) \in SV$  at 0

# $X$ comparable with its large jump processes-(one-sided version)

**theorem** suppose  $\bar{\Pi}^+(0+) = \infty$ . then

$$\frac{X_t}{\Delta X_t^{(1)}} \xrightarrow{P} 1, \text{ as } t \downarrow 0$$

iff  $\bar{\Pi}^+(x) \in SV$  at 0,  $X$  is of bounded variation with drift 0, and  $\lim_{x \downarrow 0} \bar{\Pi}^-(x)/\bar{\Pi}^+(x) = 0$ .

## comparing positive & negative jumps

define  $\Delta X_t^+ := \max(\Delta X_t, 0)$ ,  $\Delta X_t^- := \max(-\Delta X_t, 0)$ , and

$$(\Delta X^+)_t^{(1)} := \sup_{0 < s \leq t} \Delta X_s^+ \quad \text{and} \quad (\Delta X^-)_t^{(1)} := \sup_{0 < s \leq t} \Delta X_s^-, \quad t > 0.$$

positive and negative jump processes are independent.

when the integrals are finite, define

$$A_{\pm}(x) := \int_0^x \bar{\Pi}^{\pm}(y) dy = x \int_0^1 \bar{\Pi}^{\pm}(xy) dy.$$

## theorem (comparing positive & negative jumps)

Suppose  $\bar{\Pi}^{\pm}(0+) = \infty$ . For (1) assume  $\sum_{0 < s \leq t} \Delta X_s^-$  is finite a.s., and for (2) assume  $\sum_{0 < s \leq t} \Delta X_s^+$  is finite a.s. For (3), assume both are finite a.s. Then

$$(1) \frac{\sum_{0 < s \leq t} \Delta X_s^-}{\sup_{0 < s \leq t} \Delta X_s^+} \xrightarrow{\mathbb{P}} 0, \text{ as } t \downarrow 0 \text{ if and only if } \lim_{x \downarrow 0} \frac{\int_0^x \bar{\Pi}^-(y) dy}{x \bar{\Pi}^+(x)} = 0;$$

also

$$(2) \frac{\sup_{0 < s \leq t} \Delta X_s^-}{\sum_{0 < s \leq t} \Delta X_s^+} \xrightarrow{\mathbb{P}} 0, \text{ as } t \downarrow 0, \text{ if and only if } \lim_{x \downarrow 0} \frac{x \bar{\Pi}^-(x)}{\int_0^x \bar{\Pi}^+(y) dy} = 0;$$

and

$$(3) \frac{\sum_{0 < s \leq t} \Delta X_s^-}{\sum_{0 < s \leq t} \Delta X_s^+} \xrightarrow{\mathbb{P}} 0, \text{ as } t \downarrow 0, \text{ if and only if } \lim_{x \downarrow 0} \frac{\int_0^x \bar{\Pi}^-(y) dy}{\int_0^x \bar{\Pi}^+(y) dy} = 0.$$

Finally,

$$\frac{\sup_{0 < s \leq t} \Delta X_s^-}{\sup_{0 < s \leq t} \Delta X_s^+} \xrightarrow{\mathbb{P}} 0, \text{ as } t \downarrow 0, \text{ if and only if } \lim_{x \downarrow 0} \frac{\bar{\Pi}^-(\varepsilon x)}{\bar{\Pi}^+(x)} = 0 \text{ for all } \varepsilon > 0$$

## $X$ dominating its large jump processes

versions of truncated first and second moment functions:

$$\nu(x) = \gamma - \int_{x < |y| \leq 1} y \Pi(dy) \text{ and } V(x) = \sigma^2 + \int_{0 < |y| \leq x} y^2 \Pi(dy), \quad x > 0.$$

## $X$ dominating its large jump processes

versions of truncated first and second moment functions:

$$\nu(x) = \gamma - \int_{x < |y| \leq 1} y \Pi(dy) \text{ and } V(x) = \sigma^2 + \int_{0 < |y| \leq x} y^2 \Pi(dy), \quad x > 0.$$

variants of  $\nu(x)$  and  $V(x)$  are Winsorised first and second:

$$A(x) = \gamma + \bar{\Pi}^+(1) - \bar{\Pi}^-(1) - \int_x^1 (\bar{\Pi}^+(y) - \bar{\Pi}^-(y)) dy$$

and

$$U(x) = \sigma^2 + 2 \int_0^x y \bar{\Pi}(y) dy, \text{ for } x > 0.$$

$A(x)$  and  $U(x)$  are continuous for  $x > 0$ .

using Fubini's theorem,

$$A(x) = \nu(x) + x(\bar{\Pi}^+(x) - \bar{\Pi}^-(x))$$

and

$$U(x) = V(x) + x^2(\bar{\Pi}^+(x) + \bar{\Pi}^-(x)) = V(x) + x^2 \bar{\Pi}(x)$$

## thm ( $X$ dominating its large jump processes)

$X$  Staying Positive Near 0 in probability. Suppose  $\bar{\Pi}^+(0+) = \infty$ .

(i) (Doney 2004) if also  $\bar{\Pi}^-(0+) > 0$ , then the following are equivalent:

$$\lim_{t \downarrow 0} P(X_t > 0) = 1; \quad (1)$$

$$\frac{X_t}{(\Delta X^-)_t^{(1)}} \xrightarrow{P} \infty, \text{ as } t \downarrow 0; \quad (2)$$

$$\sigma^2 = 0 \quad \text{and} \quad \lim_{x \downarrow 0} \frac{A(x)}{x \bar{\Pi}^-(x)} = \infty; \quad (3)$$

$$\lim_{x \downarrow 0} \frac{A(x)}{\sqrt{U(x) \bar{\Pi}^-(x)}} = \infty; \quad (4)$$

there is a nonstochastic nondecreasing function  $\ell(x) > 0$ , which is slowly varying at 0, such that

$$\frac{X_t}{t \ell(t)} \xrightarrow{P} \infty, \text{ as } t \downarrow 0. \quad (5)$$



## thm ( $X$ dominating its large jump processes)

(ii) Suppose  $X$  is spectrally positive, so  $\bar{\Pi}^-(x) = 0$  for  $x > 0$ .  
Then (1) is equivalent to

$$\sigma^2 = 0 \text{ and } A(x) \geq 0 \text{ for all small } x, \quad (6)$$

and this happens if and only if  $X$  is a subordinator. Furthermore,  
we then have  $A(x) \geq 0$ , not only for small  $x$ , but for all  $x > 0$ .

## relative stability & dominance

assume  $\bar{\Pi}(0+) > 0$ .

- (cf Kallenberg (2002)): relative stability (RS):  $X(t)/b(t) \xrightarrow{P} \pm 1$   
for some measurable nonstochastic function  $b(t) > 0$   
iff

$$\lim_{t \downarrow 0} t\bar{\Pi}(xb_t) = 0, \quad \lim_{t \downarrow 0} \frac{tA(xb_t)}{b_t} = \pm 1, \quad \lim_{t \downarrow 0} \frac{tU(xb_t)}{b_t^2} = 0$$

- (Griffin & Maller (2013)): there is a measurable nonstochastic function  $b_t > 0$  such that

$$\frac{|X_t|}{b_t} \xrightarrow{P} 1, \quad \text{as } t \downarrow 0,$$

iff  $X \in RS$  at 0, equivalently, iff  $X \in RS$  at 0, equivalently, iff

$$\sigma^2 = 0 \quad \text{and} \quad \lim_{x \downarrow 0} \frac{|A(x)|}{x\bar{\Pi}(x)} = \infty$$

## positive relative stability & positive dominance

**theorem** if  $\bar{\Pi}^+(0+) = \infty$ , then the following are equivalent:

$$\frac{X_t}{(\Delta X^+)_t^{(1)}} \xrightarrow{\mathbb{P}} \infty, \text{ as } t \downarrow 0;$$

$$\frac{X_t}{|\widetilde{\Delta X}_t^{(1)}|} \xrightarrow{\mathbb{P}} \infty, \text{ as } t \downarrow 0;$$

$$\sigma^2 = 0 \text{ and } \lim_{x \downarrow 0} \frac{A(x)}{x\bar{\Pi}(x)} = \infty;$$

$X \in \text{PRS}$  at 0;

$$\lim_{x \downarrow 0} \frac{A(x)}{\sqrt{U(x)\bar{\Pi}(x)}} = \infty;$$

$$\lim_{x \downarrow 0} \frac{x A(x)}{U(x)} = \infty.$$

## domain of attraction to normality

We say  $X \in D(N)$  at 0 if there are  $a_t \in \mathbb{R}$ ,  $b_t > 0$ , such that  $(X_t - a_t)/b_t \xrightarrow{D} N(0, 1)$  as  $t \downarrow 0$

if  $a_t$  may be taken as 0, we write  $X \in D_0(N)$

• (Doney and Maller (2002))  $X \in D$  iff  $\lim_{x \downarrow 0} \frac{U(x)}{x^2 \bar{\Pi}(x)} = \infty$ ;  
in fact,  $D(N) = D_0(N)$  (Maller & Mason (2010));

$$X \in D \quad \text{iff} \quad X \in D_0 \quad \lim_{x \downarrow 0} \frac{U(x)}{x|A(x)| + x^2 \bar{\Pi}(x)} = \infty$$

**corollary**[to theorem] if  $\bar{\Pi}^+(0+) = \infty$ , then the following are equivalent:

$$X \in D(N)$$

there is a nonstochastic function  $c_t > 0$  such that  $\frac{V_t}{c_t} \xrightarrow{P} 1$ , as  $t \downarrow 0$ ;

$$\frac{V_t}{\sup_{0 < s \leq t} |\Delta X_s|^2} \xrightarrow{P} \infty, \text{ as } t \downarrow 0.$$

# relative stability, attraction to normality & two-sided dominance

**theorem** if  $\bar{\Pi}(0+) = \infty$ , then the following are equivalent:

$$\frac{|X_t|}{|\widetilde{\Delta X}_t^{(1)}|} \xrightarrow{P} \infty, \text{ as } t \downarrow 0;$$

$$\lim_{x \downarrow 0} \frac{x|A(x)| + U(x)}{x^2 \bar{\Pi}(x)} = \infty;$$

$$\lim_{x \downarrow 0} \frac{U(x)}{x|A(x)| + x^2 \bar{\Pi}(x)} = +\infty, \quad \text{or} \quad \lim_{x \downarrow 0} \frac{|A(x)|}{x \bar{\Pi}(x)} = +\infty;$$

$$X \in D_0(N) \cup RS \text{ at } 0$$

## outlook

- talk was based on B.B., Y. Fan, Y. & R.A. Maller (2016). Distributional Representations and Dominance of a Lévy Process over its Maximal Jump Processes. *Bernoulli* **22**(4), 2325–2371.
- Yuguang Fan proved NASC convergence of trimmed Lévy processes in the domain of attraction of normal and stable to trimmed counterparts, including point process versions in her thesis and/or articles
  - Study in trimmed Lévy processes*. PhD thesis. ANU.
  - Convergence of trimmed Lévy processes to trimmed stable random variables at 0. To appear in *Stoch. Pro. & its Appl.*
  - Tightness and Convergence of Trimmed Lévy Processes to Normality at Small Times. To appear in *J of Theo Prob.*
- current work includes small-time behaviour of extremal processes (ongoing project with Ross Maller & Ana Feirerra) & extremal processes as infinite state space Markov processes (ongoing project with Ross Maller & Sid Resnick) indexed by its order

as usual: finis delectat

**Je vous remercie de votre attention!!**