

Remains to prove that

in case (1) $\lim_{t \rightarrow \infty} \tilde{z}_t = \infty$

Case (2) $\lim_{t \rightarrow +\infty} \tilde{z}_t = -\infty$.

Recall that from previous calculations, we

can represent
$$I_\infty = \inf \{ t > 0 : X_t^{(x)} = 0 \}$$

$$= r^\alpha \int_0^\infty e^{r \tilde{z}_s} ds.$$

$$\left[\tilde{z}_t = \log \left(X_t^{(x)} / r I_t \right) \right]$$

We need a Lemma

(Nothing in \tilde{z})

$$\int_0^\infty e^{r \tilde{z}_t} dt < \infty \text{ a.s.}$$

$$\iff \lim_{t \rightarrow \infty} \tilde{z}_t = -\infty$$

Note: \forall L.P. we have either

$$\lim_{t \rightarrow \infty} \tilde{z}_t = \infty \text{ (s)} \text{ or } \lim_{t \rightarrow \infty} \tilde{z}_t = -\infty \text{ (a.s.)}$$

$$\text{or } \limsup_{t \rightarrow \infty} \tilde{z}_t = -\liminf_{t \rightarrow \infty} \tilde{z}_t = \infty \text{ (s.)}$$

In fact: either $\lim_{t \rightarrow \infty} \frac{\tilde{z}_t}{t} = c_+ \in (0, \infty)$ a.s. $\Rightarrow \int_0^\infty e^{r \tilde{z}_t} dt = \infty$

or $\lim_{t \rightarrow \infty} \frac{\tilde{z}_t}{t} = c_- \in (-\infty, 0)$ a.s. $\Rightarrow \int_0^\infty e^{r \tilde{z}_t} dt < \infty$

or $\limsup_{t \rightarrow \infty} \tilde{z}_t = -\liminf_{t \rightarrow \infty} \tilde{z}_t = \infty$.

look at $T_1 = \inf \{ t > 0 : \tilde{z}_t > 2 \}$

$S_1 = \inf \{ t > T_1 : \tilde{z}_t < 1 \}$

$T_2 = \inf \{ t > S_1 : \tilde{z}_t > 2 \}$

$S_2 = \inf \{ t > T_2 : \tilde{z}_t < 1 \}$

\vdots

$$I_\infty := \int_0^\infty e^{r \tilde{z}_t} dt \geq e^{r \sum_{n=1}^\infty (S_n - T_n)}$$

\hookrightarrow stochastically dominating indep. copies of $T_1 = \inf \{ s > 0 : \tilde{z}_s < -1 \}$ under $\tilde{z}_0 = 0$.

i.e. $I_\infty \geq I_\infty' = e^{r \sum_{n \geq 1} T_1(n)} = \infty$ a.s.

show $T_1(n) \sim \text{i.i.d. } T_1$

Stable Process:

A stable process is a Lévy process which respects the scaling property

$$\left\{ c Y_{c^{-\alpha}t}^{(cx)} : t \geq 0 \right\} \stackrel{d}{=} \left\{ Y_t^{(cx)} : t \geq 0 \right\}$$

Note: A stable process is NOT a \mathbb{P}^{ss} M.p. because in general Y is \mathbb{R} -valued!

Note: necessarily $\alpha \in (0, 2]$ with $\alpha = 2$ being (scaled) BM.

First properties on Lévy processes:

A L.P. (1) $\xi_0 = 0$

(2) $0 < s < t < \infty$, $\xi_t - \xi_s \perp \left\{ \xi_u : u \leq s \right\}$

(3) $\xi_t - \xi_s \stackrel{d}{=} \xi_{t-s}$

(4) ξ has a.s. right- & left limits in its paths.

* unless Y is a stable subordinator

Character exponent:

$$\textcircled{*} \quad \sum_t = \sum_{t/n} + \underbrace{\left(\sum_{\frac{t}{n}} - \sum_{\frac{t}{n}} \right)}_{= \sum_{t/n}} + \underbrace{\left(\sum_{\frac{2t}{n}} - \sum_{\frac{t}{n}} \right)}_{= \sum_{t/n}} + \dots + \underbrace{\left(\sum_t - \sum_{\frac{t(n-1)}{n}} \right)}_{= \sum_{t/n}}$$

$n \in \mathbb{N}$. (distⁿ of \sum_t is infinitely divisible).

$$\Psi_t(\theta) := -\log \mathbb{E}[e^{i\theta \sum_t}]$$

$$\textcircled{*} \Rightarrow \begin{array}{l} \text{on the one hand} \\ \text{on the other hand} \end{array} \quad \begin{array}{l} \Psi_m(\theta) = m \Psi_1(\theta) \\ \Psi_{m/n}(\theta) = n \Psi_{1/n}(\theta) \end{array} \quad m, n \in \mathbb{N}$$

$$\Rightarrow \Psi_{\frac{m}{n}}(\theta) = \frac{m}{n} \Psi_1(\theta)$$

$$\forall t \in \mathbb{Q} \cap [0, \infty) \quad \Psi_t(\theta) = t \boxed{\Psi_1(\theta)} = \Psi(\theta)$$

$$\textcircled{**} \cdot \mathbb{E}[e^{i\theta \sum_t}] = e^{-t\Psi(\theta)} \quad \text{for } t \in \mathbb{Q} \cap [0, \infty)$$

use DCT + right-continuity to deduce ($t_n \in \mathbb{Q} : t_n \downarrow t \in [0, \infty)$)

$\Rightarrow \textcircled{**}$ holds $\forall t$.

$\Psi(\theta)$ is the char. exp. of $\sum_t \leftarrow$ inf-div. distⁿ.

all inf-div. distⁿ necessarily have char. exponents of the form (Lévy-Khintchine formula)

$$\Psi(\theta) = \underbrace{ia\theta + \frac{1}{2}\sigma^2\theta^2}_{\text{Gaussian part}} + \underbrace{\int_{\mathbb{R}} (1 - e^{i\theta x} + i\theta x \mathbb{1}_{(|x| < 1)}) \Pi(dx)}_{\text{Lévy measure}}$$

$a \in \mathbb{R}, \sigma^2 \geq 0$ Π is a measure on $\mathbb{R} \setminus \{0\}$

$$\text{s.t.} \quad \int_{\mathbb{R}} (1 \wedge x^2) \Pi(dx) < \infty.$$

In fact: Theorem Class of L.P.s corresponds to the class of inf-div. distⁿ through the L-K-formula.

Exercise: Consider a compound Poisson process with drift: λ arrival rate, iid jumps have distⁿ F and c is the drift.

$$\sum_t = ct + \sum_{i=1}^{N_t} J_i \quad : \quad J_i \stackrel{iid}{\sim} F$$

is a L.P.

$$\begin{aligned} \mathbb{E}[e^{i\theta \sum_t}] &= e^{i\theta ct} \mathbb{E}\left[e^{i\theta \sum_{k=1}^{N_t} J_k}\right] \\ &= e^{i\theta ct} \mathbb{E}\left[\mathbb{E}[e^{i\theta J_1}]^{N_t}\right] \\ &= e^{i\theta ct} \mathbb{E}\left[\left(\int_{\mathbb{R}} e^{i\theta x} F(dx)\right)^{N_t}\right] \end{aligned}$$

$$\mathbb{E}[s^{N_t}] = e^{\lambda t(s-1)}$$

$$= e^{i\theta ct} \exp\left(\lambda t \left(\int_{\mathbb{R}} (e^{i\theta x} - 1) F(dx)\right)\right)$$

$$= e^{-t\Psi(\theta)}$$

$$\Psi(\theta) = -ci\theta + \int_{\mathbb{R}} (1 - e^{i\theta x}) \lambda F(dx).$$

suppose $\int_{\mathbb{R}} |x| F(dx) < \infty$ we take $c = -\int_{\mathbb{R}} x \lambda F(dx)$

(Exercise: this makes \sum_t a m.g.)

$$\text{also } \Psi(\theta) = \int_{\mathbb{R}} (1 - e^{i\theta x} + i\theta x) \lambda F(dx)$$

$$\Psi(\theta) = i\alpha\theta + \frac{1}{2}\sigma^2\theta^2$$

$$+ \int_{\mathbb{R}} (1 - e^{i\theta x} + i\theta x \mathbb{1}_{(|x| < 1)}) \Pi(dx)$$

$$\begin{aligned} & \int_{|x| \geq 1} (1 - e^{i\theta x}) \frac{\Pi(dx)}{\Pi(|x| \geq 1)} \cdot \Pi(|x| \geq 1) \\ & + \sum_{n=1}^{\infty} \int_{|x| \in [2^{-n}, 2^{-(n-1)})} (1 - e^{i\theta x} + i\theta x) \frac{\Pi(dx)}{\Pi(|x| \in [2^{-n}, 2^{-(n-1)})}} \cdot \Pi(|x| \in [2^{-n}, 2^{-(n-1)}]) \end{aligned}$$

Note: $\int_{\mathbb{R}} (1 \wedge x^2) \Pi(dx) < \infty$

$$\Rightarrow \Pi(A) < \infty \quad \forall A \text{ s.t. } \{0\} \notin \bar{A}$$

Note: \mathbb{Z} and \mathbb{Z}^2 are two indep. L.P.
with exponents $\Psi^1(\theta)$ & $\Psi^2(\theta)$

$$\mathbb{E}[e^{i\theta(\mathbb{Z}_t^1 + \mathbb{Z}_t^2)}] = e^{-t(\Psi^1(\theta) + \Psi^2(\theta))}$$

and $\mathbb{Z}_t^1 + \mathbb{Z}_t^2$ is a L.P.

Can we add an infinite # L.P. together?

In the exponent

$$\lim_{k \rightarrow \infty} \sum_{n=1}^k \int_{|x| \in [2^{-n}, 2^{-(n-1)})} (1 - e^{i\theta x} + i\theta x) \Pi(dx)$$

corresponds to the "infinite" sum of centred CPPs
which are also mgf.

an the sum of processes converges thanks to L^2 -mgf
theory which will use $\int_{(-1,1)} x^2 \Pi(dx) < \infty$

How do stable processes fit into this picture?

Usual defⁿ: $\bar{v} = 0$ ($\alpha \in (0, 2)$)

$$\Pi(dx) = \frac{c_1}{|x|^{1+\alpha}} \mathbb{1}_{(x>0)} dx + \frac{c_2}{|x|^{1+\alpha}} \mathbb{1}_{(x<0)} dx$$

[A jump of size x arrives in $(t, t+dt)$ u.p. $\Pi(dx)dt + o(dt)$]

There exists an appropriate choice of $a \in \mathbb{R}$ s.t.

$$\Psi(\theta) = c |\theta|^\alpha \left(1 - \beta \tan(\pi\alpha/2) \operatorname{sgn}(\theta) \right)$$

$$c = -(c_1 + c_2) \cos(\pi\alpha/2) \Gamma(-\alpha)$$

$$\beta = \frac{c_1 - c_2}{c_1 + c_2}$$

Note:
$$\mathbb{E} \left[e^{i\theta(cY_{c^{-\alpha}t})} \right] = \mathbb{E} \left[e^{i(\theta c^{-\alpha}) Y_{c^{-\alpha}t}} \right]$$

$$= e^{-c^{-\alpha}t \Psi(c\theta)} = e^{-t \Psi(\theta)}$$

Henceforth: we will use a particular normalization of \mathbb{P} .

Theorem Because of scaling $\mathbb{P}(Y_t > 0) = p$
 where p is a constant in $[0, 1]$ and

$$p = \frac{1}{2} + \frac{1}{\pi\alpha} \tan^{-1}(\beta \tan(\pi\alpha/2))$$

By choosing $c_1 = \Gamma(1+\alpha) \sin(\pi\alpha p)$

$$c_2 = \Gamma(1+\alpha) \sin(\pi\alpha(1-p)) \quad \hat{p} = 1-p$$

then
$$\Psi(\theta) = |\theta|^\alpha \left(e^{\pi i \alpha (\frac{1}{2} - \hat{p})} \mathbb{1}_{(\theta > 0)} + e^{-\pi i \alpha (\frac{1}{2} - \hat{p})} \mathbb{1}_{(\theta < 0)} \right)$$

In fact the parameter range of (α, p) turns out to be a bit more restrictive than $\alpha \in (0, 2)$ and $p \in [0, 1]$

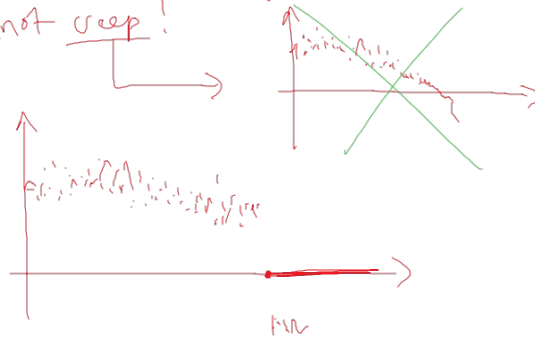
$$A = \left\{ \alpha \in (0, 1), p \in [0, 1] \right\} \cup \left\{ \alpha = 1, p = \frac{1}{2} \right\} \cup \left\{ \alpha \in (1, 2), p \in [1-\frac{1}{\alpha}, \frac{1}{\alpha}] \right\}$$

Casady Process

Two examples of pssMp using stable Lévy processes

(1) $X_t := Y_t \mathbb{1}_{(Y_t \geq 0)} = Y_t \mathbb{1}_{(t < \underline{T}_0^-)$

[Note: a stable process always jumps across the origin
i.e. cannot creep!]



(2) $A_t = \int_0^t \mathbb{1}_{(Y_s > 0)} ds$

$\gamma(t) = A_t^{-1} \dots \int_0^{\gamma(t)} \mathbb{1}_{(Y_s > 0)} ds = t$

$\check{X}_t = Y_{\gamma(t)}$

$T_0 = \inf \{ t > 0 : \check{X}_t = 0 \}$

$X_t = \check{X}_t \mathbb{1}_{(t < T_0)}$

