

# Self-regulating processes

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## Motivation

1. Modelling phenomena with varying local regularity is important in various applications: finance, geophysics, internet traffic modelling, ...
2. Often, local regularity varies as a function not of time or space but as a function of the value of the process itself.
3. When local regularity is measured in terms of Hölder exponents, this gives rise to *self-regulating processes*.
4. However, when dealing with discontinuous processes, it is necessary to account for an intensity of jumps that would depend on the value of the process. We refer to those processes as *self-regulating processes*.

## Motivation

More precisely, we construct jump processes of a self-regulating nature, that is variants on  $\alpha$ -stable processes where the stability index  $\alpha$  around time  $t$  depends on the value of the process at time  $t$ .

The construction utilizes the Poisson sum representation of  $\alpha$ -stable processes as a sum over a point set in the plane.

## Symmetric $\alpha$ -stable Lévy motion

$\{L_\alpha(t), t \geq 0\}$  ( $0 < \alpha \leq 2$ ), is the stochastic process with stationary independent increments such that  $L_\alpha(0) = 0$  almost surely, and  $L_\alpha(t) - L_\alpha(s)$  has the distribution of  $S_\alpha((t-s)^{1/\alpha}, 0, 0)$ , where  $S_\alpha(c, \beta, \mu)$  denotes a stable random variable with stability-index  $\alpha$ , with scale parameter  $c$ , skewness parameter  $\beta$ , and shift  $\mu$ .

$L_\alpha$  admits a version with càdlàg sample paths. It may be represented as:

$$L_\alpha(t) = C_\alpha \sum_{(X,Y) \in \Pi} 1_{(0,t]}(X) Y^{\langle -1/\alpha \rangle} \quad (2.1)$$

where  $C_\alpha$  is a normalising constant,  $\Pi$  is a Poisson point process on  $\mathbb{R}^+ \times \mathbb{R}$  with plane Lebesgue measure  $\mathcal{L}^2$  as mean measure, and

$$r^{\langle s \rangle} = \text{sign}(r)|r|^s, \quad r \in \mathbb{R}, s \in \mathbb{R}.$$

We are interested in processes where the intensity of jumps  $\alpha$  varies. The simplest version of such a process is the *multistable symmetric Lévy motion*  $\{M_\alpha(t), t \geq 0\}$  with following representation:

$$M_\alpha(t) = C_\alpha(t) \sum_{(X,Y) \in \Pi} 1_{(0,t]}(\mathbf{X}) Y^{\langle -1/\alpha(t) \rangle} \quad (2.2)$$

where  $\alpha$  is a suitable function from  $\mathbb{R}^+$  to  $(0, 2)$ .

$\{M_\alpha(t), t \geq 0\}$  is *localizable* in the following sense: for each  $t > 0$  and  $u \in \mathbb{R}$ ,

$$\frac{M_\alpha(t + ru) - M_\alpha(t)}{r^{1/\alpha(t)}} \xrightarrow{\text{dist}} L_{\alpha(t)}(u)$$

as  $r \searrow 0$ , where convergence is in distribution with respect to the Skorohod metric.

Thus  $M$  'looks like'  $\alpha(t)$ -stable process near  $t$ .

With multistable motion, the local stability parameter depends on the time  $t$ .

Our aim is to construct a process  $Z$  where the local stability parameter at time  $t$  depends instead on the value of the process at time  $t$ : for suitable  $\alpha : \mathbb{R} \rightarrow (0, 2)$ ,  $Z(t)$  would be localizable in the sense that for all  $t \in [t_0, t_1)$  and  $u > 0$

$$\frac{Z(t + ru) - Z(t)}{r^{1/\alpha(Z(t))}} \Big|_{\mathcal{F}_t} \xrightarrow{\text{dist}} L_{\alpha(Z(t))}^0(u)$$

as  $r \searrow 0$ , where  $\mathcal{F}_t$  indicates conditioning on the process up to time  $t$ .



We first deal with the case where the local jump intensity function ranges in  $(0, 1)$  and start by a deterministic construction.

Given a countable discrete point set  $\Pi$  in the plane we define real valued functions  $f$  on an interval  $[t_0, t_1)$  such that  $f(t)$  'jumps' when  $t = x$  for each  $(x, y) \in \Pi$ , the magnitude of the jump depending both on  $y$  and on the value of  $\lim_{t \nearrow x} f(t)$ .

For  $t_0 < t_1$  let  $D[t_0, t_1)$  denote the càdlàg functions on  $[t_0, t_1)$ . The space  $D[t_0, t_1)$  is complete under the supremum norm  $\|\cdot\|_\infty$ .

Fix  $0 < a < b < 1$ . Let  $\alpha : \mathbb{R} \rightarrow [a, b]$  be continuously differentiable with bounded derivative and let  $\Pi \subset (t_0, t_1) \times \mathbb{R}$  be a set of points such that

$$\sum_{(x,y) \in \Pi} |y|^{-1/b'} < \infty \quad (3.3)$$

for some  $b < b' < 1$ .

Let  $a_0 \in \mathbb{R}$ . Define  $K$  on  $D[t_0, t_1)$  by

$$K(f)(t) = a_0 + \sum_{(x,y) \in \Pi} 1_{(t_0, t]}(x) y^{\langle -1/\alpha(f(x_-)) \rangle} \quad (t_0 \leq t < t_1), \quad (3.4)$$

where the sum is absolutely convergent by (3.3).

### Lemma

The operator  $K$  maps  $D[t_0, t_1)$  into itself.

## Theorem

With  $a_0, \alpha$  and  $\Pi$  as above, there exists a unique  $f \in D[t_0, t_1)$  such that

$$f(t) = a_0 + \sum_{(x,y) \in \Pi} 1_{(t_0, t]}(x) y^{\langle -1/\alpha(f(x_-)) \rangle} \quad (t_0 \leq t < t_1). \quad (3.5)$$

In particular  $f(t_0) = a_0$ . Moreover, for each  $s$  and  $t$  with  $t_0 \leq s < t < t_1$ ,  $f(t)$  is completely determined given  $f(s)$  and the points of the set  $\Pi \cap ((s, t] \times \mathbb{R})$ .

## Idea of proof:

We would like to use that  $K$  is a contracting operator on  $D[t_0, t_1)$  and apply Banach's contraction theorem. However,  $K$  is contracting only if the value of  $|y|$  is not too small at points  $(x, y) \in \Pi$ .

We use that  $K$  is contracting in intervals where  $|y|$  is not too small and incorporate the jumps at these points 'by hand'.

The function  $f$  may be approximated as follows: define a sequence of functions  $f_n$  ( $n \in \mathbb{N}$ ) by restricting the sums to points with  $|y| \leq n$ , that is :

$$f_n(t) = a_0 + \sum_{(x,y) \in \Pi: |y| \leq n} 1_{(t_0, t]}(x) y^{\langle -1/\alpha(f_n(x_-)) \rangle} \quad (3.6)$$

for  $t_0 \leq t < t_1$ . Then  $f_n \in D[t_0, t_1)$  is uniquely defined as a sum over a finite set of points and is piecewise constant, so it may be evaluated using a finite number of inductive steps.

## Theorem

$\{f_n\}$  is a Cauchy sequence in  $(D[t_0, t_1], \|\cdot\|_\infty)$ . Moreover,  $f_n \rightarrow f$  in  $\|\cdot\|_\infty$  and we have:

$$\|f_n - f\|_\infty \leq \exp\left(M \sum_{(x,y) \in \Pi, |y| \leq n} |y|^{-1/(a,b)}\right) \sum_{(x,y) \in \Pi: |y| > n} |y|^{-1/b},$$

where

$$M = \sup_{\xi \in \mathbb{R}} \frac{|\alpha'(\xi)|}{\alpha(\xi)^2}.$$

## Idea of proof:

The main difficulty with the sequence of functions  $f_n(t)$  is that when, as  $n$  increases, a new point  $(x, y)$  enters the sum then, for all existing  $(x', y')$  with  $x' > x$  and smaller  $|y'|$ , the summands  $y'^{\langle -1/\alpha(f(x'-)) \rangle}$  will change, leading to a change in  $f_n(t)$  for  $t > x$  that is amplified as  $t$  increases past larger  $x$  with  $(x, y) \in \Pi$ .

We deal with this difficulty by carefully controlling these changes.



Going back to the random case, we have the following result:

### Theorem

Let  $\Pi \subset (t_0, t_1) \times \mathbb{R}$  be a Poisson point process with  $\mathcal{L}^2$  as mean measure. Then there exists a Markov process  $Z$  on  $[t_0, t_1)$  such that, almost surely, the sample paths are in  $D[t_0, t_1)$  with  $Z(t_0) = a_0$  and

$$Z(t) = a_0 + \sum_{(X,Y) \in \Pi} 1_{(t_0, t]}(\mathbf{X}) Y^{\langle -1/\alpha(Z(\mathbf{X}_-)) \rangle} \quad (t_0 \leq t < t_1). \quad (3.7)$$

Writing

$$Z_n(t) = a_0 + \sum_{(X,Y) \in \Pi: |Y| \leq n} 1_{(t_0, t]}(\mathbf{X}) Y^{\langle -1/\alpha(Z_n(\mathbf{X}_-)) \rangle} \quad (t_0 \leq t < t_1) \quad (3.8)$$

then almost surely,  $\|Z_n - Z\|_\infty \rightarrow 0$  as  $n \rightarrow \infty$ .

We investigate the local properties of  $Z$ .

### Proposition

Given  $0 < \epsilon < 1/b$ , for each  $t \in [t_0, t_1)$  there exists almost surely a random  $C > 0$  such that for all  $0 < h < t_1 - t$ ,

$$|Z(t+h) - Z(t)| \leq Ch^{1/\alpha(Z(t)) - \epsilon}. \quad (3.9)$$

## Idea of proof:

The main ingredient is a comparison with the  $\alpha$ -stable subordinator  $S_\alpha$ , for constant  $0 < \alpha < 1$ :

$$S_\alpha(t) := \sum_{(X,Y) \in \Pi} 1_{(0,t]}(X) |Y|^{-1/\alpha}.$$

$S_\alpha$  is a self-similar process with stationary increments such that for all  $0 < \epsilon < 1/\alpha$  there is almost surely a random constant  $C < \infty$  such that

$$S_\alpha(t) \leq Ct^{(1/\alpha) - \epsilon} \quad (t \geq 0). \quad (3.10)$$

## Theorem

$Z$  is right-localizable at each  $t \in [t_0, t_1)$ , in the sense that

$$\frac{Z(t + ru) - Z(t)}{r^{1/\alpha(Z(t))}} \Big|_{\mathcal{F}_t} \xrightarrow{\text{dist}} L_{\alpha(Z(t))}^0(u) \quad (3.11)$$

as  $r \searrow 0$ , where convergence is in distribution with respect to  $(D[0, t_1), \rho_S)$ , with  $\rho_S$  the Skorohod metric.

Idea of proof:

We compare  $Z(t + ru) - Z(t)$  and  $L_{\alpha(Z(t))}^0(t + ru) - L_{\alpha(Z(t))}^0(t)$ , where

$$L_{\alpha}^0(t) = \sum_{(X,Y) \in \Pi} 1_{(0,t]}(\mathbf{X}) Y^{\langle -1/\alpha \rangle} \quad (t \geq 0). \quad (3.12)$$

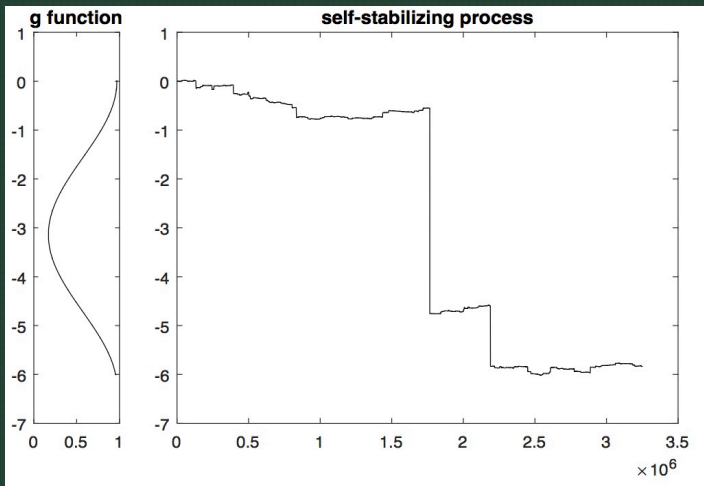


Figure: left : self-stabilizing function  $\alpha(z) = 0.57 + 0.4 \cos(z)$ . Right: corresponding realization of a self-stabilizing process.

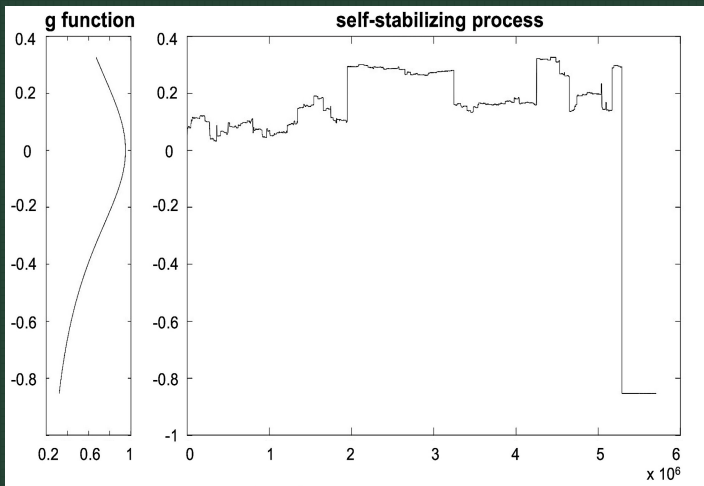


Figure: left : self-stabilizing function  $\alpha(z) = 0.15 + \frac{0.8}{1+5z^2}$ . Right: corresponding realization of a self-stabilizing process.

When  $\alpha : \mathbb{R}^+ \rightarrow (0, 1)$ , we started by showing that there exists a deterministic function  $f \in D[t_0, t_1)$  satisfying the relation

$$f(t) = a_0 + \sum_{(x,y) \in \Pi} 1_{(t_0, t]}(x) y^{\langle -1/\alpha(f(x_-)) \rangle}$$

for a fixed point set  $\Pi$ , and then randomising to get a random function  $Z$  such that

$$Z(t) = a_0 + \sum_{(X,Y) \in \Pi} 1_{(t_0, t]}(X) Y^{\langle -1/\alpha(Z(X_-)) \rangle} \quad (t_0 \leq t < t_1).$$

However, this approach depends on the infinite sums being absolutely convergent, which need not be the case if  $\alpha(t) \geq 1$  for some  $t$ .

The general case is more involved.



We show first that for a *fixed* point set  $\Pi^+ \subset (t_0, t_1) \times \mathbb{R}^+$  and independent random ‘signs’  $S(x, y) = \pm 1$  there exists almost surely a random function  $Z^p \in D[t_0, t_1)$  satisfying

$$Z^p(t) = a_0 + \sum_{(x,y) \in \Pi^+} 1_{(t_0, t]}(x) S(x, y) y^{-1/\alpha(Z^p(x-))} \quad (t_0 \leq t < t_1).$$

To achieve this we work with partial sums

$$Z_n^p(t) = a_0 + \sum_{(x,y) \in \Pi^+ : |y| \leq n} 1_{(t_0, t]}(x) S(x, y) y^{-1/\alpha(Z_n^p(x-))} \quad (t_0 \leq t < t_1)$$

and show that the limit as  $n \rightarrow \infty$  exists in a norm given by  $\mathbb{E}(\|\cdot\|_\infty^2)^{1/2}$ , where  $\mathbb{E}$  denotes expectation.

## Theorem

$$\mathbb{E}(\|Z_n^p - Z^p\|_\infty^2) \leq 4 \prod_{(x,y) \in \Pi^+} (1 + M^2 y^{-2/(a,b)}) \sum_{(x,y) \in \Pi^+ : y > n} y^{-2/b} \rightarrow 0. \quad (4.13)$$

## Proposition

Let  $Z \in \mathcal{D}$  be the random function given by Theorem 4.1 and let  $t \in [t_0, t_1)$ . Suppose that for some  $\beta > 0$ ,

$$\sum_{(x,y) \in \Pi^+ : t < x \leq t+h} y^{-2/\alpha(Z(t))} = O(h^\beta) \quad (0 < h < t_1 - t). \quad (4.14)$$

Then, conditional on  $\mathcal{F}_t$ , given  $0 < \epsilon < \beta$  there exist almost surely a random number  $C_1 < \infty$  such that for all  $0 \leq h < t_1 - t$ ,

$$|Z(t+h) - Z(t)| \leq C_1 h^{(\beta-\epsilon)/2}. \quad (4.15)$$

We now randomise the construction further by taking  $\Pi^+$  to be a Poisson point process on  $(t_0, t_1) \times \mathbb{R}^+$  with mean measure  $2\mathcal{L}^2$ .

The key idea is that the distribution of the point sets  $\{(X, Y) \in \Pi\}$  where  $\Pi \subset (t_0, t_1) \times \mathbb{R}$  is a Poisson point process with plane Lebesgue measure  $\mathcal{L}^2$  as mean measure, is identical to that of  $\{(X, S(X, Y)Y) : (X, Y) \in \Pi_2^+, S(X, Y) = \pm 1\}$ , where  $\Pi_2^+$  is a Poisson point process on  $(t_0, t_1) \times \mathbb{R}^+$  with double Lebesgue measure  $2\mathcal{L}^2$  as mean measure and with the  $S(X, Y)$  independently taking the values  $\pm 1$  with equal probability  $\frac{1}{2}$  for each  $(X, Y) \in \Pi_2^+$  (this follows from the superposition property of Poisson processes).

Hence  $\Pi$  can be realised by first sampling  $(X, Y)$  from  $\Pi_2^+$  and then assigning random signs to the  $Y$  coordinates.

## Theorem

Let  $\Pi \subset (t_0, t_1) \times \mathbb{R}$  be a Poisson point process with mean measure  $\mathcal{L}^2$ , let  $\alpha : \mathbb{R} \rightarrow [a, b]$  where  $0 < a < b < 2$  and let  $a_0 \in \mathbb{R}$ . Then there exists  $Z \in \mathcal{D}$  such that  $\lim_{n \rightarrow \infty} \mathbb{E}(\|Z_n - Z\|_\infty^2) = 0$  where  $Z_n$  is defined as

$$Z_n(t) = a_0 + \sum_{(\mathbf{X}, \mathbf{Y}) \in \Pi: |\mathbf{Y}| \leq n} 1_{(t_0, t]}(\mathbf{X}) \mathbf{Y}^{(-1/\alpha(Z_n(\mathbf{X}_-)))} \quad (t_0 \leq t < t_1).$$

## Theorem

Let  $y_0 > 0$  and let  $\Pi$  be a Poisson point process on  $(t_0, t_1) \times (-\infty, -y_0] \cup [y_0, \infty)$  with mean measure  $\mathcal{L}^2$  restricted to this domain. Then,

$$\mathbb{E}(\|Z_n - Z\|_\infty^2) \leq \frac{8b(t_1 - t_0)}{2 - b} \exp\left(2M^2(t_1 - t_0) \int_{y_0}^{\infty} y^{-2/(a,b)} dy\right) n^{-(2-b)/b}$$

## Proposition

Conditional on  $\mathcal{F}_t$ , given  $0 < \epsilon < 1/b$ , there exist almost surely random numbers  $C_1, C_2 < \infty$  such that for all  $0 \leq h < t_1 - t$ ,

$$|Z(t+h) - Z(t)| \leq C_1 h^{1/\alpha(Z(t)) - \epsilon}$$

and

$$\left| (Z(t+h) - Z(t)) - (L_{\alpha(t)}^0(t+h) - L_{\alpha(t)}^0(t)) \right| \leq C_2 h^{1/\alpha(Z(t)) + 1/b - \epsilon},$$

where  $L_{\alpha(t)}^0$  is the  $\alpha(t)$ -stable process defined using the same realisations of  $\Pi$  as  $Z$ .

## Theorem

Let  $t \in [t_0, t_1)$ . Then, conditional on  $\mathcal{F}_t$ , almost surely  $Z$  is right-localisable at  $t$ , in the sense that

$$\frac{Z(t+ru) - Z(t)}{r^{1/\alpha(Z(t))}} \Big| \mathcal{F}_t \xrightarrow{\text{dist}} L_{\alpha(Z(t))}^0(u) \quad (0 \leq u \leq 1)$$

as  $r \searrow 0$ , where convergence is in distribution with respect to  $(D[0, 1], \rho_S)$ , with  $\rho_S$  the Skorohod metric.