

# Looking for martingales associated to a self-decomposable law

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**Abstract** We construct martingales whose 1-dimensional marginals are those of a centered self-decomposable variable multiplied by some power of time  $t$ . Many examples involving quadratic functionals of Bessel processes are discussed.

**Key words** Convex order, Self-decomposable law, Sato process, Karhunen-Loeve representation, Perturbed Bessel process, Ray-Knight theorem.

## 1 Introduction, Motivation

### 1.1

We first introduce some notation which will be used throughout our paper.

If  $A$  and  $B$  are two random variables,  $A \stackrel{d}{=} B$  means that these variables have the same law.

If  $(X_t, t \geq 0)$  and  $(Y_t, t \geq 0)$  are two processes,  $(X_t) \stackrel{(1,d)}{=} (Y_t)$  means that the processes  $(X_t, t \geq 0)$  and  $(Y_t, t \geq 0)$  have the same

one-dimensional marginals, that is, for any fixed  $t$ ,  $X_t \stackrel{d}{=} Y_t$ .

If  $(X_t, t \geq 0)$  and  $(Y_t, t \geq 0)$  are two processes,  $(X_t) \stackrel{(d)}{=} (Y_t)$  means that the two processes are identical in law.

All random variables and processes which will be considered are assumed to be real valued.

## 1.2

In a number of applied situations involving randomness, it is a quite difficult problem to single out a certain stochastic process  $(Y_t, t \geq 0)$ , or rather its law, which is coherent with the real-world data.

In some cases, it is already nice to be able to consider that the one-dimensional marginals of  $(Y_t)$  are accessible. The random situation being studied may suggest, for instance, that:

- (i) there exists a martingale  $(M_t)$  such that

$$(Y_t) \stackrel{(1,d)}{=} (M_t)$$

(this hypothesis may indicate some kind of “equilibrium” with respect to time),

- (ii) there exists  $H > 0$  such that

$$(Y_t) \stackrel{(1,d)}{=} (t^H Y_1)$$

(there is a “scaling” property involved in the randomness).

It is a result due to Kellerer [12] that (i) is satisfied for a given process  $(Y_t)$  if and only if this process is increasing in the convex order, that is: it is integrable ( $\forall t \geq 0, \mathbb{E}[|Y_t|] < \infty$ ), and for every convex function  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$t \geq 0 \longrightarrow \mathbb{E}[\varphi(Y_t)] \in (-\infty, +\infty]$$

is increasing.

In the sequel, we shall use the acronym PCOC for such processes, since, in French, the name of such processes becomes: Processus Croissant pour l'Ordre Convexe.

A martingale  $(M_t)$  which has the same one-dimensional marginals as a PCOC is said to be *associated* to this PCOC. Note that several different martingales may be associated to a given PCOC. We shall see several striking occurrences of this in our examples.

Roynette [19] has exhibited two large families,  $(F1)$  and  $(F2)$ , of PCOC's: If  $(N_s)$  is a martingale satisfying some integrability condition, then

$$\left( \frac{1}{t} \int_0^t N_s \, ds, t \geq 0 \right) \text{ is a PCOC in } (F1);$$

$$\left( \int_0^t (N_s - N_0) \, ds, t \geq 0 \right) \text{ is a PCOC in } (F2).$$

### 1.3

It is a non-trivial problem to exhibit, for either of these PCOC's, an associated martingale. We have been able to do so concerning some examples in  $(F1)$ , in the Brownian context, with the help of the Brownian sheet, in our paper [8]. Concerning the class  $(F2)$ , note that, considering a trivial filtration, it follows that  $(tX)$ , where  $X$  is a centered random variable, is a PCOC. Even with this reduction, it is not obvious to find a martingale which is associated to  $(tX)$ . In order to exhibit examples, we were led to introduce the class  $(S)$  of processes  $(Y_t)$  satisfying the above condition (ii) and such that  $Y_1$  is a self-decomposable integrable random variable. It is a result due to Sato (see Sato [20, Chapter 3, Sections 15-17]) that, if  $(Y_t) \in (S)$ , then there exists a process  $(U_t)$  which has independent increments, is  $H$ -self-similar ( $\forall c > 0, (U_{ct}) \stackrel{(d)}{=} (c^H U_t)$ ) and satisfies  $Y_1 \stackrel{d}{=} U_1$ . This process  $(U_t)$ , which is unique in law, will be called the  $H$ -Sato process associated to  $Y_1$ . Clearly, then  $(U_t - \mathbb{E}[U_t])$  is a  $H$ -self-similar martingale which is associated to the PCOC  $(V_t)$  defined by:  $V_t = Y_t - \mathbb{E}[Y_t]$ . Moreover,  $(V_t) \stackrel{(1,d)}{=} (t^H (Y_1 - \mathbb{E}[Y_1]))$ .

We note that the self-decomposability property has also been used in Madan-Yor [15, Theorem 4, Theorem 5] in a very different manner than in this paper, to construct martingales with one-dimensional marginals those of  $(tX)$ .

### 1.4

We look for some interesting processes in the class  $(S)$ , in a Brownian framework.

**Example 1** A most simple example is the process:

$$Y_t := \int_0^t B_s \, ds, \quad t \geq 0$$

Then,

$$\left( \int_0^t B_s \, ds \right) \stackrel{(1.d)}{=} \left( \int_0^t s \, dB_s \right)$$

and the RHS is a centered  $(3/2)$ -Sato process. Moreover the process  $(Y_t)$  obviously belongs to the class  $(F2)$ .

**Example 2** The process

$$V_1(t) := \int_0^t (B_s^2 - s) \, ds \quad , \quad t \geq 0$$

and more generally the process

$$V_N(t) := \int_0^t (R_N^2(s) - Ns) \, ds \quad , \quad t \geq 0$$

where  $(R_N(s))$  is a Bessel process of dimension  $N > 0$  starting from 0, belongs to the family  $(F2)$  and is 2-self-similar. We show in Section 4 that the centered 2-Sato process:

$$\frac{N^2}{4} \int_0^{\tau_t} 1_{(|B_s| \leq \frac{2}{N} \ell_s)} \, ds - \frac{Nt^2}{2} \quad , \quad t \geq 0$$

where  $(\ell_s)$  is the local time in 0 of the Brownian motion  $B$ , and

$$\tau_t = \inf\{s ; \ell_s > t\}$$

is a martingale associated to the PCOC  $V_N$ .

**Example 3** We extend our discussion of Example 2 by considering, for  $N > 0$  and  $K > 0$ , the process:

$$V_{N,K}(t) := \frac{1}{K^2} \int_0^t s^{2(\frac{1}{K}-1)} (R_N^2(s) - Ns) \, ds \quad , \quad t \geq 0$$

Then, in Section 5, a centered  $(2/K)$ -Sato process (and hence a martingale) associated to the PCOC  $V_{N,K}$  may be constructed from the process of first hitting times of a perturbed Bessel process  $R_{K,1-\frac{N}{2}}$  as defined and studied first in Le Gall-Yor [13, 14] and then in Doney-Warren-Yor [5]. We remark that, if  $0 < K < 2$ , then the process

$$V_{N,K}(t^{\frac{K}{2-K}}) \quad , \quad t \geq 0$$

belongs to  $(F2)$ .

**Example 4** In Section 6, we generalize again our discussion by considering the process

$$V_N^{(\mu)}(t) := \int_{(0,\infty)} (R_N^2(ts) - Nts) d\mu(s) \quad , \quad t \geq 0$$

for  $\mu$  a nonnegative measure on  $(0, \infty)$  such that  $\int_{(0,\infty)} s d\mu(s) < \infty$ . We show that  $V_N^{(\mu)}$  is a PCOC to which we are able to associate two very different martingales. The first one is purely discontinuous and is a centered 1-Sato process, the second one is continuous. The method of proof is based on a Karhunen-Loeve type decomposition (see, for instance, [4] and the references therein, notably Kac-Siebert [11]). For this, we need to develop a precise spectral study of the operator  $K^{(\mu)}$  defined on  $L^2(\mu)$  by :

$$K^{(\mu)}f(t) = \int_{(0,\infty)} f(s) t \wedge s d\mu(s)$$

This spectral study is certainly classical, but we present it for the convenience of the reader.

## 1.5

It is still an open problem (for the authors) whether the PCOC's (in  $(F2)$ ):

$$\int_0^t H_n(B_s, s) ds \quad , \quad t \geq 0$$

defined from the 2 variables Hermite polynomials

$$H_n(x, s) = s^{n/2} h_n(x/\sqrt{s})$$

lend themselves to our method, namely: is it true that the random variable

$$\int_0^1 H_n(B_s, s) ds$$

is self decomposable for  $n = 3, 4, \dots$ .

## 1.6

We now present more precisely the organisation of our paper:

- in Section 2, we recall some basic results about various representations of self-decomposable variables, and we complete the discussion of Subsection 1.3 above;

- in Section 3, we consider the simple situation, as in Subsection 1.3, where  $Y_t = R_N^2(t)$ , for  $R_N$  a Bessel process of dimension  $N$  starting from 0;
- the contents of Sections 4, 5, 6 have already been discussed in the above Subsection 1.4;
- in a short final Section 7, we prove some negative results concerning further self-decomposability properties for squared Bessel processes: indeed, it is well-known, and goes back to Shiga-Watanabe [21], that  $R_N^2(\bullet)$ , considered as a random variable taking values in  $C(\mathbb{R}_+, \mathbb{R}_+)$  is infinitely divisible. Furthermore, in the present paper, we exploit the self-decomposability of  $\int_{(0,\infty)} R_N^2(s) d\mu(s)$  for any positive measure  $\mu$ . It then seemed natural to wonder about the self-decomposability of  $R_N^2(\bullet)$ , but this property is ruled out, as the 2-dimensional vectors:  $(R_N^2(t_1), R_N^2(t_1 + t_2))$  are not self-decomposable.

## 2 Sato processes and PCOC's

### 2.1 Self-decomposability and Sato processes

We recall, in this subsection, some general facts concerning the notion of self-decomposability. We refer the reader, for background, complements and references, to Sato [20, Chapter 3].

A random variable  $X$  is said to be *self-decomposable* if, for each  $u$  with  $0 < u < 1$ , there is the equality in law:

$$X \stackrel{d}{=} uX + \widehat{X}_u$$

for some variable  $\widehat{X}_u$  independent of  $X$ .

On the other hand, an *additive process*  $(U_t, t \geq 0)$  is a stochastically continuous process with càdlàg paths, independent increments, and satisfying  $U_0 = 0$ .

An additive process  $(U_t)$  which is *H-self-similar* for some  $H > 0$ , meaning that, for each  $c > 0$ ,  $(U_{ct}) \stackrel{(d)}{=} (c^H U_t)$ , will be called a *Sato process* or, more precisely, a *H-Sato process*.

The following theorem, for which we refer to Sato's book [20, Chapter 3, Sections 16-17], gives characterizations of the self-decomposability property that we state in the following theorem:

**Theorem 2.1** *Let  $X$  be a real valued random variable. Then,  $X$  is self-decomposable if and only if one of the following equivalent properties is satisfied:*

- 1)  $X$  is infinitely divisible and its Lévy measure is  $\frac{h(x)}{|x|} dx$  with  $h$  increasing on  $(-\infty, 0)$  and decreasing on  $(0, +\infty)$ .
- 2) There exists a Lévy process  $(C_s, s \geq 0)$  such that

$$X \stackrel{d}{=} \int_0^\infty e^{-s} dC_s .$$

- 3) For any (or some)  $H > 0$ , there exists a  $H$ -Sato process  $(U_t, t \geq 0)$  such that  $X \stackrel{d}{=} U_1$ .

In 2) (resp. 3)) the Lévy process  $(C_s)$  (resp. the  $H$ -Sato process  $(U_t)$ ) is uniquely determined in law by  $X$ , and will be said to be *associated* with  $X$ . We note that, if  $X \geq 0$ , then the function  $h$  vanishes on  $(-\infty, 0)$ ,  $(C_s)$  is a subordinator and  $(U_t)$  is an increasing process.

The relation between  $(C_s)$  and  $(U_t)$  was precised by Jeanblanc-Pitman-Yor [9, Theorem 1]:

**Theorem 2.2** *If  $(U_t)$  is a  $H$ -Sato process, then the formulae:*

$$C_s^{(-)} = \int_{e^{-s}}^1 r^{-H} dU_r \quad \text{and} \quad C_s^{(+)} = \int_1^{e^s} r^{-H} dU_r \quad , \quad s \geq 0$$

*define two independent and identically distributed Lévy processes from which  $(U_t, t \geq 0)$  can be recovered by:*

$$U_t = \int_{-\log t}^\infty e^{-sH} dC_s^{(-)} \quad \text{if} \quad 0 \leq t \leq 1$$

and

$$U_t = U_1 + \int_0^{\log t} e^{sH} dC_s^{(+)} \quad \text{if} \quad t \geq 1 .$$

*In particular, the Lévy process associated with the self-decomposable random variable  $U_1$  is*

$$C_s = C_{s/H}^{(-)} \quad , \quad s \geq 0 .$$

## 2.2 Sato processes and PCOC's

We recall (see Subsection 1.2) that a PCOC is an integrable process which is increasing in the convex order. On the other hand, a process  $(V_t, t \geq 0)$  is said to be a *1-martingale* if there exists, on some filtered probability space, a martingale  $(M_t, t \geq 0)$  such that  $(V_t) \stackrel{(1,d)}{=} (M_t)$ . Such a martingale  $M$  is said to be *associated* with  $V$ . It is a direct consequence of Jensen's inequality that, if  $V$  is a 1-martingale, then  $V$  is a PCOC. As indicated in Subsection 1.2, the converse holds true (Kellerer [12]).

The following proposition, which is central in the following, summarizes the method sketched in Subsection 1.3.

**Proposition 2.3** *Let  $H > 0$ . Suppose that  $Y = (Y_t, t \geq 0)$  satisfies:*

- (a)  $Y_1$  is an integrable self-decomposable random variable;
- (b)  $(Y_t) \stackrel{(1,d)}{=} (t^H Y_1)$ .

Then the process

$$V_t := Y_t - t^H \mathbb{E}[Y_1] \quad , \quad t \geq 0$$

is a PCOC, and an associated martingale is

$$M_t := U_t - t^H \mathbb{E}[Y_1] \quad , \quad t \geq 0$$

where  $(U_t)$  denotes the  $H$ -Sato process associated with  $Y_1$  according to Theorem 2.1.

## 3 About the process $(R_N^2(t), t \geq 0)$

In the sequel, we denote by  $(R_N(t), t \geq 0)$  the Bessel process of dimension  $N > 0$ , starting from 0.

### 3.1 Self-decomposability of $R_N^2(1)$

As is well-known (see, for instance, Revuz-Yor [18, Chapter XI]) one has

$$\mathbb{E}[\exp(-\lambda R_N^2(1))] = (1 + 2\lambda)^{-N/2} .$$

In other words,

$$R_N^2(1) \stackrel{d}{=} 2 \gamma_{N/2}$$



where, for  $a > 0$ ,  $\gamma_a$  denotes a gamma random variable of index  $a$ . Now,

$$\frac{N}{2} \log(1 + 2\lambda) = \frac{N}{2} \int_0^\infty (1 - e^{-\lambda t}) \frac{e^{-t/2}}{t} dt.$$

Then,  $R_N^2(1)$  satisfies the property 1) in Theorem 2.1 with

$$h(x) = \frac{N}{2} 1_{(0,\infty)}(x) e^{-x/2}$$

and it is therefore self-decomposable.

The process  $R_N^2$  is 1-self-similar and  $\mathbb{E}[R_N^2(1)] = N$ . By Proposition 2.3, the process

$$V_t^N := R_N^2(t) - tN \quad , \quad t \geq 0$$

is a PCOC, and an associated martingale is

$$M_t^N := U_t^N - tN \quad , \quad t \geq 0$$

where  $(U_t^N)$  denotes the 1-Sato process associated with  $R_N^2(1)$  by Theorem 2.1.

We remark that, in this case, the process  $(V_t^N)$  itself is a *continuous martingale* and therefore obviously a PCOC. In the following subsections, we give two expressions for the process  $(U_t^N)$ . As we will see, this process is purely discontinuous with finite variation; consequently, the martingales  $(V_t^N)$  and  $(M_t^N)$ , which have the same one-dimensional marginals, do not have the same law.

### 3.2 Expression of $(U_t^N)$ from a compound Poisson process

We denote by  $(\Pi_s, s \geq 0)$  the compound Poisson process with Lévy measure:

$$1_{(0,\infty)}(t) e^{-t} dt.$$

This process allows to compute the distributions of a number of perpetuities

$$\int_0^\infty e^{-\Lambda_s} d\Pi_s$$

where  $(\Lambda_s)$  is a particular Lévy process, independent of  $\Pi$ ; see, e.g., Nilsen-Paulsen [17]. In the case  $\Lambda_s = r s$ , the following result seems to go back at least to Harrison [7].

**Proposition 3.1** *The Lévy process  $(C_s^N)$  associated with the self-decomposable random variable  $R_N^2(1)$  in the sense of Theorem 2.1 is*

$$C_s^N = 2 \Pi_{Ns/2} \quad , \quad s \geq 0 .$$

**Proof**

We set  $C_s^N = 2 \Pi_{Ns/2}$ . Then,

$$\mathbb{E} \left[ \exp \left( -\lambda \int_0^\infty e^{-s} dC_s^N \right) \right] = \exp \left( -\frac{N}{2} \int_0^\infty F(2\lambda e^{-s}) ds \right)$$

with, for  $x > 0$ ,

$$F(x) = \int_0^\infty (1 - e^{-tx}) e^{-t} dt = \frac{x}{1+x} .$$

Consequently,

$$\mathbb{E} \left[ \exp \left( -\lambda \int_0^\infty e^{-s} dC_s^N \right) \right] = (1 + 2\lambda)^{-N/2} ,$$

which proves the result. □

By application of Theorem 2.2 we get:

**Corollary 3.1.1** *Let  $\Pi^{(+)}$  and  $\Pi^{(-)}$  two independent copies of the Lévy process  $\Pi$ . Then*

$$U_t^N = 2 \int_{-\frac{N}{2} \log t}^\infty e^{-2s/N} d\Pi_s^{(-)} \quad \text{if } 0 \leq t \leq 1$$

and

$$U_t^N = U_1^N + 2 \int_0^{\frac{N}{2} \log t} e^{2s/N} d\Pi_s^{(+)} \quad \text{if } t \geq 1 .$$

### 3.3 Expression of $(U_t^N)$ from the local time of a perturbed Bessel process

There is by now a wide literature on perturbed Bessel processes, a notion originally introduced by Le Gall-Yor [13, 14], and then studied by Chaumont-Doney [3], Doney-Warren-Yor [5]. We also refer the interested reader to Doney-Zhang [6].

We first introduce the perturbed Bessel process  $(R_{1,\alpha}(t), t \geq 0)$  starting from 0, for  $\alpha < 1$ , as the nonnegative continuous strong solution  $(R_t, t \geq 0)$  of the equation

$$R_t = B_t + \frac{1}{2} L_t(R) + \alpha M_t(R) \quad (1)$$

where  $L_t(R)$  is the semi-martingale local time of  $R$  in 0 at time  $t$ , and

$$M_t(R) = \sup_{0 \leq s \leq t} R_s,$$

$(B_t)$  denoting a standard linear Brownian motion starting from 0. (The strong solution property has been established in Chaumont-Doney [3].)

It is clear that the process  $R_{1,0}$  is nothing else but the Bessel process  $R_1$  (reflected Brownian motion).

We also denote by  $T_t(R)$  the hitting time:

$$T_t(R) = \inf\{s; R_s > t\}.$$

We set  $L_{T_t}(R)$  for  $L_{T_t(R)}(R)$ .

Finally, in the sequel, we set

$$\alpha_N = 1 - \frac{N}{2}.$$

**Proposition 3.2** *For any  $\alpha < 1$ , the process  $(L_{T_t}(R_{1,\alpha}), t \geq 0)$  is a 1-Sato process, and we have*

$$(U_t^N) \stackrel{(d)}{=} (L_{T_t}(R_{1,\alpha_N})).$$

### Proof

By the uniqueness in law of the solution to the equation (1), the process  $R_{1,\alpha}$  is (1/2)-self-similar. As a consequence, the process  $(L_{T_t}(R_{1,\alpha}), t \geq 0)$  is 1-self-similar.

On the other hand, the pair  $(R_{1,\alpha}, M(R_{1,\alpha}))$  is strong Markov (see Doney-Warren-Yor [5, p. 239]). As

$$R_{1,\alpha}(u) = M_u(R_{1,\alpha}) = t \quad \text{if} \quad u = T_t(R_{1,\alpha}),$$

the fact that  $(L_{T_t}(R_{1,\alpha}), t \geq 0)$  is an additive process follows from standard arguments.

Finally, we need to prove:

$$R_N^2(1) \stackrel{d}{=} L_{T_1}(R_{1,\alpha_N}).$$

We denote below  $R_{1,\alpha_N}$  by  $R$ , and  $L_t(R)$ ,  $T_t(R)$ ,  $M_t(R) \dots$  are simply denoted respectively by  $L_t$ ,  $T_t$ ,  $M_t \dots$ . As a particular case of the ‘‘balayage formula’’ (Yor [22]) we deduce from equation (1), that:

$$\begin{aligned} \exp(-\lambda L_t) R_t &= \int_0^t \exp(-\lambda L_s) dR_s \\ &= \int_0^t \exp(-\lambda L_s) dB_s + \frac{1 - \exp(-\lambda L_t)}{2\lambda} + \alpha_N \int_0^t \exp(-\lambda L_s) dM_s. \end{aligned}$$

Hence,

$$\begin{aligned} \exp(-\lambda L_t) (1 + 2\lambda R_t) &= 1 + 2\lambda \int_0^t \exp(-\lambda L_s) dB_s \\ &\quad + 2\lambda \alpha_N \int_0^t \exp(-\lambda L_s) dM_s. \end{aligned}$$

By time changing, we get:

$$\int_0^{T_t} \exp(-\lambda L_s) dM_s = \int_0^t \exp(-\lambda L_{T_u}) du.$$

Therefore, the optional stopping theorem yields:

$$\mathbb{E}[\exp(-\lambda L_{T_t})] (1 + 2\lambda t) = 1 + 2\lambda \alpha_N \int_0^t \mathbb{E}[\exp(-\lambda L_{T_u})] du.$$

Setting

$$\varphi_\lambda(t) = \mathbb{E}[\exp(-\lambda L_{T_t})],$$

we obtain:

$$\varphi_\lambda(t) = \frac{1}{1 + 2\lambda t} + \frac{2\lambda \alpha_N}{1 + 2\lambda t} \int_0^t \varphi_\lambda(u) du.$$

Consequently

$$\varphi_\lambda(t) = (1 + 2\lambda t)^{-N/2}.$$

Therefore,

$$\mathbb{E}[\exp(-\lambda L_{T_1})] = (1 + 2\lambda)^{-N/2} = \mathbb{E}[\exp(-\lambda R_N^2(1))],$$

which proves the desired result. □

## 4 About the process $\left(\int_0^t R_N^2(s) ds, t \geq 0\right)$

### 4.1 A class of Sato processes

Let  $(\ell_t, t \geq 0)$  be the local time in 0 of a linear Brownian motion  $(B_t, t \geq 0)$  starting from 0. We denote, as usual, by  $(\tau_t, t \geq 0)$  the inverse of this local time:

$$\tau_t = \inf\{s \geq 0; \ell_s > t\}.$$

**Proposition 4.1** *Let  $f(x, u)$  be a Borel function on  $\mathbb{R}_+ \times \mathbb{R}_+$  such that*

$$\forall t > 0 \quad \int \int_{\mathbb{R}_+ \times [0, t]} |f(x, u)| dx du < \infty. \quad (2)$$

*Then the process  $A^{(f)}$  defined by:*

$$A_t^{(f)} = \int_0^{\tau_t} f(|B_s|, \ell_s) ds, \quad t \geq 0$$

*is an integrable additive process. Furthermore,*

$$\mathbb{E}[A_t^{(f)}] = 2 \int \int_{\mathbb{R}_+ \times [0, t]} f(x, u) dx du.$$

#### Proof

Assume first that  $f$  is nonnegative. Then,

$$A_t^{(f)} = \sum_{0 \leq u \leq t} \int_{\tau_{u-}}^{\tau_u} f(|B_s|, u) ds.$$

By the theory of excursions (Revuz-Yor [18, Chapter XII, Proposition 1.10]) we have

$$\mathbb{E}[A_t^{(f)}] = \int_0^t du \int n(d\varepsilon) \int_0^{V(\varepsilon)} ds f(|\varepsilon_s|, u)$$

where  $n$  denotes the Itô measure of Brownian excursions and  $V(\varepsilon)$  denotes the life time of the excursion  $\varepsilon$ . The entrance law under  $n$  is given by:

$$n(\varepsilon_s \in dx; s < V(\varepsilon)) = (2\pi s^3)^{-1/2} |x| \exp(-x^2/(2s)) dx.$$

Therefore

$$\mathbb{E}[A_t^{(f)}] = 2 \int_0^t du \int_0^\infty dx f(x, u).$$

The additivity of the process  $A^{(f)}$  follows easily from the fact that, for any  $t \geq 0$ ,  $(B_{\tau_t+s}, s \geq 0)$  is a Brownian motion starting from 0, which is independent of  $\mathcal{B}_{\tau_t}$  (where  $(\mathcal{B}_u)$  is the natural filtration of  $B$ ).

□

**Corollary 4.1.1** *We assume that  $f$  is a Borel function on  $\mathbb{R}_+ \times \mathbb{R}_+$  satisfying (2) and which is  $m$ -homogeneous for  $m > -2$ , meaning that*

$$\forall a > 0, \forall (x, u) \in \mathbb{R}_+ \times \mathbb{R}_+, \quad f(ax, au) = a^m f(x, u).$$

*Then the process  $A^{(f)}$  is a  $(m + 2)$ -Sato process.*

**Proof**

This is a direct consequence of the scaling property of the Brownian motion.

□

## 4.2 A particular case

Let  $N > 0$ . We denote by  $A^{(N)}$  the process  $A^{(f)}$  with

$$f(x, u) = \frac{N^2}{4} 1_{(x \leq \frac{2}{N} u)}.$$

By Proposition 4.1,  $(A_t^{(N)})$  is an integrable process and

$$\mathbb{E}[A_t^{(N)}] = \frac{N t^2}{2}.$$

We now consider the process  $Y_N$  defined by

$$Y_N(t) = \int_0^t R_N^2(s) ds, \quad t \geq 0.$$

**Theorem 4.2** *The process  $A^{(N)}$  is a 2-Sato process and*

$$(Y_N(t)) \stackrel{(1.d)}{=} (A_t^{(N)}).$$

**Proof**

It is a direct consequence of Corollary 4.2.1 that  $A^{(N)}$  is a 2-Sato process.

By Mansuy-Yor [16, Theorem 3.4, p.38], the following extension of the Ray-Knight theorem holds:

For any  $u > 0$ ,

$$(L_{\tau_u}^{a-(2u/N)}, 0 \leq a \leq (2u/N)) \stackrel{(d)}{=} (R_N^2(a), 0 \leq a \leq (2u/N))$$

where  $L_t^x$  denotes the local time of the semi-martingale  $(|B_s| - \frac{2}{N} \ell_s, s \geq 0)$  in  $x$  at time  $t$ .

We remark that

$$s \in [0, \tau_t] \implies |B_s| - \frac{2}{N} \ell_s \geq -\frac{2t}{N}.$$

Therefore, the occupation times formula entails:

$$A_t^{(N)} = \frac{N^2}{4} \int_{-2t/N}^0 L_{\tau_t}^x dx = \frac{N^2}{4} \int_0^{2t/N} L_{\tau_t}^{x-(2t/N)} dx.$$

Thus, by the above mentioned extension of the Ray-Knight theorem,

$$(A_t^{(N)}) \stackrel{(1.d)}{=} \left( \frac{N^2}{4} \int_0^{2t/N} R_N^2(s) ds \right).$$

The scaling property of  $R_N$  also yields the identity in law:

$$(A_t^{(N)}) \stackrel{(1.d)}{=} \left( \int_0^t R_N^2(s) ds \right),$$

and the result follows from the definition of  $Y_N$ . □

We may now apply Proposition 2.3 to get:

**Corollary 4.2.1** *The process  $V_N$  defined by:*

$$V_N(t) = Y_N(t) - \frac{Nt^2}{2}, \quad t \geq 0$$

*is a PCOC and an associated martingale is  $M_N$  defined by:*

$$M_N(t) = A_t^{(N)} - \frac{Nt^2}{2}, \quad t \geq 0.$$

*Moreover,  $M_N$  is a centered 2-Sato process.*

### 4.3 Representation of $A^{(N)}$ as a process of hitting times

**Theorem 4.3** *The process  $A^{(N)}$  is identical in law to the process*

$$T_t(R_{1,\alpha_N}) \quad , \quad t \geq 0$$

where  $R_{1,\alpha_N}$  denotes the perturbed Bessel process defined in Subsection 3.3 and

$$T_t(R_{1,\alpha_N}) = \inf\{s ; R_{1,\alpha_N}(s) > t\} .$$

The proof can be found in Le Gall-Yor [14]. Nevertheless, for the convenience of the reader, we give again the proof below. A more general result, based on Doney-Warren-Yor [5], shall also be stated in the next section.

#### Proof

In this proof, we adopt the following notation:  $(B_t)$  still denotes a standard linear Brownian motion starting from 0,  $S_t = \sup_{0 \leq s \leq t} B_s$  and  $\sigma_t = \inf\{s ; B_s > t\}$ . Moreover, for  $a < 1$  and  $t \geq 0$ , we set

$$X_t^a = \int_0^t 1_{(B_s > a S_s)} ds \quad \text{and} \quad Z_t^a = \inf\{s ; X_s^a > t\} .$$

**Lemma 4.3.1** *Let  $a < 1$ . Then*

$$\sup_{0 \leq s \leq t} (B_s - a S_s)^+ = (1 - a) S_t .$$

#### Proof

Since  $a < 1$ , we have, for  $0 \leq s \leq t$ ,

$$(B_s - a S_s)^+ \leq (1 - a) S_s \leq (1 - a) S_t .$$

Moreover, there exists  $s_t \in [0, t]$  such that  $B_{s_t} = S_t$  and therefore  $S_{s_t} = S_t$ . Hence,  $B_{s_t} - a S_{s_t} = (1 - a) S_t$ . □

**Lemma 4.3.2** *Let  $a < 1$  and  $\alpha = -a/(1 - a)$ . We set*

$$R_t^a = (B_t - a S_t)^+ \quad \text{and} \quad U_t^a = R_{Z_t^a}^a .$$

*Then the processes  $U^a$  and  $R_{1,\alpha}$  are identical in law.*



**Proof**

By Tanaka's formula,

$$R_t^a = \int_0^t 1_{(R_s^a > 0)} dR_s^a + \frac{1}{2} L_t(R^a)$$

where  $L_t(R^a)$  denotes the local time of the semi-martingale  $R^a$  in 0 at time  $t$ . Now,

$$\int_0^t 1_{(R_s^a > 0)} dR_s^a = \int_0^t 1_{(B_s - a S_s > 0)} d(B_s - a S_s).$$

If  $s > 0$  belongs to the support of  $dS_s$ , then  $B_s = S_s$  and, since  $a < 1$ ,  $B_s - a S_s > 0$ . Therefore,

$$R_t^a = \int_0^t 1_{(B_s - a S_s > 0)} dB_s - a S_t + \frac{1}{2} L_t(R^a).$$

By Lemma 4.3.1,  $-a S_t = \alpha M_t(R^a)$  where

$$M_t(R^a) = \sup_{0 \leq s \leq t} R_s^a.$$

Consequently,

$$U_t^a = \int_0^{Z_t^a} 1_{(B_s - a S_s > 0)} dB_s + \frac{1}{2} L_{Z_t^a}(R^a) + \alpha M_{Z_t^a}(R^a).$$

The process

$$\int_0^{Z_t^a} 1_{(B_s - a S_s > 0)} dB_s \quad , \quad t \geq 0$$

is a continuous martingale whose bracket is  $t$ , therefore it is a Brownian motion.

On the other hand, it is easy to see that

$$L_{Z_t^a}(R^a) = L_t(U^a) \quad \text{and} \quad M_{Z_t^a}(R^a) = M_t(U^a).$$

Therefore, the process  $U^a$  is a solution to equation (1), which obviously is continuous and nonnegative. □

By Lévy's theorem, the process  $A^{(N)}$  is identical in law to the process

$$\frac{N^2}{4} \int_0^{\sigma_t} 1_{(B_s > (1 - \frac{2}{N}) S_s)} ds \quad , \quad t \geq 0.$$

By the scaling property of  $B$ , the above process has the same law as

$$\int_0^{\sigma_{Nt/2}} 1_{(B_s > (1 - \frac{2}{N})S_s)} ds = X_{\sigma_{Nt/2}}^{1 - \frac{2}{N}} \quad , \quad t \geq 0 .$$

Now,

$$X_{\sigma_{Nt/2}}^{1 - \frac{2}{N}} = \inf\{X_u^{1 - \frac{2}{N}} ; S_u > \frac{Nt}{2}\}$$

and, by Lemma 4.3.1,

$$X_{\sigma_{Nt/2}}^{1 - \frac{2}{N}} = \inf\{X_u^{1 - \frac{2}{N}} ; R_u^{1 - \frac{2}{N}} > t\} = \inf\{v ; U_v^{1 - \frac{2}{N}} > t\} .$$

The result then follows from Lemma 4.3.2. □

**Corollary 4.3.1** *The process*

$$T_t(R_{1, \alpha_N}) \quad , \quad t \geq 0$$

*is a 2-Sato process and*

$$\left( \int_0^t R_N^2(s) ds \right) \stackrel{(1.d)}{=} (T_t(R_{1, \alpha_N})) .$$

## 5 About the process

$$\left( \frac{1}{K^2} \int_0^t s^{\frac{2(1-K)}{K}} R_N^2(s) ds \quad , \quad t \geq 0 \right)$$

In this section we extend Corollary 4.3.1. We fix two positive real numbers  $N$  and  $K$ . We first recall some important results on general perturbed Bessel processes  $R_{K, \alpha}$  with  $\alpha < 1$ .

### 5.1 Perturbed Bessel processes

We follow, in this subsection, Doney-Warren-Yor [5]. We first recall the definition of the process  $R_{K, \alpha}$  with  $K > 0$  and  $\alpha < 1$ .

The case  $K = 1$  was already introduced in Subsection 3.3. For  $K > 1$ ,  $R_{K, \alpha}$  is defined as a continuous nonnegative solution to

$$R_t = B_t + \frac{K-1}{2} \int_0^t \frac{1}{R_s} ds + \alpha M_t(R) , \quad (3)$$

and, for  $0 < K < 1$ ,  $R_{K,\alpha}$  is defined as the square root of a continuous nonnegative solution to

$$X_t = 2 \int_0^t \sqrt{X_s} dB_s + K t + \alpha M_t(X). \quad (4)$$

We note that, for any  $K > 0$ ,  $(R_{K,0}(t)) \stackrel{(d)}{=} (R_K(t))$ . As in the case  $K = 1$ , for any  $K > 0$ , the pair  $(R_{K,\alpha}, M(R_{K,\alpha}))$  is strong Markov.

We denote, as before,

$$T_t(R_{K,\alpha}) = \inf\{s ; R_{K,\alpha}(s) > t\}.$$

The following theorem, due to Doney-Warren-Yor ([5, Theorem 5.2, p. 246]) is an extension of the Ciesielski-Taylor theorem and of the Ray-Knight theorem.

**Theorem 5.1** 1)

$$\int_0^\infty 1_{(R_{K+2,\alpha}(s) \leq 1)} ds \stackrel{d}{=} T_1(R_{K,\alpha})$$

2)

$$(L_\infty^a(R_{K+2,\alpha}), a \geq 0) \stackrel{(d)}{=} \left( \frac{a^{1-K}}{K} R_{2(1-\alpha)}(a^K), a \geq 0 \right)$$

## 5.2 Identification of the Sato process associated to $Y_{N,K}$

We denote, for  $N > 0$  and  $K > 0$ , by  $Y_{N,K}$  the process:

$$Y_{N,K}(t) = \frac{1}{K^2} \int_0^t s^{\frac{2(1-K)}{K}} R_N^2(s) ds, \quad t \geq 0.$$

We also recall the notation:

$$\alpha_N = 1 - \frac{N}{2}.$$

**Theorem 5.2** *The process*

$$T_{t^{1/K}}(R_{K,\alpha_N}), \quad t \geq 0$$

*is a  $(2/K)$ -Sato process and*

$$(Y_{N,K}(t)) \stackrel{(1,d)}{=} (T_{t^{1/K}}(R_{K,\alpha_N})).$$

**Proof**

In the following proof, we denote  $R_{K,\alpha_N}$  simply by  $R$ , and we set  $T_t$  and  $M_t$  for, respectively,  $T_t(R)$  and  $M_t(R)$ .

The first part of the statement follows from the  $(1/2)$ -self-similarity of  $R$  and from the strong Markovianity of  $(R, M)$ , taking into account that, for any  $t \geq 0$ ,

$$R_{T_t} = M_{T_t} = t.$$

By occupation times formula, we deduce from 1) in Theorem 5.1,

$$\int_0^1 L_\infty^x(R_{K+2,\alpha_N}) dx \stackrel{d}{=} T_1.$$

Using then 2) in Theorem 5.1, we obtain:

$$\int_0^1 L_\infty^x(R_{K+2,\alpha_N}) dx \stackrel{d}{=} \int_0^1 \frac{x^{1-K}}{K} R_N^2(x^K) dx.$$

By change of variable, the last integral is equal to  $Y_{N,K}(1)$ , and hence,

$$Y_{N,K}(1) \stackrel{d}{=} T_1.$$

The final result now follows by self-similarity. □

**Corollary 5.2.1** *The process*

$$V_{N,K}(t) := Y_{N,K}(t) - \frac{N}{2K} t^{2/K}, \quad t \geq 0$$

is a PCOC, and an associated martingale is

$$M_{N,K}(t) := T_{t^{1/K}}(R_{K,\alpha_N}) - \frac{N}{2K} t^{2/K}, \quad t \geq 0,$$

which is a centered  $(2/K)$ -Sato process.

Finally, we have proven, in particular, that for any  $\rho > -2$  and any  $N > 0$ , the random variable

$$\int_0^1 s^\rho R_N^2(s) ds$$

is self-decomposable. This result will be generalized and made precise in the next section, using completely different arguments.

## 6 About the random variables $\int R_N^2(s) d\mu(s)$

In this section, we fix a measure  $\mu$  on  $\mathbb{R}_+^* = (0, \infty)$  such that

$$\int_{\mathbb{R}_+^*} s d\mu(s) < \infty.$$

### 6.1 Spectral study of an operator

We associate with  $\mu$  an operator  $K^{(\mu)}$  on  $E = L^2(\mu)$  defined by

$$\forall f \in E \quad K^{(\mu)} f(t) = \int_{\mathbb{R}_+^*} f(s) t \wedge s d\mu(s)$$

where  $\wedge$  denotes the infimum. Though the spectral study of this operator is certainly classical, we give the details for the convenience of the reader.

**Lemma 6.1** *The operator  $K^{(\mu)}$  is a nonnegative symmetric Hilbert-Schmidt operator.*

**Proof**

As a consequence of the obvious inequality:

$$(t \wedge s)^2 \leq t s,$$

we get

$$\int \int_{(\mathbb{R}_+^*)^2} (t \wedge s)^2 d\mu(t) d\mu(s) \leq \left( \int_{\mathbb{R}_+^*} s d\mu(s) \right)^2,$$

and therefore  $K^{(\mu)}$  is a Hilbert-Schmidt operator.

On the other hand, denoting by  $(\bullet, \bullet)_E$  the scalar product in  $E$ , we have:

$$(K^{(\mu)} f, g)_E = \mathbb{E} \left[ \int f(t) B_t d\mu(t) \int g(t) B_t d\mu(t) \right]$$

where  $B$  is a standard Brownian motion starting from 0. This entails that  $K^{(\mu)}$  is nonnegative symmetric. □

**Lemma 6.2** *Let  $\lambda \in \mathbb{R}$ . Then  $\lambda$  is an eigenvalue of  $K^{(\mu)}$  if and only if  $\lambda > 0$  and there exists  $f \in L^2(\mu)$ ,  $f \neq 0$ , such that:*

i)

$$\lambda f'' + f \cdot \mu = 0 \quad \text{in the distribution sense on } \mathbb{R}_+^* \quad (5)$$

ii)  $f$  admits a representative which is absolutely continuous on  $\mathbb{R}_+$ ,  $f'$  admits a representative which is right-continuous on  $\mathbb{R}_+^*$ ;  
(In the sequel,  $f$  and  $f'$  respectively always denote such representatives.)

iii)

$$f(0) = 0 \quad \text{and} \quad \lim_{t \rightarrow \infty} f'(t) = 0.$$

**Proof**

Let  $f \in L^2(\mu)$  and  $g = K^{(\mu)}f$ . We have, for  $\mu$ -a.e.  $t > 0$ ,

$$g(t) = \int_0^t du \int_{(u, \infty)} f(s) d\mu(s). \quad (6)$$

Thus  $g$  admits a representative (still denoted by  $g$ ) which is absolutely continuous on  $\mathbb{R}_+$  and  $g(0) = 0$ . Moreover,  $g'$  admits a representative which is right-continuous on  $\mathbb{R}_+^*$  and is given by:

$$g'(t) = \int_{(t, \infty)} f(s) d\mu(s). \quad (7)$$

In particular

$$|g'(t)| \leq t^{-1/2} \left[ \int_{(t, \infty)} f^2(s) d\mu(s) \int_{(t, \infty)} s d\mu(s) \right]^{1/2}. \quad (8)$$

Hence:

$$\lim_{t \rightarrow \infty} g'(t) = 0.$$

Besides, (7) entails:

$$g'' + f \cdot \mu = 0 \quad \text{in the distribution sense on } \mathbb{R}_+^*.$$

Consequently, 0 is not an eigenvalue of  $K^{(\mu)}$  and the “only if” part is proven.

Conversely, let  $f \in L^2(\mu)$ ,  $f \neq 0$ , and  $\lambda > 0$  such that properties i),ii),iii) hold. Then

$$\lambda f'(t) = \int_{(t, \infty)} f(s) d\mu(s).$$

Hence

$$\lambda f(t) = \int_0^t du \int_{(u, \infty)} f(s) d\mu(s) = K^{(\mu)}f(t),$$

which proves the “if” part. □

We note that, since 0 is not an eigenvalue of  $K^{(\mu)}$ ,  $K^{(\mu)}$  is actually a *positive* symmetric operator. On the other hand, by the previous proof, the functions  $f \in L^2(\mu)$ ,  $f \neq 0$ , satisfying properties i),ii),iii) in the statement of Lemma 6.2, are the eigenfunctions of the operator  $K^{(\mu)}$  corresponding to the eigenvalue  $\lambda > 0$ .

**Lemma 6.3** *Let  $f$  be an eigenfunction of  $K^{(\mu)}$ . Then,*

$$|f(t)| = o(t^{1/2}) \quad \text{and} \quad |f'(t)| = o(t^{-1/2})$$

when  $t$  tends to  $\infty$ .

**Proof**

This is a direct consequence of (8). □

**Lemma 6.4** *Let  $f_1$  and  $f_2$  be eigenfunctions of  $K^{(\mu)}$  with respect to the same eigenvalue. Then,*

$$\forall t > 0 \quad f_1'(t) f_2(t) - f_1(t) f_2'(t) = 0.$$

**Proof**

By (5),

$$(f_1' f_2 - f_1 f_2')' = 0 \quad \text{in the sense of distributions on } \mathbb{R}_+^*.$$

By right-continuity, there exists  $C \in \mathbb{R}$  such that

$$\forall t > 0 \quad f_1'(t) f_2(t) - f_1(t) f_2'(t) = C.$$

Letting  $t$  tend to  $\infty$ , we deduce from Lemma 6.3 that  $C = 0$ . □

**Lemma 6.5** *Let  $f$  be a solution of (5) with  $\lambda > 0$ , and let  $a > 0$ . We assume as previously that  $f$  (resp.  $f'$ ) denotes the representative which is absolutely continuous (resp. right-continuous) on  $\mathbb{R}_+^*$ . If  $f(a) = f'(a) = 0$ , then, for any  $t \geq a$ ,  $f(t) = 0$ .*

**Proof**

This lemma is quite classical if the measure  $\mu$  admits a density with respect to the Lebesgue measure. The proof may be easily adapted to this more general case. □

We are now able to state the main result of this section.

**Theorem 6.6** *The operator  $K^{(\mu)}$  is a positive symmetric compact operator whose all eigenvalues are simple (i.e. the dimension of the eigenspaces is 1).*

**Proof**

It only remains to prove that the eigenvalues are simple. Let then  $\lambda > 0$  be an eigenvalue and let  $f_1$  and  $f_2$  be eigenfunctions with respect to this eigenvalue. Let  $a > 0$  with  $\mu(\{a\}) = 0$ . By Lemma 6.4, there exist  $c_1$  and  $c_2$  with  $c_1^2 + c_2^2 > 0$  such that, setting  $f = c_1 f_1 + c_2 f_2$ , we have

$$f(a) = f'(a) = 0.$$

By Lemma 6.5,  $f(t) = 0$  for any  $t \geq a$ . But, since  $\mu(\{a\}) = 0$ ,  $f'$  also is left-continuous at  $a$ . Then, we may reason on  $(0, a]$  as on  $[a, \infty)$  and therefore we also have  $f(t) = 0$  for  $0 < t \leq a$ . Finally,

$$c_1 f_1 + c_2 f_2 = 0,$$

which proves the result. □

In the following, we denote by  $\lambda_1 > \lambda_2 > \dots$  the decreasing sequence (possibly finite) of the eigenvalues of  $K^{(\mu)}$ . Of course, this sequence depends on  $\mu$ , which we omit in the notation. As  $K^{(\mu)}$  is Hilbert-Schmidt,

$$\sum_{n \geq 1} \lambda_n^2 < \infty.$$

It will be shown in Subsection 6.3 (see Theorem 6.7) that actually

$$\sum_{n \geq 1} \lambda_n < \infty,$$

i.e.  $K^{(\mu)}$  is trace-class. The following corollary plays an essential role in the sequel.

**Corollary 6.6.1** *There exists a Hilbert basis  $(f_n)_{n \geq 1}$  in  $L^2(\mu)$  such that*

$$\forall n \geq 1 \quad K^{(\mu)} f_n = \lambda_n f_n.$$



## 6.2 Examples

In this subsection, we consider two particular types of measures  $\mu$ .

**6.2.1**  $\mu = \sum_{j=1}^n a_j \delta_{t_j}$

Let  $a_1, \dots, a_n$  positive real numbers and  $0 < t_1 < \dots < t_n$ . We denote by  $\delta_t$  the Dirac measure at  $t$  and we consider, in this paragraph,

$$\mu = \sum_{j=1}^n a_j \delta_{t_j}.$$

By the previous study, the sequence of eigenvalues of  $K^{(\mu)}$  is finite if and only if the space  $L^2(\mu)$  is finite dimensional, that is if  $\mu$  is of the above form. In this case, the eigenvalues of  $K^{(\mu)}$  are the eigenvalues of the matrix  $(m_{i,j})_{1 \leq i,j \leq n}$  with

$$m_{i,j} = \sqrt{a_i a_j} t_{i \wedge j}.$$

In particular, by the previous study, such a matrix has  $n$  distinct eigenvalues, which are  $> 0$ .

**6.2.2**  $\mu = C t^\rho 1_{(0,1]}(t) dt$

In this paragraph, we consider

$$\mu = C t^\rho 1_{(0,1]}(t) dt$$

with  $C > 0$  and  $\rho > -2$ . By Lemma 6.2, the eigenfunctions  $f$  of  $K^{(\mu)}$  associated with  $\lambda > 0$  are characterized by:

$$\lambda f''(x) + C x^\rho f(x) = 0 \quad \text{on } (0, 1), \quad (9)$$

$$f(0) = 0 \quad , \quad f'(1) = 0.$$

We set  $\sigma = (\rho + 2)^{-1}$  and  $\nu = \sigma - 1$ . For  $a > -1$ , we recall the definition of the Bessel function  $J_a$ :

$$J_a(x) = \sum_{k=0}^{\infty} \frac{(-1)^k (x/2)^{a+2k}}{k! \Gamma(a+k+1)}.$$

Then, the only function  $f$  satisfying (9) and  $f(0) = 0$  is, up to a multiplicative constant,

$$f(x) = x^{1/2} J_\sigma \left( 2\sigma \sqrt{\frac{C}{\lambda}} x^{1/2\sigma} \right).$$

We deduce from the equality, valid for  $a > 1$ ,

$$a J_a(x) + x J'_a(x) = x J_{a-1}(x)$$

that  $f'(1) = 0$  if and only if

$$J_\nu \left( 2\sigma \sqrt{\frac{C}{\lambda}} \right) = 0.$$

Denote by  $(j_{\nu,k}, k \geq 1)$  the sequence of the positive zeros of  $J_\nu$ . Then the sequence  $(\lambda_k, k \geq 1)$  of eigenvalues of  $K^{(\mu)}$  is given by:

$$\lambda_k = 4C(\nu + 1)^2 j_{\nu,k}^{-2}, \quad k \geq 1.$$

**Particular case** Suppose  $\rho = 0$ . Then  $\nu = -1/2$  and

$$J_\nu(x) = J_{-1/2}(x) = \left( \frac{2}{\pi x} \right)^{1/2} \cos(x).$$

Hence,

$$\lambda_k = 4C\pi^{-2}(2k-1)^{-2}, \quad k \geq 1.$$

### 6.3 Representation of $\int B_s^2 d\mu(s)$

We again consider the general setting defined in Subsection 6.1, the notation of which we keep.

In this subsection, we study the random variable

$$Y_1^{(\mu)} := \int B_s^2 d\mu(s).$$

The use of the operator  $K^{(\mu)}$  and of its spectral decomposition in the type of study we develop below, is called the Karhunen-Loeve decompositions method. It has a long history which goes back at least to Kac-Siegert [10, 11]. We also refer to the recent paper [4] and to the references therein.

**Theorem 6.7** *The eigenvalues  $(\lambda_k, k \geq 1)$  of the operator  $K^{(\mu)}$  satisfy*

$$\sum_{k \geq 1} \lambda_k = \int_{\mathbb{R}_+^*} t d\mu(t) (< \infty, \text{ by hypothesis}).$$

*Moreover, there exists a sequence  $(\Gamma_n, n \geq 1)$  of independent normal variables such that:*

$$Y_1^{(\mu)} \stackrel{d}{=} \sum_{n \geq 1} \lambda_n \Gamma_n^2.$$

**Proof**

We deduce from Corollary 6.6.1, by the Bessel-Parseval equality,

$$Y_1^{(\mu)} = \sum_{n \geq 1} \left( \int B_s f_n(s) \, d\mu(s) \right)^2 \quad \text{a.s.}$$

Taking the expectation, we get

$$\int_{\mathbb{R}_+^*} t \, d\mu(t) = \sum_{n \geq 1} (K^{(\mu)} f_n, f_n)_E = \sum_{n \geq 1} \lambda_n.$$

We set, for  $n \geq 1$ ,

$$\Gamma_n = \frac{1}{\sqrt{\lambda_n}} \int B_s f_n(s) \, d\mu(s).$$

Then  $(\Gamma_n, n \geq 1)$  is a Gaussian sequence and

$$\mathbb{E}[\Gamma_n \Gamma_m] = \frac{1}{\sqrt{\lambda_n \lambda_m}} (K^{(\mu)} f_n, f_m)_E = \delta_{n,m}$$

where  $\delta_{n,m}$  denotes Kronecker's symbol. Hence, the result follows. □

**Corollary 6.7.1** *The Laplace transform of  $Y_1^{(\mu)}$  is*

$$F_1^{(\mu)}(t) = \prod_{n \geq 1} (1 + 2t \lambda_n)^{-1/2}.$$

**Proof**

This is a direct consequence of the previous theorem, taking into account that, if  $\Gamma$  is a normal variable, then

$$\Gamma^2 \stackrel{d}{=} 2 \gamma_{1/2}.$$

□

## 6.4 Representation of $\int R_N^2(s) d\mu(s)$

We now consider the random variable

$$Y_N^{(\mu)} := \int R_N^2(s) d\mu(s).$$

**Theorem 6.8** *There exists a sequence  $(\Theta_{N,n}, n \geq 1)$  of independent variables with, for any  $n \geq 1$ ,*

$$\Theta_{N,n} \stackrel{d}{=} R_N^2(1) \stackrel{d}{=} 2\gamma_{N/2}$$

such that

$$Y_N^{(\mu)} \stackrel{d}{=} \sum_{n \geq 1} \lambda_n \Theta_{N,n}. \quad (10)$$

Moreover, the Laplace transform of  $Y_N^{(\mu)}$  is

$$F_N^{(\mu)}(t) = \prod_{n \geq 1} (1 + 2t\lambda_n)^{-N/2}. \quad (11)$$

### Proof

It is clear, for instance from Revuz-Yor [18, Chapter XI, Theorem 1.7], that

$$F_N^{(\mu)}(t) = [F_1^{(\mu)}(t)]^N.$$

Therefore, (11) holds and (10) follows directly. □

**Corollary 6.8.1** *The random variable  $Y_N^{(\mu)}$  is self-decomposable. The function  $h$ , which is decreasing on  $(0, \infty)$  and associated with  $Y_N^{(\mu)}$  in Theorem 2.1, is*

$$h(x) = \frac{N}{2} \sum_{n \geq 1} \exp\left(-\frac{1}{2\lambda_n} x\right).$$

As a consequence, following Bondesson [1], we see that  $Y_N^{(\mu)}$  is a generalized gamma convolution (GGC) whose Thorin measure is the discrete measure:

$$\frac{N}{2} \sum_{n \geq 1} \delta_{1/2\lambda_n}.$$

**Particular case** We consider here, as in Section 5, the particular case:

$$\mu = \frac{1}{K^2} t^{\frac{2(1-K)}{K}} 1_{(0,1]}(t) dt.$$

Then,  $Y_N^{(\mu)}$  is the random variable  $Y_{N,K}(1)$  studied in Section 5. As a consequence of Paragraph 6.2.2 with

$$C = \frac{1}{K^2} \quad \text{and} \quad \rho = \frac{2}{K} - 2,$$

we have

$$\lambda_k = j_{\nu,k}^{-2}, \quad k \geq 1$$

with  $\nu = \frac{K}{2} - 1$ . Moreover, by Theorem 5.2,

$$Y_2^{(\mu)} \stackrel{d}{=} T_1(R_K).$$

It is known (see for instance Borodin-Salminen [2, formula 2.0.1, p. 387]) that

$$\mathbb{E}[\exp(-t T_1(R_K))] = \frac{2^{-\nu}}{\Gamma(\nu + 1)} \frac{(\sqrt{2t})^\nu}{I_\nu(\sqrt{2t})}$$

where  $I_\nu$  denotes the modified Bessel function:

$$I_\nu(x) = \sum_{k=0}^{\infty} \frac{(x/2)^{\nu+2k}}{k! \Gamma(\nu + k + 1)}.$$

We set:

$$\widehat{I}_\nu(x) = \sum_{k=0}^{\infty} \frac{(x/2)^{2k}}{k! \Gamma(\nu + k + 1)}.$$

Therefore, by formula (11) in the case  $N = 2$ , we recover the following representation:

$$\widehat{I}_\nu(x) = \frac{1}{\Gamma(\nu + 1)} \prod_{k \geq 1} \left( 1 + \frac{x^2}{j_{\nu,k}^2} \right).$$

In particular ( $\nu = -1/2$ ),

$$\cosh(x) = \prod_{k \geq 1} \left( 1 + \frac{4x^2}{\pi^2 (2k - 1)^2} \right).$$

Likewise we obtain, for  $\nu = 1/2$ ,

$$\frac{\sinh(x)}{x} = \prod_{k \geq 1} \left( 1 + \frac{x^2}{\pi^2 k^2} \right).$$

## 6.5 Sato process associated to $Y_N^{(\mu)}$

**Theorem 6.9** *Let  $(U_t^N)$  be the 1-Sato process associated to  $R_N^2(1)$  (cf. Section 3). Then, the 1-Sato process associated to  $Y_N^{(\mu)}$  is  $(U_t^{(N,\mu)})$  defined by:*

$$U_t^{(N,\mu)} = \sum_{n \geq 1} \lambda_n U_t^{N,n} \quad , \quad t \geq 0$$

where  $((U_t^{N,n}) , n \geq 1)$  denotes a sequence of independent processes such that, for  $n \geq 1$ ,

$$(U_t^{N,n}) \stackrel{(d)}{=} (U_t^N).$$

### Proof

This is a direct consequence of Theorem 6.8. □

**Corollary 6.9.1** *The process*

$$V_t^{(N,\mu)} := \int_{\mathbb{R}_+^*} (R_N^2(t s) - N t s) d\mu(s) \quad , \quad t \geq 0$$

is a PCOC and an associated martingale is

$$M_t^{(N,\mu)} := U_t^{(N,\mu)} - N t \int_{\mathbb{R}_+^*} s d\mu(s) \quad , \quad t \geq 0.$$

The above martingale  $(M_t^{(N,\mu)})$  is purely discontinuous. We also may associate to the PCOC  $(V_t^{(N,\mu)})$  a continuous martingale, as we now state.

**Theorem 6.10** *A continuous martingale associated to the PCOC  $(V_t^{(N,\mu)})$  is*

$$\sum_{n \geq 1} \lambda_n ((R_N^{(n)})^2(t) - N t) \quad , \quad t \geq 0$$

where  $((R_N^{(n)}(t)) , n \geq 1)$  denotes a sequence of independent processes such that, for  $n \geq 1$ ,

$$(R_N^{(n)}(t)) \stackrel{(d)}{=} (R_N(t)).$$

**Proof**

This is again a direct consequence of Theorem 6.8. □

We can also explicit the relation between  $U^{(N,\mu)}$  and  $U^{(N',\mu)}$ . Let  $C^{(N,\mu)}$  (resp.  $C^{(N',\mu)}$ ) be the Lévy process associated with  $Y_N^{(\mu)}$  (resp.  $Y_{N'}^{(\mu)}$ ). We see, by Laplace transform, that

$$(C_s^{(N',\mu)}) \stackrel{(d)}{=} (C_{N's/N}^{(N,\mu)}).$$

Then, using the relations between the processes  $U$  and  $C$  given in Theorem 2.2, we obtain:

**Proposition 6.11** *We have:*

$$(U_t^{(N',\mu)}, t \geq 0) \stackrel{(d)}{=} \left( \int_0^{t^{N'/N}} s^{\frac{N-N'}{N'}} dU_s^{(N,\mu)}, t \geq 0 \right).$$

**Corollary 6.11.1** *For  $N > 0$  and  $K > 0$ , we set, with the notation of Section 5,*

$$T_t^{N,K} = T_t(R_{K,\alpha_N}) \quad , \quad t \geq 0.$$

*Then, for  $N > 0$ ,  $N' > 0$  and  $K > 0$ , for any  $t \geq 0$ ,*

$$T_t^{N',K} = \int_0^{t^{N'/N}} s^{2\frac{N-N'}{N'}} dT_s^{N,K}.$$

**Proof**

By Theorem 5.2,  $(T_{t^{1/2}}^{N,K})$  is the 1-Sato process associated with  $Y_N^{(\mu)}$  defined from

$$\mu = \frac{1}{K^2} t^{\frac{2(1-K)}{K}} 1_{(0,1]}(t) dt.$$

□

## 7 Some negative results

### 7.1 Squared Bessel process started from $x > 0$

Let  $Y_u^{N,x}$  be the value at time  $u > 0$  of the squared Bessel process of dimension  $N \geq 0$ , starting from  $x \geq 0$ .

**Proposition 7.1** *The random variable  $Y_u^{N,x}$  is self-decomposable if and only if  $x^2 \leq Nu$ .*

**Proof**

One has:

$$\mathbb{E}[\exp(-t Y_u^{N,x})] = (1 + 2tu)^{-N/2} \exp\left(-\frac{x^2 t}{1 + 2tu}\right).$$

It is then easy to see that  $Y_u^{N,x}$  is infinitely divisible and its Lévy measure admits on  $(0, \infty)$  the density

$$\varphi(y) = \left(N + \frac{x^2}{2u^2} y\right) \frac{1}{2y} e^{-y/2u}.$$

Hence, the result follows from the characterisation 1) in Theorem 2.1. □

### 7.2 Pairs of values of a squared Bessel process

We now consider, for  $N > 0$ ,  $t_1, t_2 > 0$ , the  $\mathbb{R}^2$ -valued random variable

$$Y := (R_N^2(t_1), R_N^2(t_1 + t_2)).$$

(Recall that  $R_N(0) = 0$ .)

For such an  $\mathbb{R}^2$ -valued random variable, we can also define the notion of self-decomposability as in Section 2. Theorem 2.1, suitably modified, is still valid.

**Proposition 7.2** *The  $\mathbb{R}^2$ -valued random variable  $Y$  is not self-decomposable.*

**Proof**

An easy computation gives

$$\mathbb{E}[\exp(-\lambda_1 R_N^2(t_1) - \lambda_2 R_N^2(t_1 + t_2))] = [P(\lambda)]^{-N/2}$$



with

$$P(\lambda) = 1 + 2\lambda_1 t_1 + 2\lambda_2(t_1 + t_2) + 4\lambda_1 \lambda_2 t_1 t_2.$$

If  $Y$  were self-decomposable, we would have

$$\log(P(\lambda)) = \int \int_{(\mathbb{R}_+^*)^2} (1 - \exp(-\lambda_1 x_1 - \lambda_2 x_2)) \frac{H(x_1, x_2)}{x_1^2 + x_2^2} dx_1 dx_2$$

with  $H$  a decreasing function on each half line with origin  $(0, 0)$ . Taking the derivative with respect to  $\lambda_1$ , we get

$$\frac{2t_1(1 + 2\lambda_2 t_2)}{P(\lambda)} = \int \int_{(\mathbb{R}_+^*)^2} \exp(-\lambda_1 x_1 - \lambda_2 x_2) \frac{x_1 H(x_1, x_2)}{x_1^2 + x_2^2} dx_1 dx_2.$$

Letting  $\lambda_2$  tend to  $\infty$ , we obtain

$$\frac{2t_1 t_2}{t_1 + t_2 + 2\lambda_1 t_1 t_2} = 0$$

which yields a contradiction. □

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