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# On the convergence of LePage series in Skorokhod space

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### 1. Introduction

We are interested in the convergence in the Skorokhod space  $\mathbb{D}^d = \mathbb{D}([0, 1], \mathbb{R}^d)$  endowed with the  $J_1$ -topology of random series of the form

$$X(t) = \sum_{i=1}^{\infty} \Gamma_i^{-1/\alpha} \varepsilon_i Y_i(t), \quad t \in [0, 1],$$
(1)

where  $\alpha \in (0, 2)$  and:

 $\infty$ 

-  $(\Gamma_i)_{i>1}$  is the increasing enumeration of the points of a Poisson point process on  $[0, +\infty)$  with Lebesgue intensity;

- $(\varepsilon_i)_{i\geq 1}$  is an i.i.d. sequence of real random variables;
- $(Y_i)_{i\geq 1}$  is an i.i.d. sequence of  $\mathbb{D}^d$ -valued random variables;
- the sequences  $(\Gamma_i)$ ,  $(\varepsilon_i)$  and  $(Y_i)$  are independent.

Note that a more constructive definition for the sequence  $(\Gamma_i)_{i>1}$  is given by

$$\Gamma_i = \sum_{j=1}^l \gamma_j, \quad i \ge 1,$$

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## ABSTRACT

We consider the problem of the convergence of the so-called LePage series in the Skorokhod space  $\mathbb{D}^d = \mathbb{D}([0, 1], \mathbb{R}^d)$  and provide a simple criterion based on the moments of the increments of the random process involved in the series. This provides a simple sufficient condition for the existence of an  $\alpha$ -stable distribution on  $\mathbb{D}^d$  with given spectral measure. © 2011 Elsevier B.V. All rights reserved.





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where  $(\gamma_i)_{i \ge 1}$  is an i.i.d. sequence of random variables with exponential distribution of parameter 1, and independent of  $(\varepsilon_i)$  and  $(Y_i)$ .

Series of the form (1) are known as LePage series. For fixed  $t \in [0, 1]$ , the convergence in  $\mathbb{R}^d$  of the series (1) is ensured as soon as one of the following conditions is satisfied:

 $\begin{array}{l} -\ 0 < \alpha < 1, \mathbb{E}|\varepsilon_1|^{\alpha} < \infty \text{ and } \mathbb{E}|Y_1(t)|^{\alpha} < \infty, \\ -\ 1 \le \alpha < 2, \mathbb{E}\varepsilon_1 = 0, \mathbb{E}|\varepsilon_1|^{\alpha} < \infty \text{ and } \mathbb{E}|Y_1(t)|^{\alpha} < \infty. \end{array}$ 

Here |.| denotes the usual Euclidean norm on  $\mathbb{R}$  or on  $\mathbb{R}^d$ . The random variable X(t) has then an  $\alpha$ -stable distribution on  $\mathbb{R}^d$ . Conversely, it is well known that any  $\alpha$ -stable distributions on  $\mathbb{R}^d$  admits a representation in terms of LePage series (see for example Samorodnitsky and Taqqu (1994), Section 3.9).

There is a vast literature on symmetric  $\alpha$ -stable distributions on separable Banach spaces (see e.g. Ledoux and Talagrand (1991), Araujo and Giné (1980)). In particular, any symmetric  $\alpha$ -stable distribution on a separable Banach space can be represented as an almost surely convergent LePage series (see Corollary 5.5 in Ledoux and Talagrand (1991)). The existence of a symmetric  $\alpha$ -stable distribution with a given spectral measure is not automatic and is linked with the notion of stable type of a Banach space; see Theorem 9.27 in Ledoux and Talagrand (1991) for a precise statement. Davydov et al. (2008) consider  $\alpha$ -stable distributions in the more general framework of abstract convex cones.

The space  $\mathbb{D}^d$  equipped with the norm

 $||x|| = \sup\{|x_i(t)|, t \in [0, 1], i = 1, ..., d\}, x = (x_1, ..., x_d) \in \mathbb{D}^d,$ 

is a Banach space but is not separable. The uniform topology associated with this norm is finer than the  $J_1$ -topology. On the other hand, the space  $\mathbb{D}^d$  with the  $J_1$ -topology is Polish, i.e. there exists a metric on  $\mathbb{D}^d$  compatible with the  $J_1$ -topology that makes  $\mathbb{D}^d$  a complete and separable metric space. However, such a metric cannot be compatible with the vector space structure since the addition is not continuous in the  $J_1$ -topology. These properties explain why the general theory of stable distributions on separable Banach space cannot be applied to the space  $\mathbb{D}^d$ .

Nevertheless, in the case where the series (1) converges, the distribution of the sum X defines an  $\alpha$ -stable distribution on  $\mathbb{D}^d$ . We can determine the associated spectral measure  $\sigma$  on the unit sphere  $\mathbb{S}^d = \{x \in \mathbb{D}^d; \|x\| = 1\}$ . It is given by

$$\sigma(A) = \frac{\mathbb{E}\Big(|\varepsilon_1|^{\alpha} \|Y_1\|^{\alpha} \mathbf{1}_{\{\operatorname{sign}(\varepsilon_1)Y_1/\|Y_1\| \in A\}}\Big)}{\mathbb{E}(|\varepsilon_1|^{\alpha} \|Y_1\|^{\alpha})}, \quad A \in \mathscr{B}(\mathbb{S}^d)$$

This is closely related to regular variations theory (see Hult and Lindskog (2006), Davis and Mikosch (2008)). For all r > 0 and  $A \in \mathcal{B}(\mathbb{S}^d)$  such that  $\sigma(\partial A) = 0$ , it holds that

$$\lim_{n\to\infty} n\mathbb{P}\left(\frac{X}{\|X\|} \in A \mid \|X\| > rb_n\right) = r^{-\alpha}\sigma(A),$$

with

 $b_n = \inf\{r > 0; \mathbb{P}(||X|| < r) \le n^{-1}\}, n \ge 1.$ 

The random variable X is said to be regularly varying in  $\mathbb{D}^d$  with index  $\alpha$  and spectral measure  $\sigma$ .

In this framework, convergence of the LePage series (1) in  $\mathbb{D}^d$  is known in some particular cases only:

- When  $0 < \alpha < 1$ ,  $\mathbb{E}|\varepsilon_1|^{\alpha} < \infty$  and  $\mathbb{E}||Y_1||^{\alpha} < \infty$ , the series (1) converges almost surely uniformly in [0, 1] (see example 4.2 in Davis and Mikosch (2008)).
- When  $1 \le \alpha < 2$ , the distribution of the  $\varepsilon_i$ 's is symmetric,  $\mathbb{E}|\varepsilon_1|^{\alpha} < \infty$  and  $Y_i(t) = \mathbf{1}_{[0,t]}(U)$  with  $(U_i)_{i\ge 1}$  an i.i.d. sequence of random variables with uniform distribution on [0, 1], the series (1) converges almost surely uniformly on [0, 1] and the limit process X is a symmetric  $\alpha$ -stable Lévy process (see Rosiński (2001)).

The purpose of this note is to complete these results and to provide a general criterion for almost sure convergence in  $\mathbb{D}^d$  of the random series (1). Our main result is the following:

**Theorem 1.** *Suppose that*  $1 \le \alpha < 2$ ,

$$\mathbb{E}\varepsilon_1 = 0, \qquad \mathbb{E}|\varepsilon_1|^{\alpha} < \infty \quad \text{and} \quad \mathbb{E}||Y_1||^{\alpha} < \infty.$$

Suppose furthermore that there exist  $\beta_1, \beta_2 > \frac{1}{2}$  and  $F_1, F_2$  nondecreasing continuous functions on [0, 1] such that, for all  $0 \le t_1 \le t \le t_2 \le 1$ ,

$$\mathbb{E}|Y_1(t_2) - Y_1(t_1)|^2 \le |F_1(t_2) - F_1(t_1)|^{\beta_1},\tag{2}$$

$$\mathbb{E}|Y_1(t_2) - Y_1(t)|^2 |Y_1(t) - Y_1(t_1)|^2 \le |F_2(t_2) - F_2(t_1)|^{2\beta_2}.$$
(3)

Then, the LePage series (1) converges almost surely in  $\mathbb{D}^d$ .

The proof of this theorem is detailed in the next section. We provide hereafter a few cases where Theorem 1 can be applied.

**Example 1.** The example considered by Davis and Mikosch (2008) follows easily from Theorem 1: let *U* be a random variable with uniform distribution on [0, 1] and consider  $Y_1(t) = \mathbf{1}_{[0,t]}(U)$ ,  $t \in [0, 1]$ . Then, for  $0 \le t_1 \le t \le t_2 \le 1$ ,

 $\mathbb{E}(Y_1(t_2) - Y_1(t_1))^2 = t_2 - t_1$  and  $\mathbb{E}(Y_1(t_2) - Y_1(t_1))^2 (Y_1(t_1) - Y_1(t_1))^2 = 0$ ,

so conditions (2) and (3) are satisfied.

**Example 2.** Example 1 can be generalized in the following way: let  $p \ge 1$ ,  $(U_i)_{1 \le i \le p}$  independent random variables on [0, 1] and  $(R_i)_{1 \le i \le p}$  random variables on  $\mathbb{R}^d$ . Consider

$$Y_1(t) = \sum_{i=1}^p R_i \mathbf{1}_{[0,t]}(U_i).$$

Assume that for each  $i \in \{1, ..., p\}$ , the cumulative distribution function  $F_i$  of  $U_i$  is continuous on [0, 1]. Assume furthermore that there is some M > 0 such that for all  $i \in \{1, ..., p\}$ 

$$\mathbb{E}[R_i^4 \mid \mathcal{F}_U] \leq M$$
 almost surely

(4)

where  $\mathcal{F}_U = \sigma(U_1, \ldots, U_p)$ . This is for example the case when the  $R_i$ 's are uniformly bounded by  $M^{1/4}$  or when the  $R_i$ 's have finite fourth moment and are independent of the  $U_i$ 's. Simple computations entail that under condition (4), it holds for all  $0 \le t_1 \le t \le t_2 \le 1$  that

$$\mathbb{E}(Y_1(t_2) - Y_1(t_1))^2 \le M^{1/2} p^2 |F(t_2) - F(t_1)|^2$$

and

$$\mathbb{E}(Y_1(t_2) - Y_1(t))^2 (Y_1(t) - Y_1(t_1))^2 \le Mp^4 |F(t_2) - F(t_1)|^4.$$

with  $F(t) = \sum_{i=1}^{p} F_i(t)$ . So conditions (2) and (3) are satisfied and Theorem 1 can be applied in this case.

**Example 3.** A further natural example is the case where  $Y_1(t)$  is a Poisson process with intensity  $\lambda > 0$  on [0, 1]. Then, for all  $0 \le t_1 \le t \le t_2 \le 1$ ,

$$\mathbb{E}(Y_1(t_2) - Y_1(t_1))^2 = \lambda |t_2 - t_1| + \lambda^2 |t_2 - t_1|^2$$

and

$$\mathbb{E}(Y_1(t_2) - Y_1(t))^2(Y_1(t) - Y_1(t_1))^2 = (\lambda |t_2 - t| + \lambda^2 |t_2 - t|^2)(\lambda |t - t_1| + \lambda^2 |t - t_1|^2)$$

and we easily see that conditions (2) and (3) are satisfied.

### 2. Proof

For the sake of clarity, we divide the proof of Theorem 1 into five steps.

Step 1. According to Lemma 1.5.1 in Samorodnitsky and Taqqu (1994), it holds almost surely that for k large enough,

$$|\Gamma_k^{-1/\alpha} - k^{-1/\alpha}| \le 2\alpha^{-1} k^{-1/\alpha} \sqrt{\frac{\ln \ln k}{k}}.$$
(5)

This implies the a.s. convergence of the series

$$\sum_{i=1}^{\infty} |\Gamma_i^{-1/\alpha} - i^{-1/\alpha}| |\varepsilon_i| \|Y_i\| < \infty.$$
(6)

The series (6) does indeed have nonnegative terms, and (5) implies that the following conditional expectation is finite:

$$\mathbb{E}\left[\sum_{i=1}^{\infty} |\Gamma_i^{-1/\alpha} - i^{-1/\alpha}| |\varepsilon_i| \|Y_i\| \middle| \mathcal{F}_{\Gamma}\right] = \mathbb{E}|\varepsilon_1| \mathbb{E}||Y_i|| \sum_{i=1}^{\infty} |\Gamma_i^{-1/\alpha} - i^{-1/\alpha}|$$

where  $\mathcal{F}_{\Gamma} = \sigma(\Gamma_i, i \geq 1)$ .

This proves that (6) holds true and it is enough to prove the a.s. convergence in  $\mathbb{D}^d$  of the series

$$Z(t) = \sum_{i=1}^{\infty} i^{-1/\alpha} \varepsilon_i Y_i(t), \quad t \in [0, 1],$$

$$\tag{7}$$

Step 2. Next, consider

$$\widetilde{Z}(t) = \sum_{i=1}^{\infty} i^{-1/\alpha} \widetilde{\varepsilon}_i Y_i(t), \quad t \in [0, 1].$$
(8)

with

$$\tilde{\varepsilon}_i = \varepsilon_i \mathbf{1}_{\{|\varepsilon_i|^{\alpha} \le i\}}, \quad i \ge 1.$$

We prove that the series (7) and (8) differ only by a finite number of terms. We have indeed

$$\sum_{i=1}^{\infty} \mathbb{P}\left(\tilde{\varepsilon}_{i} \neq \varepsilon_{i}\right) = \sum_{i=1}^{\infty} \mathbb{P}\left(|\varepsilon_{i}|^{\alpha} > i\right) \leq \mathbb{E}|\varepsilon_{1}|^{\alpha} < \infty$$

and the Borel–Cantelli lemma implies that almost surely  $\tilde{\varepsilon}_i = \varepsilon_i$  for *i* large enough. So, the two series (7) and (8) have the same nature and it is enough to prove the convergence in  $\mathbb{D}^d$  of the series (8).

Step 3. As a preliminary for step 4, we prove several estimates involving the moments of the random variables  $(\tilde{\varepsilon}_i)_{i\geq 1}$ . First, for all  $m > \alpha$ ,

$$C(\alpha, m) := \sum_{i=1}^{\infty} i^{-m/\alpha} \mathbb{E}(|\tilde{\varepsilon}_i|^m) < \infty.$$
(9)

We have indeed

$$C(\alpha, m) = \sum_{i=1}^{\infty} i^{-m/\alpha} \mathbb{E}(|\varepsilon_i|^m \mathbf{1}_{\{|\varepsilon_i| \le i^{1/\alpha}\}})$$
$$= \mathbb{E}\left(|\varepsilon_1|^m \sum_{i=1}^{\infty} i^{-m/\alpha} \mathbf{1}_{\{i \ge |\varepsilon_1|^\alpha\}}\right)$$
$$\le C\mathbb{E}(|\varepsilon_1|^m |\varepsilon_1|^{\alpha-m}) = C\mathbb{E}(|\varepsilon_1|^{\alpha}) < \infty$$

where the constant  $C = \sup_{x>0} x^{m/\alpha-1} \sum_{i \ge x} i^{-m/\alpha}$  is finite since for  $m > \alpha$ 

$$\lim_{x\to\infty} x^{m/\alpha-1} \sum_{i\geq x}^{\infty} i^{-m/\alpha} = \frac{\alpha}{m-\alpha}.$$

Similarly, we also have

$$C(\alpha, 1) := \sum_{i=1}^{\infty} i^{-1/\alpha} |\mathbb{E}(\tilde{\varepsilon}_i)| < \infty.$$
(10)

Indeed, the assumption  $\mathbb{E}\varepsilon_i = 0$  implies  $\mathbb{E}(\tilde{\varepsilon}_i) = \mathbb{E}(\varepsilon_i \mathbf{1}_{\{|\varepsilon_i|^{\alpha} > i\}})$ . Hence,

$$\begin{split} \sum_{i=1}^{\infty} i^{-1/\alpha} |\mathbb{E}(\tilde{\varepsilon}_i)| &\leq \sum_{i=1}^{\infty} i^{-1/\alpha} \mathbb{E}(|\varepsilon_1| \mathbf{1}_{\{|\varepsilon_1| > i^{1/\alpha}\}}) \\ &= \mathbb{E}\Big(|\varepsilon_1| \sum_{i=1}^{[|\varepsilon_1|^{\alpha}]} i^{-1/\alpha}\Big) \\ &\leq \mathbb{E}\Big(|\varepsilon_1| C'(|\varepsilon_1|^{\alpha})^{1-1/\alpha}\Big) = C' \mathbb{E}|\varepsilon_1|^{\alpha} < \infty \end{split}$$

where the constant  $C' = \sup_{x>0} x^{1/\alpha-1} \sum_{i=1}^{[x]} i^{-1/\alpha}$  is finite. Step 4. For  $n \ge 1$ , consider the partial sum

$$\widetilde{Z}_n(t) = \sum_{i=1}^n i^{-1/\alpha} \widetilde{\varepsilon}_i Y_i(t), \quad t \in [0, 1].$$
(11)

We prove that the sequence of processes  $(\widetilde{Z}_n)_{n\geq 1}$  is tight in  $\mathbb{D}^d$ . Following Theorem 3 in Gikhman and Skorokhod (2004), Chapter 6, Section 3, it is enough to show that there exists  $\beta > 1/2$  and a nondecreasing continuous function F on [0, 1] such that

$$\mathbb{E}|\widetilde{Z}_{n}(t_{2}) - \widetilde{Z}_{n}(t)|^{2}|\widetilde{Z}_{n}(t) - \widetilde{Z}_{n}(t_{1})|^{2} \leq |F(t_{2}) - F(t_{1})|^{2\beta},$$
(12)

for all  $0 \le t_1 \le t \le t_2 \le 1$ . Remark that in Gikhman and Skorokhod (2004), the result is stated only for  $F(t) \equiv t$ . However, the case of a general continuous nondecreasing function F follows easily from a simple change of variable.

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We use the notation  $Y(t) = (Y^p(t))_{1 \le p \le d}$ ,  $[[1, n]] = \{1, ..., n\}$  and  $\mathbf{i} = (i_1, i_2, i_3, i_4) \in [[1, n]]^4$ . We have

$$\mathbb{E}|\widetilde{Z}_{n}(t_{2}) - \widetilde{Z}_{n}(t)|^{2}|\widetilde{Z}_{n}(t) - \widetilde{Z}_{n}(t_{1})|^{2} = \mathbb{E}\left|\sum_{i=1}^{n} i^{-1/\alpha} \widetilde{\varepsilon}_{i}(Y_{i}(t) - Y_{i}(t_{1}))\right|^{2} \left|\sum_{j=1}^{n} j^{-1/\alpha} \widetilde{\varepsilon}_{j}(Y_{j}(t_{2}) - Y_{j}(t))\right|^{2} \\ = \sum_{1 \le p,q \le d} \sum_{\mathbf{i} \in [[1,n]]^{4}} (i_{1}i_{2}i_{3}i_{4})^{-1/\alpha} \mathbb{E}(\widetilde{\varepsilon}_{i_{1}}\widetilde{\varepsilon}_{i_{2}}\widetilde{\varepsilon}_{i_{3}}\widetilde{\varepsilon}_{i_{4}}) \mathbb{E}[(Y_{i_{1}}^{p}(t) - Y_{i_{1}}^{p}(t_{1}))$$
(13)

$$(Y_{i_{2}}^{p}(t) - Y_{i_{2}}^{p}(t_{1}))(Y_{i_{3}}^{q}(t_{2}) - Y_{i_{3}}^{q}(t))(Y_{i_{4}}^{q}(t_{2}) - Y_{i_{4}}^{q}(t))]$$

$$\leq d^{2} \sum_{i_{1}} (i_{1}i_{2}i_{3}i_{4})^{-1/\alpha} |\mathbb{E}(\tilde{\varepsilon}_{i_{1}}\tilde{\varepsilon}_{i_{2}}\tilde{\varepsilon}_{i_{3}}\tilde{\varepsilon}_{i_{4}})|D_{\mathbf{i}}(t, t_{1}, t_{2})$$
(14)
(15)

where

$$D_{\mathbf{i}}(t, t_1, t_2) = \mathbb{E}|Y_{i_1}(t) - Y_{i_1}(t_1)||Y_{i_2}(t) - Y_{i_2}(t_1)||Y_{i_3}(t_2) - Y_{i_3}(t)||Y_{i_4}(t_2) - Y_{i_4}(t)|.$$

Consider  $\sim_i$ , the equivalence relation on  $\{1, \ldots, 4\}$  defined by

$$j \sim_{\mathbf{i}} j'$$
 if and only if  $i_j = i_{j'}$ 

 $i \in [[1,n]]^4$ 

Let  $\mathcal{P}$  be the set of all partitions of  $\{1, \ldots, 4\}$  and  $\tau(\mathbf{i})$  be the partition of  $\{1, 2, 3, 4\}$  given by the equivalence classes of  $\sim_{\mathbf{i}}$ . We introduce these definitions because, since the  $Y_i$ 's are i.i.d., the term  $D_{\mathbf{i}}(t, t_1, t_2)$  depends on  $\mathbf{i}$  only through the associated partition  $\tau(\mathbf{i})$ . For example, if  $\tau(\mathbf{i}) = \{1, 2, 3, 4\}$ , i.e. if  $i_1 = i_2 = i_3 = i_4$ , then

$$D_{\mathbf{i}}(t, t_1, t_2) = \mathbb{E}|Y_1(t) - Y_1(t_1)|^2 |Y_1(t_2) - Y_1(t)|^2.$$

Or if  $\tau$  (i) = {1}  $\cup$  {2}  $\cup$  {3}  $\cup$  {4}, i.e. if the indices  $i_1, \ldots, i_4$  are pairwise distinct, then

$$D_{\mathbf{i}}(t, t_1, t_2) = (\mathbb{E}|Y_1(t) - Y_1(t_1)|\mathbb{E}|Y_1(t_2) - Y_1(t)|)^2$$

For  $\tau \in \mathcal{P}$ , we denote by  $D_{\tau}(t, t_1, t_2)$  the common value of the terms  $D_i(t, t_1, t_2)$  corresponding to indices i such that  $\tau(\mathbf{i}) = \tau$ . Define also

$$S_{n,\tau} = \sum_{\mathbf{i} \in \{1,\dots,n\}^4; \tau(\mathbf{i}) = \tau} (i_1 i_2 i_3 i_4)^{-1/\alpha} |\mathbb{E}(\tilde{\varepsilon}_{i_1} \tilde{\varepsilon}_{i_2} \tilde{\varepsilon}_{i_3} \tilde{\varepsilon}_{i_4})|$$

With this notation, Eq. (15) can be rewritten as

$$\mathbb{E}|\widetilde{Z}_n(t_2) - \widetilde{Z}_n(t)|^2 |\widetilde{Z}_n(t) - \widetilde{Z}_n(t_1)|^2 \le d^2 \sum_{\tau \in \mathcal{P}} S_{n,\tau} D_\tau(t, t_1, t_2).$$
(16)

Under conditions (2) and (3), we will prove that for each  $\tau \in \mathcal{P}$ , there exist  $\beta_{\tau} > 1/2$ , a nondecreasing continuous function  $F_{\tau}$  on [0, 1] and a constant  $S_{\tau} > 0$  such that

$$D_{\tau}(t, t_1, t_2) \le |F_{\tau}(t_1) - F_{\tau}(t_2)|^{2\beta_{\tau}}, \quad 0 \le t_1 \le t \le t_2,$$
(17)

and

$$S_{n,\tau} \le S_{\tau}, \quad n \ge 1. \tag{18}$$

Eqs. (16)–(18) together imply inequality (12) for some suitable choices of  $\beta > 1/2$  and F.

It remains to prove inequalities (17) and (18). If  $\tau = \{1, 2, 3, 4\}$ ,

$$D_{\tau}(t, t_1, t_2) \leq \mathbb{E}|Y_1(t) - Y_1(t_1)|^2 |Y_1(t_2) - Y_1(t)|^2 \leq |F_2(t_2) - F_2(t_1)|^{2\beta_2}$$

and

$$S_n^{\tau} = \sum_{i=1}^n i^{-4/\alpha} \mathbb{E} \tilde{\varepsilon}_i^4 \le C(\alpha, 4).$$

If  $\tau = \{1\} \cup \{2\} \cup \{3\} \cup \{4\}$ , the Cauchy–Schwartz inequality entails

$$D_{\tau}(t, t_1, t_2) \le (\mathbb{E}|Y_1(t) - Y_1(t_1)|\mathbb{E}|Y_1(t_2) - Y_1(t)|)^2 \le |F_1(t_2) - F_1(t_1)|^{2\beta_1}$$

and

$$S_n^{\tau} \leq \sum_{\mathbf{i} \in \{1,\dots,n\}^4; \tau(\mathbf{i}) = \tau} (i_1 i_2 i_3 i_4)^{-1/\alpha} |\mathbb{E} \tilde{\varepsilon}_{i_1}| |\mathbb{E} \tilde{\varepsilon}_{i_2}| |\mathbb{E} \tilde{\varepsilon}_{i_3}| |\mathbb{E} \tilde{\varepsilon}_{i_4}| \leq C(\alpha, 1)^4.$$

Similarly, for  $\tau = \{1, 2, 3\} \cup \{4\}$ ,

$$\begin{aligned} \mathsf{D}_{\tau}(t,t_{1},t_{2}) &= \mathbb{E}|Y_{1}(t)-Y_{1}(t_{1})|^{2}|Y_{1}(t_{2})-Y_{1}(t)|\mathbb{E}|Y_{1}(t_{2})-Y_{1}(t)| \\ &\leq |F_{1}(t)-F_{1}(t_{1})|^{\beta_{1}/2}|F_{2}(t_{2})-F_{2}(t_{1})|^{\beta_{2}}|F_{1}(t_{2})-F_{1}(t)|^{\beta_{1}/2} \\ &\leq |(F_{1}+F_{2})(t_{2})-(F_{1}+F_{2})(t_{1})|^{\beta_{1}+\beta_{2}} \end{aligned}$$

and

$$S_n^{\tau} \leq \sum_{1 \leq i \neq j \leq n} (i^3 j)^{-1/\alpha} \mathbb{E} |\tilde{\varepsilon}_i|^3 |\mathbb{E} \tilde{\varepsilon}_j| \leq C(\alpha, 3) C(\alpha, 3).$$

or for  $\tau = \{1, 2\} \cup \{3\} \cup \{4\}$ ,

$$\begin{aligned} D_{\tau}(t, t_1, t_2) &= \mathbb{E}|Y_1(t) - Y_1(t_1)|^2 (\mathbb{E}|Y_1(t_2) - Y_1(t)|)^2 \\ &\leq |F_1(t) - F_1(t_1)|^{\beta_1} |F_1(t_2) - F_1(t)|^{\beta_1} \\ &\leq |F_1(t_2) - F_1(t_1)|^{2\beta_1} \end{aligned}$$

and

$$S_n^{\tau} \leq \sum_{1 \leq i \neq j \neq k \leq n} (i^2 j k)^{-1/\alpha} \mathbb{E} |\tilde{\varepsilon}_i|^2 |\mathbb{E} \tilde{\varepsilon}_j \parallel \mathbb{E} \tilde{\varepsilon}_k| \leq C(\alpha, 2) C(\alpha, 1)^2.$$

Similar computations can be checked in all remaining cases. The cardinality of  $\mathcal{P}$  is equal to 13.

Step 5. We prove Theorem 1. For each fixed  $t \in [0, 1]$ , Kolmogorov's three-series theorem implies that  $\widetilde{Z}_n(t)$  converges almost surely as  $n \to \infty$ . So the finite-dimensional distributions of  $(\widetilde{Z}_n)_{n\geq 1}$  converge. The tightness in  $\mathbb{D}^d$  of the sequence has already been proved in step 4, so  $(\widetilde{Z}_n)_{n\geq 1}$  weakly converges in  $\mathbb{D}^d$  as  $n \to \infty$ . We then apply Theorem 1 of Kallenberg (1974) and deduce that  $\widetilde{Z}_n$  converges almost surely in  $\mathbb{D}^d$ . In view of step 1 and step 2, this yields the almost sure convergence of the series (1).

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