

ON SELF-SIMILAR STOCHASTIC INTEGRAL PROCESSES

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1. Introduction

A stochastic process $X = \{X(t) : t \in T\}$, with an index set T , is said to be infinitely divisible (*ID*) if its finite dimensional distributions are all *ID*. An *ID* process X is said to be a stochastic integral process if $X = \{X(t) : t \in T\} \stackrel{\mathcal{D}}{=} \{\int f_t d\Lambda : t \in T\}$ where $f : T \times S \rightarrow R$ is a deterministic function and Λ is an *ID* random measure on a δ -ring \mathcal{S} of subsets of an arbitrary non-empty set S with the property; there exists an increasing sequence $\{S_n\}$ of sets in \mathcal{S} with $\cup_n S_n = S$. Here $\stackrel{\mathcal{D}}{=}$ denotes equality in all finite dimensional distributions.

Let $M_\alpha (0 < \alpha \leq 2)$ be a symmetric α -stable random measure on measurable space (S, \mathcal{S}) with a control measure m and $I(f) = \int_S f(s) dM_\alpha$ for all measurable function $f : S \rightarrow R$ satisfies the condition

$$\int_S |f_t(s)|^\alpha m(ds) < \infty \text{ for each } t.$$

Then $\{\int_S f_t(s) dM_\alpha : t \in T\}$ is symmetric α -stable process. [4, 8] gives us a Lepage representation of symmetric α -stable process and a characterization of self-similar α -stable process.

We consider integrals with respect to *ID* random measure which generalizes the stable random measure and are mostly interested in characteristic functions, series representation and self-similarities of stochastic integral process $\{X(t) : t \in T\} \stackrel{\mathcal{D}}{=} \{\int f_t(s) d\Lambda : t \in T\}$.

First, series representations involving arrival times in a Poisson process have been given by [2], for real independent increment processes without Gaussian components and with positive jumps. Series representation derived as a special case of a generalized shot noise generalizes

various Lepage type representations of stochastic integral processes. (See [3, 5]).

Chapter 2 is to review technical back-ground such as identification of the space of \wedge -integrable functions as a certain Musielak-Orlicz spaces L_{Φ_p} and characteristic functions of \wedge -integrable functions in terms of certain parameters of an ID random measure.

Chapter 3 gives us a characterizations and necessary conditions of self-similar stochastic integral processes $\{\int f_t d\wedge : t \in T\}$ where $f_t \in L_{\Phi_p}$.

DEFINITION 1.1. A random measure \wedge is said to be an independently scattered ID random measure (for short, ID random measure) if for each $A \in \mathcal{S}$, $\wedge(A)$ is ID random variable, and \wedge is independently scattered random measure.

DEFINITION 1.2. A positive Borel measure on $R \setminus \{0\}$ is Lévy measure if it integrates the function $\min\{1, x^2\}$.

DEFINITION 1.3. A σ -finite measure ν on $\sigma(\mathcal{S})$ (= the smallest σ -field generated by \mathcal{S}) is said to be a control measure of the random measure \wedge if \wedge and ν have the same families of zero sets.

DEFINITION 1.4. A measurable function $f : (S, \sigma(\mathcal{S})) \rightarrow (R, \mathcal{B}(R))$ is said to be \wedge -integrable if there exists a sequence $\{f_n\}$ of simple function such that

- (1) $f_n \rightarrow f$ ν -a.s.
- (2) for every $A \in \sigma(\mathcal{S})$, the sequence $\{\int_A f_n d\wedge\}$ converges in probability, as $n \rightarrow \infty$.

If f is \wedge -integrable, then we put

$$\int_A f d\wedge = p - \lim_{n \rightarrow \infty} \int_A f_n d\wedge,$$

where $\{f_n\}$ satisfies (1) and (2).

DEFINITION 1.5. A stochastic process $\{X(t); t \in T\}$ is said to be self-similar with index $H(H - ss)$ if for some $H \in R$.

$$X(ct) \stackrel{\mathcal{D}}{=} c^H X(t) \text{ for any } c > 0.$$

2. \wedge -integrable functions and Musielak-Orlicz space

We assume $\Lambda(A)$ is an *ID* random variable for $A \in \mathcal{S}$. Then its characteristic function can be written in the Lévy-Khintchin form

$$(2.1) \quad \mathcal{L}(\Lambda(A))^\wedge(u) = \exp\left\{iuv_0(A) - \frac{1}{2}u^2\nu_1(A) + \int_R \{e^{iux} - 1 - iuxI_{(|x|\leq 1)}\}F_A(dx)\right\},$$

where $-\infty < \nu_0(A) < \infty, 0 \leq \nu_1(A) < \infty, F_A$ is a Lévy measure on R and I_B is a indicator function of B .

We know that there is one-to-one correspondence between the class of *ID* random measures $\Lambda(\cdot)$ on one hand and the class of parameters $\nu_0(\cdot), \nu_1(\cdot)$ and $F_{(\cdot)}$ on the other.

The following lemma introduces an explicit form of a control measure for a general *ID* random measure.

LEMMA 2.1. *Let ν_0, ν_1 and F . be as in (2.1) and define*

$$\nu(A) = |\nu_0|(A) + \nu_1(A) + \int_R \min\{1, x^2\} F_A(dx) , A \in \mathcal{S}$$

Then $\nu : \sigma(\mathcal{S}) \rightarrow [0, \infty]$ is a control measure of Λ .

Proof. Let $A_n \in \mathcal{S}, A_n \downarrow \phi$. Since $\Lambda(A_n) \rightarrow 0$ in probability, we have that

$$\nu_0(A_n) \rightarrow 0, \nu_1(A_n) \rightarrow 0 \text{ and } \int_R \min\{1, x^2\} F_{A_n}(dx) \rightarrow 0.$$

We get $\nu(A_n) \rightarrow 0$ proving that ν is countably additive. Since $\nu(S_n) < \infty$, where $S_n \in \mathcal{S}$ is increasing and $\cup_n S_n = \mathcal{S}$, we may uniquely extend ν to a σ -finite measure on $(\mathcal{S}, \sigma(\mathcal{S}))$.

To prove that the σ -finite measure ν is a control measure of Λ , let $A \in \sigma(\mathcal{S})$ be a \wedge -zero set. Then, for any $A_1 \in \mathcal{S}$ satisfying $A_1 \subset A, \Lambda(A_1) = 0$ a.s. Decompose $A_1 = A'_1 \cup A''_1$ such that $\nu_0(A'_1) = \nu_0^+(A_1)$ and $\nu_0(A''_1) = -\nu_0^-(A_1)$. Since $\Lambda(A'_1) = \Lambda(A''_1) = 0$, we get that $\nu_0(A'_1) = \nu_0(A''_1) = 0$. Hence $|\nu_0|(A_1) = 0$.

We know that $\nu_1(A_1) = 0$ and $\int_R \min\{1, x^2\} F_{A_1}(dx) = 0$. Thus $\nu(A) = 0$ by an approximation theorem. ([1, Theorem 11.4]).

Conversely, let $A \in \sigma(\mathcal{S})$ be a ν -zero set. Then $\nu(A_1) = 0$ for any $A_1 \subset A, A_1 \in \mathcal{S}$. Therefore, $|\nu_0(A_1)| \leq |\nu_0|(A_1) = 0, \nu_1(A_1) = 0$ and $\int_R \min\{1, x^2\} F_{A_1}(dx) = 0$, i.e., $\wedge(A_1) = 0$ a.s. Let λ be an arbitrary but fixed control measure of \wedge .

LEMMA 2.2. *Let F be as in (2.1). Then there exists a unique σ -finite measure F on $\sigma(\mathcal{S}) \times \mathcal{B}(R)$ such that*

$$(2.2) \quad F(A \times B) = F_A(B), \text{ for all } A \in \mathcal{S}, B \in \mathcal{B}(R)$$

Moreover, there exists a function $Q : S \times \mathcal{B}(R) \rightarrow [0, \infty]$ such that

- (1) $Q(s, \cdot)$ is a Borel measure on $\mathcal{B}(R)$, for every $s \in S$.
- (2) $Q(\cdot, B)$ is a $\sigma(\mathcal{S})$ - measurable function, for every $B \in \mathcal{B}(R)$.
- (3) $\int_{S \times R} h(s, x) F(ds, dx) = \int_S [\int_R h(s, x) Q(s, dx)] \lambda(ds)$
for every $\sigma(\mathcal{S}) \times \mathcal{B}(R)$ -measurable function $h : S \times R \rightarrow [0, \infty]$.
- (4) $\int_R \min\{1, x^2\} Q(s, dx) < \infty$, for every $s \in S$,
- (5) $\lambda\{s \in S : a(s) = \sigma^2(s) = Q(s, R) = 0\} = 0$.
- (6) $\mathcal{L}(\wedge(A))^\wedge(u) = \exp\{\int_A K(u, s) \lambda(ds)\}$ where $K(u, s) = iua(s) - \frac{1}{2}u^2\sigma^2(s) + \int_R \{e^{iux} - 1 - iuxI_{(|x| \leq 1)}\} Q(s, dx)$,
 $a(s) = \frac{d\nu_0}{d\lambda}(s), \sigma^2(s) = \frac{d\nu_1}{d\lambda}(s)$

Proof. By [7, Lemma 2.3], there exists a unique σ -finite measure F on $\sigma(\mathcal{S}) \times \mathcal{B}(R)$ satisfying (2.2), and we can find a function $\rho : S \times \mathcal{B}(R) \rightarrow [0, \infty]$ such that

- (a) $\rho(s, \cdot)$ is a Lévy measure on $\mathcal{B}(R)$, for every $s \in S$,
- (b) $\rho(\cdot, B)$ is a $\sigma(\mathcal{S})$ - measurable function, for every $B \in \mathcal{B}(R)$.
- (c) $\int_{S \times R} h(s, x) F(ds, dx) = \int_S [\int_R h(s, x) \rho(s, dx)] \lambda(ds)$
for every $\sigma(\mathcal{S}) \times \mathcal{B}(R)$ -measurable function $h : S \times R \rightarrow [0, \infty]$.

Since λ and ν are equivalent σ -finite measure on $\sigma(\mathcal{S})$ there exists a strictly positive and finite version of ψ of Radon-Nikodym derivative $\frac{d\nu}{d\lambda}$. Put

$$Q(s, dx) = \psi(s)\rho(s, dx).$$

Then (1), (2), (3) and (6) follow because

$$F_A(B) = F(A \times B) = \int_A \int_R I_B(x) Q(s, dx) \lambda(ds).$$

Since $\rho(s, \cdot)$ is a Lévy measure, (4) is satisfied.

Finally, note that $A_0 = \{s : a(s) = \sigma^2(s) = Q(s, R) = 0\}$ is a Λ -zero set, so that $\lambda(A_0) = 0$.

The following provides a necessary and sufficient condition for the existence of $\int_S f d\Lambda$ in terms of the deterministic characteristic of Λ .

THEOREM 2.3. *Let $f : S \rightarrow R$ be a $\sigma(S)$ -measurable function. Then f is Λ -integrable if and only if the following three conditions hold;*

- (1) $\int_S |U(f(s), s)| \lambda(ds) < \infty$,
- (2) $\int_S |f(s)|^2 \sigma^2(s) \lambda(ds) < \infty$ and
- (3) $\int_S V_0(f(s), s) \lambda(ds) < \infty$,

where

$$U(u, s) = ua(s) + \int_R \{xuI_{(|xu| \leq 1)} - uxI_{(|x| \leq 1)}\} Q(s, dx),$$

$$V_0(u, s) = \int_R \min\{1, |ux|^2\} Q(s, dx)$$

Proof. See [7, Theorem 2.7]

We shall define a certain Musielak-Orlicz space and identify the set of Λ - integrable functions as Musielak-Orlicz space.

Let q be a nonnegative number such that

$$E|\wedge(A)|^q < \infty \text{ for all } A \in S.$$

For $0 \leq p \leq q$,

Define

$$(2.3) \quad \Phi_p(u, s) = U^*(u, s) + u^2 \sigma^2(s) + V_p(u, s)$$

where

$$U^*(u, s) = \sup_{|c| \leq 1} (|U(cu, s)|)$$

$$V_p(u, s) = \int_R \{|ux|^p I_{(|ux| > 1)} + |ux|^2 I_{(|ux| \leq 1)}\} Q(s, dx)$$

Then the following three conditions are satisfied

- (1) For every $s \in S$, $\Phi_p(\cdot, s)$ is a continuous non-decreasing function on $[0, \infty)$ with $\Phi_p(0, s) = 0$
- (2) $\lambda\{s : \Phi_p(u, s) = 0 \text{ for some } u \neq 0\} = 0$
- (3) There exists a numerical constant $C > 0$ such that $\Phi_p(2u, s) \leq C\Phi_p(u, s)$, for all $u \geq 0$ and $s \in S$.

Now, we can define the so-called Musielak-Orlicz space

$$L_{\Phi_p}(S; \lambda) = \{f \in L_0(S; \lambda) : \int_S \Phi_p(|f(s)|, s)\lambda(ds) < \infty\}.$$

The space $L_{\Phi_p}(S; \lambda)$ is a complete linear metric space with norm defined by

$$\|f\|_{\Phi_p} = \inf\{c > 0; \int_S \Phi_p(c^{-1}|f(s)|, s)\lambda(ds) \leq c\}.$$

If the function Φ_p is independent of S , i.e. $\Phi_p(u, s) = \Phi_p(u)$, then the corresponding space $L_{\Phi_p}(S; \lambda)$ is called an Orlicz space.

THEOREM 2.4. *Let $0 \leq p \leq q$ and Φ_p be as in (2.3). Then*

$$\{f : f \text{ is } \wedge - \text{integrable and } E|\int f d\lambda|^p < \infty = L_{\Phi_p}(S; \lambda)$$

$p = 0$ signifies that $\{f : f \text{ is } \wedge - \text{integrable}\} = L_{\Phi_0}(S; \lambda)$.

Proof. See [7, Theorem 3.3].

Let $\lambda^{(1)}$ be a arbitrary probability measure on $(S, \sigma(S))$ equivalent to λ .

Set

$$R(r, s) = \inf\{x > 0; Q(s, [-x, x]^c) \leq r\}, r > 0$$

$$R^{(1)}(r, s) = R(r \frac{d\lambda^{(1)}}{d\lambda}(s), s), r > 0, s \in S$$

where the version of the Radon-Nikodym derivative $\frac{d\lambda^{(1)}}{d\lambda}$ is chosen to be strictly positive and finite everywhere.

Let f be a \wedge -integrable function i.e., $f \in L_{\Phi_0}(S; \lambda)$. Define

$$(2.4) \quad F_f(A) = \int_S \int_R I_{A \setminus \{0\}}(xf(s))Q(s, dx)\lambda(ds)$$

Define $a_f = \int_S U(f(s), s)\lambda(ds)$, $\sigma_f^2 = \int_S |f(s)|^2 \sigma^2(s)\lambda(ds)$.

THEOREM 2.5. Assume $f \in L_{\Phi_0}(S; \lambda)$. Then F_f is Lévy measure and

$$\begin{aligned} & \mathcal{L}\left(\int f d\Lambda\right)^\wedge(u) \\ &= \exp\left\{iua_f - \frac{1}{2}u^2\sigma f^2 + \int_R \{e^{iux} - 1 - iuxI_{(|x|\leq 1)}\}F_f(dx)\right\} \\ &= \exp\left\{iu \int_S a(s)f(s)\lambda(ds) - \frac{1}{2} \int_S |f(s)|^2\sigma^2(s)\lambda(ds) \right. \\ & \quad \left. + \int_R \{e^{iuR^{(1)}(r,s)f(s)} - 1 - iuR^{(1)}(r,s)f(s)I_{(R^{(1)}(r,s)\leq 1)}\}\lambda^{(1)}(ds)dr\right\}. \end{aligned}$$

Proof. F_f is Lévy measure, since we have

$$\begin{aligned} \int_{\{|x|\leq 1\}} x^2 F_f(dx) &= \int_S \int_{\{|f(s)x|\leq 1\}} |f(s)x|^2 Q(s, dx)\lambda(ds) \\ &\leq \int_S \Phi_0(|f(s)|, s)\lambda(ds) \\ &< \infty, \end{aligned}$$

$$\begin{aligned} \int_{\{|x|>1\}} F_f(dx) &= \int_S \int_{\{|f(s)x|>1\}} Q(s, dx)\lambda(ds) \\ &\leq \int_S \Phi_0(|f(s)|, s)\lambda(ds) \\ &< \infty. \end{aligned}$$

Since, for every $x \geq 0$ and $s \in S$

$$Leb\{r > 0; R^{(1)}(r, s) > x\} = \frac{d\lambda}{d\lambda^{(1)}}(s)Q(s, [-x, x]^c)$$

We get,

$$(2.5) \quad F_f(A) = \int_0^\infty \int_S I_{A \setminus \{0\}}(R^{(1)}(r, s)f(s))\lambda^{(1)}(ds)dr.$$

By [7, Theorem 2.7] we complete the proof.

3. Self-similarity of stochastic integral processes

Stochastic integral processes $\{\int f_t d\Lambda; t \in T\}$ are always assumed to be real-valued and defined for $t \in T = [0, \infty)$. All stochastic processes will be discussed in terms of finite-dimensional distributions.

By $\{\int f_t d\Lambda; t \in T\} \stackrel{\mathcal{D}}{=} \{\int g_t d\Lambda; t \in T\}$ we mean the equality of all finite-dimensional distributions. $\int f_t d\Lambda \stackrel{\mathcal{L}}{=} \int g_t d\Lambda$ means the equality of one-dimensional distributions for fixed t . Recall any stochastic process $\{\int f_t d\Lambda; t \in T\}$ is called H -ss,si(stationary increment) if, for every constant $c > 0, H \in R$,

$$\left\{ \int f_{ct} d\Lambda; t \in T \right\} \stackrel{\mathcal{D}}{=} \left\{ c^H \int f_t d\Lambda; t \in T \right\}$$

$$\left\{ \int f_{t+h} d\Lambda - \int f_h d\Lambda; t \in T \right\} \stackrel{\mathcal{D}}{=} \left\{ \int f_t d\Lambda - \int f_0 d\Lambda; t \in T \right\} \text{ for any } h > 0.$$

Let $\Lambda = \{\Lambda(A); A \in \mathcal{S}\}$ be a symmetric ID random measure without Gaussian component. Then the characteristic function of $\int f d\Lambda$ can be written in Lévy's form

$$\mathcal{L}\left(\int f d\Lambda\right)^\wedge(u) = \exp\left\{2 \int_0^\infty \{\cos ux - 1\} F_f(dx)\right\}, u \in R$$

where, $f \in L_{\Phi_0}(S; \lambda)$ and F_f is a symmetric Lévy measure.

THEOREM 3.1. Assume that $f_t \in L_{\Phi_0}(S; \lambda)$ for each $t \in T$ and $\{\int f_t d\Lambda; t \in T\}$ is H -ss,si stochastic integral process,

- (1) if $H < 0$, then $\|f_t\|_{\Phi_0} = 0$.
- (2) if $f_1 \in L_{\Phi_1}$ and $H \neq 1$, then $f_t \in L_{\Phi_1}$ and $E \int f_t d\Lambda = 0$.

Proof. (1) By H -ss, $\int f_0 d\Lambda = 0$ a.s. and

$$\int f_t d\Lambda \stackrel{\mathcal{L}}{=} t^H \int f_1 d\Lambda \rightarrow 0 \text{ as } t \rightarrow \infty.$$

On the other hand, by si,

$$\int f_t d\Lambda = \int f_t d\Lambda - \int f_0 d\Lambda \stackrel{\mathcal{L}}{=} \int f_{t+h} d\Lambda - \int f_h d\Lambda$$

which tends to 0 as $h \rightarrow \infty$. Hence $\int f_t d\Lambda = 0$ a.s. From the facts that $\int f_t d\Lambda = 0$ a.s. and $f_t \in L_{\phi_0}$, there exists a sequence $\{f_{t,n}\}$ of simple functions such that $f_{t,n} \rightarrow f_t$ λ -a.s. and $\int f_{t,n} d\Lambda \rightarrow 0$ in probability as $n \rightarrow \infty$. Since Λ is symmetric, we have $a(s) = 0$ and $Q(s, \cdot)$ is symmetric. Thus $U(\cdot, s) \equiv 0$ λ -a.s. which implies

$$\begin{aligned} \int \Phi_0(|f_{t,n}(s)|, s)\lambda(ds) &= \int V_0(|f_{t,n}(s)|, s)\lambda(ds) \\ &= \int \min\{1, x^2\}F_{f_{t,n}}(dx) \rightarrow 0 \end{aligned}$$

i.e.,

$$\|f_{t,n}\|_{\phi_0} \rightarrow 0, \text{ as } n \rightarrow \infty.$$

From the triangle inequality, we get $\|f_t\|_{\phi_0} = 0$.

Proof. (2) By $\int f_t d\Lambda \stackrel{d}{=} t^H \int f_1 d\Lambda$ and uniqueness of Lévy measure corresponding to same distribution functions, we have

$$\begin{aligned} \int \Phi_1(|f_t(s)|, s)\lambda(ds) &= \int \{|x|I_{(|x|>1)} + |x|^2 I_{(|x|\leq 1)}\}F_{f_t}(dx) \\ &= \int \{|x|I_{(|x|>1)} + |x|^2 I_{(|x|\leq 1)}\}F_{t^H f_1}(dx) \\ &= \int \Phi_1(|t^H f_1(s)|, s)\lambda(ds) < \infty, \end{aligned}$$

i.e., $f_t \in L_{\phi_1}$, which implies $E|\int f_t d\Lambda| < \infty$.

By H-ss,si, it follows that

$$\begin{aligned} E\left[\int f_t d\Lambda\right] &= E\left[\int f_{2t} d\Lambda - \int f_t d\Lambda\right] \\ &= (2^H - 1)E\left[\int f_t d\Lambda\right] \end{aligned}$$

so $(2^H - 2)E[\int f_t d\Lambda] = 0$. Since $H \neq 1$, we have $E[\int f_t d\Lambda] = 0$.

The following theorems gives us necessary conditions of H-ss stochastic integral process $\{\int f_t d\Lambda; t \in T\}$.

Let $\lambda^{(1)}$ be an arbitrary probability measure on $(S, \sigma(S))$ equivalent to λ .

THEOREM 3.2. Assume that $f_1 \in L_{\Phi_0}(S; \lambda)$ and $f_t = t^H f_1$ λ -a.s. for any $t \in T$. Then $\{\int f_t d\Lambda; t \in T\}$ is H -ss stochastic integral process.

Proof. For any $t \in [0, \infty]$, obviously $f_t \in L_{\Phi_0}$ and $f_{ct} = c^H f_t$ $\lambda^{(1)}$ -a.s. We have

$$\begin{aligned} \mathcal{L}(\int f_{ct} d\Lambda)^\wedge(u) &= \exp\{2 \int_0^\infty \{\cos ux - 1\} F_{f_{ct}}(dx)\} \\ &= \exp\{2 \int_0^\infty \int_S \{\cos(uR^{(1)}(r, s)f_{ct}(s)) - 1\} \lambda^{(1)}(ds) dr\} \\ &= \exp\{2 \int_0^\infty \int_S \{\cos(uR^{(1)}(r, s)c^H f_t(s)) - 1\} \lambda^{(1)}(ds) dr\} \\ &= \exp\{2 \int_0^\infty \{\cos ux - 1\} F_{c^H f_t}(dx)\} \end{aligned}$$

and $\sum a_j f_{ct_j} = c^H \sum_{j=1}^n a_j f_{t_j}$ $\lambda^{(1)}$ -a.s for any $a_j \in R, t_j \in T$. Thus,

$$\sum_{j=1}^n a_j \int f_{ct_j} d\Lambda \stackrel{d}{=} c^H \sum_{j=1}^n a_j \int f_{t_j} d\Lambda.$$

Let $\{\xi_j\}, \{\Gamma_j\}$, and $\{\varepsilon_j\}$ be independent sequence of random variables such that $\{\xi_j\}$ is a sequence of i.i.d. random variables in $(S, \sigma(S))$ with $\mathcal{L}(\xi_j) = \lambda^{(1)}, \{\Gamma_j\}$ is j^{th} arrival time of a Poisson process N_t with parameter 1 i.e. $\Gamma_j = \inf\{t > 0; N_t = j\}$. and $\{\varepsilon_n\}$ is a sequence of i.i.d. random variable with $P\{\varepsilon_n = -1\} = P\{\varepsilon_n = 1\} = \frac{1}{2}$. Put

$$X_n(t) = \sum_{j=1}^n \varepsilon_j R^{(1)}(\Gamma_j, \xi_j) f_t(\xi_j).$$

Under the same assumption as Theorem 3.2 , we know that $f_t \in L_{\Phi_0}$. By [5, Proposition 2], $X_n(t)$ converges a.s. (to $X(t)$) and $\{X(t); t \in T\} \stackrel{D}{=} \{\int f_t d\Lambda; t \in T\}$. Therefore, the following theorem holds.

THEOREM 3.3. Assume that $f_1 \in L_{\Phi_0}(S; \lambda)$ and $f_t = t^H f_1$ λ -a.s. for any $t \in T$ Then

- (1) $X_n(t)$ converges a.s (to $X(t)$),
- (2) $X(t) \stackrel{D}{=} \int f_t d\Lambda$,
- (3) $\{X(t) : t \in T\}$ is H -ss stochastic process.

From now on, we will assume that $\Lambda = \{\wedge(A) : A \in \mathcal{S}\}$ is a *ID* random measure without Gaussian component and have the characteristic function as the following form

$$\mathcal{L}\left(\int f d\wedge\right)^\wedge(u) = \exp\left\{\int_R \{e^{iux} - 1 - iux\} F_f(dx)\right\},$$

where $f \in L_{\phi_p}, p \geq 1$, and F_f is defined in (2.4).

Put

$$Y_n(t) = \sum_{j=1}^n R^{(1)}(\Gamma_j, \xi_j) f_t(\xi_j) - C_{f_t}(\Gamma_n)$$

where $f_t \in L_{\phi_p}, p \geq 1$ and

$$C_{f_t}(a) = \int_0^a \int_S R^{(1)}(r, s) f_t(s) \lambda^{(1)}(ds) dr, a > 0.$$

THEOREM 3.4. Assume that $f_1 \in L_{\phi_p}(S; \lambda), p \geq 1$, and $f_t = t^H f_1 \lambda$ -a.s. Then

- (1) $Y_n(t)$ converges a.s. (to $Y(t)$),
- (2) $Y(t) \stackrel{\mathcal{D}}{=} \int f_t d\wedge$,
- (3) $\{Y(t); t \in T\}$ is *H*-ss stochastic process.

Proof. Let $f \in L_{\phi_p}, p \geq 1$. First note that $C_f(a)$ is well-defined. Indeed,

$$\begin{aligned} & \int_0^a \int_S R^{(1)}(r, s) f(s) \lambda^{(1)}(ds) dr \\ & \leq a + \int_0^a \int_S |R^{(1)}(r, s) f(s)| I_{(|R^{(1)}(r, s) f(s)| > 1)} \lambda^{(1)}(ds) dr \\ & = a + \int_{\{|x| > 1\}} (|x|)^p F_f(dx) \\ & = a + \int_S \int_{\{|f(s)x| > 1\}} (|f(s)x|)^p Q(s, dx) \lambda(ds) \\ & = a + \int_S \Phi_p(|f(s)|, s) \lambda(ds) \\ & < \infty. \end{aligned}$$

We showed in the proof of Theorem 2.4 that F_f is a Lévy measure.
 Put

$$H(r, s) = R^{(1)}(r, s)f(s) \text{ in [6, Theorem 3.1]}$$

Then we get (1) and

$$\begin{aligned} & \mathcal{L}\left(\sum_{j=1}^{\infty} R^{(1)}(\Gamma_j, \xi_j)f(\xi_j) - \lim_{n \rightarrow \infty} C_f(\Gamma_n)\right) \\ &= \mathcal{L}\left(\int f d\Lambda\right) \\ &= \exp\left\{\int \{e^{iuR^{(1)}(r,s)f(s)} - 1 - iuR^{(1)}(r,s)f(s)\}\lambda^{(1)}(ds)dr\right\}. \end{aligned}$$

Let $f = \sum_{i=1}^m a_i f_{t_i}$. Then we get (2), since we have

$$\begin{aligned} \sum_{i=1}^m a_i \int f_{t_i} d\Lambda &= \int f d\Lambda \\ &= \sum_{j=1}^{\infty} R^{(1)}(\Gamma_j, \xi_j)f(\xi_j) - \lim_{n \rightarrow \infty} C_f(\Gamma_n) \\ &= \sum_{j=1}^{\infty} R^{(1)}(\Gamma_j, \xi_j)\left\{\sum_{i=1}^m a_i f_{t_i}(\xi_j)\right\} - \lim_{n \rightarrow \infty} C_f(\Gamma_n) \\ &= \sum_{i=1}^m a_i \left\{\sum_{j=1}^{\infty} R^{(1)}(\Gamma_j, \xi_j)f_{t_i}(\xi_j) - \lim_{n \rightarrow \infty} C_{f_{t_i}}(\Gamma_n)\right\} \\ &= \sum_{i=1}^m a_i Y(t_i). \end{aligned}$$

By the same argument as in Theorem 3.2,

$$\{Y(t); t \in T\} \stackrel{\mathcal{D}}{=} \left\{\int f_t d\Lambda; t \in T\right\}$$

is H-ss stochastic integral process.

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