65. A Constructive Approach to the Law Equivalence of Infinitely Divisible Random Measures

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We are concerned with our method to construct infinitely divisible random measures on T based on Poisson random measures on $S = T \times (R \setminus \{0\})$. As an application we discuss the equivalence problem for infinitely divisible random measures on T.

§1. Preliminaries. Let T be an arbitrary nonempty set and \mathcal{T} be a δ -ring of subsets of T. We assume there exists an increasing sequence $\{T_n; n \geq 1\} \subset \mathcal{T}$ with $T = \bigcup_{n=1}^{\infty} T_n$ and $\{t\} \in \mathcal{T}$ for each $t \in T$. Let $\Lambda = \{\Lambda(A); A \in \mathcal{T}\}$ be an infinitely divisible random measure (or ID random measure) on T with no Gaussian component, which is defined on a basic probability space (Ω, \mathcal{F}, P) (see [3]). In other words, Λ is a real stochastic process characterized by

(1.1)
$$E[\exp(iz\Lambda(A))] = \exp\left[izv(A) + \int \int_{A\times \mathbf{R}_0} g(z, x) M(dtdx)\right]$$

$$(z \in \mathbf{R}, A \in \mathcal{I}).$$

where $g(z, x) = \exp(izx) - 1 - izx\mathbf{1}_{J}(x)$, J = (-1,1) and $\mathbf{R}_{0} = \mathbf{R} \setminus \{0\}$. Here v is an \mathbf{R} -valued signed measure on \mathcal{T} and \mathbf{M} is a measure on $\mathbf{S} = T \times \mathbf{R}_{0}$ satisfying

(1.2)
$$\int \int_{A \times \mathbf{R}_0} (1 \wedge x^2) M(dt dx) < \infty \quad (A \in \mathcal{I}).$$

We mean by $\mathbf{\Lambda} = {}^{d}[v, M]$ that the probability law of $\mathbf{\Lambda}$ is determined by parameters v and M. We denote by \mathbf{P}_{Λ} the probability measure on a measurable space $(\mathbf{R}^{\mathcal{T}}, \mathcal{F}(\mathbf{R}^{\mathcal{T}}))$ induced by the map $\Lambda: \Omega \ni \omega \to \Lambda(\cdot, \omega) \in \mathbf{R}^{\mathcal{T}}$, where $\mathbf{R}^{\mathcal{T}}$ is the set of all \mathbf{R} -valued functions on \mathcal{T} and $\mathcal{F}(\mathbf{R}^{\mathcal{T}})$ is the σ -algebra on $\mathbf{R}^{\mathcal{T}}$ generated by all coordinate functions. The product measurable space $(\mathbf{S}, \mathcal{S})$ is given by $\mathcal{S} = \sigma(\mathcal{T}) \otimes \mathcal{B}(\mathbf{R}_0)$, where $\sigma(\mathcal{T})$ is the σ -algebra on \mathbf{T} generated by \mathcal{T} and $\mathcal{B}(\mathbf{R}_0)$ is the Borel σ -algebra on \mathbf{R}_0 . Let $\mathcal{N} = \mathcal{N}(\mathbf{S})$ be the totality of nonnegative (possibly infinite) integer-valued measures on $(\mathbf{S}, \mathcal{S})$. Let $\mathcal{F}^+(\mathbf{S})$ be the set of all nonnegative measurable functions on $(\mathbf{S}, \mathcal{S})$. We denote by $\mathcal{B}(\mathcal{N})$ the σ -algebra on \mathcal{N} generated by all functions f^* on \mathcal{N} given by

$$f^*(\nu) = \langle \nu, f \rangle = \int_S f d\nu \text{ for } f \in \mathcal{F}^+(S) \text{ and } \nu \in \mathcal{N}.$$

An \mathcal{N} -valued random element ξ is called a *Poisson random measure* on S with intensity M if it is defined on (Ω, \mathcal{F}, P) and its Laplace transform is given by

(1.3)
$$\mathbf{E}[\exp(-\langle \xi, f \rangle)] = \exp\left[-\int \int_{S} \{1 - \exp(-f(t, x))\} M(dt dx)\right]$$
 for $f \in \mathcal{F}^{+}(S)$.

§2. A construction of infinitely divisible random measures. In this section we shall construct a version of $\Lambda = {}^{d}[v, M]$ based on a Poisson random measure on S. For simplicity we may assume M(S) > 0.

Case (I): $M(S) < \infty$. For each $k \ge 1$, let $(S^k, \mathcal{S}^k, P_k)$ be a probability space given by $P_k = M(S)^{-k}M^k$, where we mean by $(S^k, \mathcal{S}^k, M^k)$ the k-fold product measure space of (S, \mathcal{S}, M) . Then we consider a probability

space
$$(\Omega^*, \mathcal{F}^*, \mathbf{P}^*)$$
 defined by (2.1) $\Omega^* = \bigcup_{k=0}^{\infty} \mathbf{S}^k$, $\mathcal{F}^* = \{A^* = \bigcup_{k=0}^{\infty} A_k; A_k \in \mathcal{S}^k \ (k \geq 0)\}$,

$$P^*(A^*) = \exp(-M(S)) \sum_{k=0}^{\infty} (k!)^{-1} M(S)^k P_k(A_k) \text{ for } A^* = \bigcup_{k=0}^{\infty} A_k \in \mathcal{F}^*,$$

where $(S^0, \mathcal{S}^0, \boldsymbol{P}_0)$ is the trivial probability space given by $S^0 = \{0\}$ and $\mathcal{S}^0 = \{\emptyset, S^0\}$. We call $(\Omega^*, \mathcal{F}^*, \boldsymbol{P}^*)$ the basic canonical probability space associated with (S, \mathcal{S}, M) . Let $\Phi: \Omega^* \to \mathcal{N}$ be an $\mathcal{F}^*/\mathfrak{B}(\mathcal{N})$ -measurable map given by $\Phi(0) = 0$ and

(2.2)
$$\langle \boldsymbol{\Phi}(\omega^*), f \rangle = \sum_{\ell=1}^k f(\boldsymbol{p}_{\ell}(\omega^*))$$
 for $f \in \mathcal{F}^+(S)$ when $\omega^* = (\boldsymbol{p}_1(\omega^*), \dots, \boldsymbol{p}_k(\omega^*)) \in S^k(k \geq 1)$. Then we obtain a Poisson

random measure Φ on S with intensity M with respect to P^* . We define

(2.3)
$$\Lambda^*(A, \omega^*) = v(A) + \int \int_{A \times \mathbf{R}_0} x \Phi(dt dx, \omega^*) - \int \int_{A \times J} x M(dt dx)$$
$$(A \in \mathcal{T}, \omega^* \in \Omega^*),$$

where we put $\Phi(U, \omega^*) = [\Phi(\omega^*)](U)$ for $U \in \mathcal{S}$ and $\omega^* \in \Omega^*$. Then we

Proposition 1. The process $\Lambda^* = \{\Lambda^*(A) ; A \in \mathcal{I}\}$ is an ID random measure on T which is defined on $(\Omega^*, \mathcal{F}^*, \mathbf{P}^*)$ and characterized by $\Lambda^* = {}^d$

Case (II): $M(S) = \infty$. On account of (1.2) we can choose a sequence $\{S_n; n \geq 1\} \subset \mathcal{S}$ of disjoint subsets of S satisfying $S = \bigcup_{n=1}^{\infty} S_n$ and 0 $< M(S_n) < \infty \ (n \ge 1)$. Let $\{M_n; n \ge 1\}$ be a sequence of finite measures on (S, \mathcal{A}) defined by $M_n(U) = M(U \cap S_n)$ for $U \in \mathcal{A}$. Let us introduce an infinite product probability space

(2.4)
$$(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\boldsymbol{P}}) = \prod_{n=1}^{\infty} (\Omega^*, \mathcal{F}^*, \boldsymbol{P}_n^*),$$

(2.4) $(\tilde{\mathcal{Q}}, \tilde{\mathcal{F}}, \tilde{\boldsymbol{P}}) = \prod_{n=1}^{\infty} (\mathcal{Q}^*, \mathcal{F}^*, \boldsymbol{P}_n^*),$ where $(\mathcal{Q}^*, \mathcal{F}^*, \boldsymbol{P}_n^*)$ is the basic canonical probability space associated with (S, \mathcal{S}, M_n) . We call $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$ the canonical probability space assoacited with decomposition $M = \sum_{n=1}^{\infty} M_n$ on (S, \mathcal{L}) . Then we have Poisson random measures $\Psi_n = \sum_{\ell=1}^n (\boldsymbol{\Phi} \circ \pi_\ell)$ and $\Psi = \sum_{n=1}^\infty (\boldsymbol{\Phi} \circ \pi_n)$ on S respectively with intensities $M_{(n)} = \sum_{\ell=1}^n M_\ell$ and M. Here π_n denotes the n-th projection map from $\tilde{\Omega} = (\Omega^*)^{\infty}$ onto Ω^* . Now we define, for each $n \geq 1$,

(2.5)
$$\tilde{\Lambda}_{n}(A, \, \tilde{\omega}) = v(A) + \int \int_{A \times \mathbf{R}_{0}} x \Psi_{n}(dt dx, \, \tilde{\omega}) - \int \int_{A \times J} x M_{(n)}(dt dx)$$

$$(A \in \mathcal{T}, \, \tilde{\omega} \in \tilde{\Omega}).$$

On account of the Lévy's equivalence theorem on the convergence of series with independent summands, we can find a random variable $\tilde{\Lambda}_{\infty}(A)$ to which $\{\tilde{\Lambda}_n(A)\}\$ converges almost surely as $n\to\infty$. Thus we have

Proposition 2. The process $\tilde{\Lambda}_{\infty} = {\{\tilde{\Lambda}_{\infty}(A) ; A \in \mathcal{I}\}}$ is an **ID** random measure on T which is defined on $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$ and characterized by $\tilde{\Lambda}_{\infty} = {}^{d} [v, M]$.

In the rest of this section we are concerned with a realization of Λ based on the space $\mathcal{N} = \mathcal{N}(S)$. We mean by $(\mathcal{N}, \mathcal{B}(\mathcal{N}), \mathbf{Q}^{M})$ a probability space given by

 $Q^M = [P^*]_{\varphi}$ in Case (I) and $Q^M = [\tilde{P}]_{\psi}$ in Case (II), where $[P^*]_{\sigma}$ and $[\tilde{P}]_{\pi}$ stand for the images of P^* and \tilde{P} induced by Φ and Ψ respectively. Then the identity map I on $(\mathcal{N}, \mathcal{B}(\mathcal{N}), \mathbf{Q}^{M})$ is considered as a Poisson random measure on S with intensity M. Furthermore we can realize $\mathbf{\Lambda} = {}^{d}[v, M]$ in the space of \mathbf{R} -valued signed measures on \mathcal{T} whenever the following conditions are satisfied.

(2.7) For each $A \in \mathcal{I}$, there exists $n \geq 1$ such that $A \subset T_n$;

(2.8)
$$\int \int_{A \times I} |x| M(dtdx) \equiv m(A) < \infty \quad (A \in \mathcal{T}).$$

Let $\mathbf{H}^+ = \{H^+(A) ; A \in \mathcal{I}\}, \mathbf{H}^- = \{H^-(A) ; A \in \mathcal{I}\}$ and $\mathbf{H} = \{H(A) ; A \in \mathcal{I}\}$ $A \in \mathcal{I}$ be **ID** random measures on **T**, which are defined on $(\mathcal{N}, \mathcal{B}(\mathcal{N}),$ Q^{M}) and expressed as follows:

(2.9)
$$H^{\pm}(A, \nu) = v^{\pm}(A) + m(A) + \int \int_{A \times R_0} x^{\pm} \nu(dtdx) - \int \int_{A \times I} x^{\pm} M(dtdx),$$

$$(2.10) \quad H(A, \nu) = v(A) + \int \int_{A \times R_0} x \nu(dt dx) - \int \int_{A \times J} x M(dt dx).$$

Here $v=v^+-v^-$ stands for the Jordan decomposition of v. We put ${m R}_{\pm}=$ $\{\pm\;x>0\}$ and $M_{\pm}(U)=M(U\cap\;(T\times R_{\pm}))$ for $U\in\mathscr{S}$ respectively. Then

Theorem 1. Assume (2.7) and (2.8). Then \mathbf{H}^{\pm} and \mathbf{H} are characterized by $\mathbf{H}^{\pm} = {}^{d} [v^{\pm} + m, M_{+}]$ and $\mathbf{H} = {}^{d} [v, M]$. Furthermore \mathbf{H}^{+} and \mathbf{H}^{-} are independent and also there exists a set $\mathcal{N}_0 \in \mathcal{B}(\mathcal{N})$ with $\mathbf{Q}^{\mathbf{M}}(\mathcal{N}_0) = 1$ satisfying (2.11) $H(A, \nu) = H^{+}(A, \nu) - H^{-}(A, \nu)$ and

$$0 \le H^{\pm}(A, \nu) < \infty \quad (A \in \mathcal{T}, \nu \in \mathcal{N}_0).$$

§3. The law equivalence of infinitely divisible random measures. In what follows we discuss the equivalence problem for ID random measures on **T** based on the method stated in Section 2. Given σ -finite measures μ and ν on a measurable space (E, \mathcal{E}) , we mean by $\mu \sim \nu$ that μ and ν are equivalent, i.e., mutually absolutely continuous. The Hellinger-Kakutani distance and inner product are defined respectively by

$$\operatorname{dist}(\mu, \, \nu) = \left[\int_{E} \left(\sqrt{d\mu} - \sqrt{d\nu} \right)^{2} \right]^{1/2} \quad \text{and} \quad \varrho(\mu, \, \nu) = \int_{E} \sqrt{d\mu d\nu}.$$

Theorem 2. Let Λ_1 and Λ_2 be **ID** random measures on T given by $\Lambda_j = {}^d$ $[v_j, M^{(j)}]$ (j = 1, 2). Then $P_{\Lambda_1} \sim P_{\Lambda_2}$ if the following three conditions hold simultaneously: (E.1) $M^{(1)} \sim M^{(2)}$, (E.2) $\operatorname{dist}(M^{(1)}, M^{(2)}) < \infty$,

(E.3)
$$v_1(A) - v_2(A) = \int \int_{A \times I} x \{M^{(1)} - M^{(2)}\} (dt dx) \quad (A \in \mathcal{T}).$$

§4. The outline of the proof of Theorem 2. First we construct versions of $oldsymbol{\Lambda}_1$ and $oldsymbol{\Lambda}_2$ based on Poisson random measures on $oldsymbol{S}$ along the procedure stated in Section 2. For simplicity we devote ourselves to the case that $M^{(j)}(S)=\infty$ $(j=1,\,2).$ Then we can find a decomposition $S=\cup_{n=1}^\infty S_n$ with $0 < M^{(j)}(S_n) < \infty (n \ge 1, j = 1, 2)$. For each j = 1, 2, we construct the canonical probability space

$$(\tilde{\Omega},\,\tilde{\mathscr{F}},\,\tilde{\boldsymbol{P}}^{(j)})=\prod_{n=1}^{\infty}\,(\Omega^*,\,\mathscr{F}^*,\,\boldsymbol{P}_n^{*(j)})$$

 $(\tilde{\varOmega},\,\tilde{\mathscr{F}},\,\tilde{\boldsymbol{P}}^{(j)}) = \prod_{n=1}^{\infty}\,(\varOmega^*,\,\mathscr{F}^*,\,\boldsymbol{P}_n^{*(j)})$ associated with decomposition $M^{(j)} = \sum_{n=1}^{\infty}M_n^{(j)}$ on $(S,\,\varnothing)$, where we put $M_n^{(j)}(U) = M^{(j)}(U\cap S_n)$ for $U\in\varnothing$. Now (E.1) implies $M_n^{(1)}\sim M_n^{(2)}$ and also $\boldsymbol{P}_n^{*(1)}\sim\boldsymbol{P}_n^{*(2)}$ for each $n\geq 1$. Further (E.2) implies $(4.1) \quad \Pi_{n=1}^{\infty}\,\varrho(\boldsymbol{P}_n^{*(1)},\,\boldsymbol{P}_n^{*(2)}) = \exp[-\,(1/2)\operatorname{dist}(M^{(1)},\,M^{(2)})^2]>0$. Therefore we obtain $\tilde{\boldsymbol{P}}^{(1)}\sim\tilde{\boldsymbol{P}}^{(2)}$ by the Kakutani's theorem on the equivalence of infinite parameters of the second seco

lence of infinite product probability measures (see [2]). By applying Proposition 2, we obtain stochastic processes $\tilde{A}_{\infty}^{(j)} = \{\tilde{A}_{\infty}^{(j)}(A) ; A \in \mathcal{I}\}\ (j=1,2)$

satisfying the following two conditions. (4.2) $\tilde{\Lambda}_{\infty}^{(j)}$ is defined on $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\boldsymbol{P}}^{(j)})$ and characterized by $\tilde{\Lambda}_{\infty}^{(j)} = {}^{d} [v_j, M^{(j)}]$; (4.3) For each $A \in \mathcal{T}$, the sequence $\{\tilde{\Lambda}_{n}^{(j)}(A) : n \geq 1\}$ converges almost surely to $\tilde{\Lambda}_{\infty}^{(j)}(A)$ with respect to $\tilde{\boldsymbol{P}}^{(j)}$ as $n \to \infty$, where we put $M_{(n)}^{(j)} = \sum_{\ell=1}^{n} M_{\ell}^{(j)}$ and

$$(4.4) \qquad \tilde{\Lambda}_{n}^{(j)}(A, \ \tilde{\omega}) = v_{j}(A) + \int \int_{A \times \mathbf{R}_{0}} x \Psi_{n}(dtdx, \ \tilde{\omega}) - \int \int_{A \times J} x M_{(n)}^{(j)}(dtdx)$$

On account of (E.2) and (4.4) we have the following equations:

$$(4.5) \quad \lim_{n\to\infty} \int \int_{A\times I} x \{M_{(n)}^{(1)} - M_{(n)}^{(2)}\} (dtdx) = \int \int_{A\times I} x \{M^{(1)} - M^{(2)}\} (dtdx),$$

$$(4.6) \quad \tilde{\Lambda}_{n}^{(1)}(A, \tilde{\omega}) - \tilde{\Lambda}_{n}^{(2)}(A, \tilde{\omega}) = v_{1}(A) - v_{2}(A) - \int \int_{A \times I} x \{M_{(n)}^{(1)} - M_{(n)}^{(2)}\} (dtdx)$$

for $A \in \mathcal{T}$, $\tilde{\omega} \in \tilde{\Omega}$, and $n \geq 1$. Therefore combining (E.3) with $\tilde{\boldsymbol{P}}^{(1)} \sim \tilde{\boldsymbol{P}}^{(2)}$ yields that $\tilde{\Lambda}_{\infty}^{(1)}(A) = \tilde{\Lambda}_{\infty}^{(2)}(A)$ a.s. with respect to $\tilde{\boldsymbol{P}}^{(1)}$ and also $\tilde{\boldsymbol{P}}^{(2)}$. Now putting $\boldsymbol{\Theta}(A, \tilde{\omega}) = \tilde{\Lambda}_{\infty}^{(1)}(A, \tilde{\omega})$ for $A \in \mathcal{T}$ and $\tilde{\omega} \in \tilde{\Omega}$, we have a process $\boldsymbol{\Theta} = \tilde{\boldsymbol{\Omega}}$

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References

- [1] O. Kallenberg: Random Measures. 4th ed., Academic Press (1986).
- S. Kakutani: On equivalence of infinite product measures. Ann. Math., 49, 239-247 (1948).
- [3] B. S. Rajput and J. Rosinski: Spectral representations of infinitely divisible processes. Probab. Th. Rel. Fields, 82, 451-487 (1989).