# 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 $\boldsymbol{T}$ based on Poisson random measures on $\boldsymbol{S}=\boldsymbol{T} \times$ ( $\boldsymbol{R} \backslash\{0\}$ ). As an application we discuss the equivalence problem for infinitely divisible random measures on $\boldsymbol{T}$.
§1. Preliminaries. Let $\boldsymbol{T}$ be an arbitrary nonempty set and $\mathscr{T}$ be a $\delta$-ring of subsets of $\boldsymbol{T}$. We assume there exists an increasing sequence $\left\{\boldsymbol{T}_{n}\right.$; $n \geq 1\} \subset \mathscr{T}$ with $\boldsymbol{T}=\cup_{n=1}^{\infty} \boldsymbol{T}_{n}$ and $\{t\} \in \mathscr{T}$ for each $t \in \boldsymbol{T}$. Let $\boldsymbol{\Lambda}=$ $\{\Lambda(A) ; A \in \mathscr{T}\}$ be an infinitely divisible random measure (or ID random measure) on $\boldsymbol{T}$ with no Gaussian component, which is defined on a basic probability space $(\Omega, \mathscr{F}, \boldsymbol{P})$ (see [3]). In other words, $\boldsymbol{\Lambda}$ is a real stochastic process characterized by

$$
\begin{equation*}
\boldsymbol{E}[\exp (i z \Lambda(A))]=\exp \left[i z v(A)+\iint_{A \times \boldsymbol{R}_{0}} g(z, x) M(d t d x)\right] \tag{1.1}
\end{equation*}
$$

$$
(z \in \boldsymbol{R}, A \in \mathscr{T})
$$

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

$$
\begin{equation*}
\iint_{A \times \boldsymbol{R}_{0}}\left(1 \wedge x^{2}\right) M(d t d x)<\infty \quad(A \in \mathscr{T}) \tag{1.2}
\end{equation*}
$$

We mean by $\boldsymbol{\Lambda}={ }^{d}[v, M]$ that the probability law of $\boldsymbol{\Lambda}$ is determined by parameters $v$ and $M$. We denote by $\boldsymbol{P}_{A}$ the probability measure on a measurable space $\left(\boldsymbol{R}^{\mathscr{T}}, \mathscr{F}\left(\boldsymbol{R}^{\mathscr{T}}\right)\right.$ ) induced by the map $\Lambda: \Omega \ni \omega \rightarrow \Lambda(\cdot, \omega) \in \boldsymbol{R}^{\mathscr{T}}$, where $\boldsymbol{R}^{\mathscr{T}}$ is the set of all $\boldsymbol{R}$-valued functions on $\mathscr{T}$ and $\mathscr{F}\left(\boldsymbol{R}^{\mathscr{T}}\right)$ is the $\sigma$-algebra on $\boldsymbol{R}^{\mathscr{T}}$ generated by all coordinate functions. The product measurable space $(\boldsymbol{S}, \mathscr{S})$ is given by $\mathscr{\&}=\sigma(\mathscr{T}) \otimes \mathscr{B}\left(\boldsymbol{R}_{0}\right)$, where $\sigma(\mathscr{T})$ is the $\sigma$-algebra on $\boldsymbol{T}$ generated by $\mathscr{T}$ and $\mathscr{B}\left(\boldsymbol{R}_{0}\right)$ is the Borel $\sigma$-algebra on $\boldsymbol{R}_{0}$. Let $\mathcal{N}=\mathcal{N}(\boldsymbol{S})$ be the totality of nonnegative (possibly infinite) integer-valued measures on $(\boldsymbol{S}, \mathscr{\infty})$. Let $\mathscr{F}^{+}(\boldsymbol{S})$ be the set of all nonnegative measurable functions on $(\boldsymbol{S}, \mathscr{\&})$. We denote by $\mathscr{B}(\mathcal{N})$ the $\sigma$-algebra on $\mathcal{N}$ generated by all functions $f^{*}$ on $\mathcal{N}$ given by

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

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

$$
\begin{array}{r}
\boldsymbol{E}[\exp (-\langle\xi, f\rangle)]=\exp \left[-\iint_{S}\{1-\exp (-f(t, x))\} M(d t d x)\right]  \tag{1.3}\\
\text { for } f \in \mathscr{F}^{+}(\boldsymbol{S}) .
\end{array}
$$

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

Case (I): $\boldsymbol{M}(\boldsymbol{S})<\infty$. For each $k \geq 1$, let $\left(\boldsymbol{S}^{k}, \delta^{k}, \boldsymbol{P}_{k}\right)$ be a probability space given by $\boldsymbol{P}_{k}=M(\boldsymbol{S})^{-k} M^{k}$, where we mean by $\left(\boldsymbol{S}^{k}, \delta^{k}, M^{k}\right)$ the $k$-fold product measure space of ( $\boldsymbol{S}, \boldsymbol{\&}, M$ ). Then we consider a probability space $\left(\Omega^{*}, \mathscr{F}^{*}, \boldsymbol{P}^{*}\right)$ defined by

$$
\begin{equation*}
\Omega^{*}=\bigcup_{k=0}^{\infty} \mathbf{S}^{k}, \quad \mathscr{F}^{*}=\left\{A^{*}=\cup_{k=0}^{\infty} A_{k} ; A_{k} \in \delta^{k}(k \geq 0)\right\} \tag{2.1}
\end{equation*}
$$

$$
\boldsymbol{P}^{*}\left(A^{*}\right)=\exp (-M(\boldsymbol{S})) \sum_{k=0}^{\infty}(k!)^{-1} M(\boldsymbol{S})^{k} \boldsymbol{P}_{k}\left(A_{k}\right) \text { for } A^{*}=\bigcup_{k=0}^{\infty} A_{k} \in \mathscr{F}^{*}
$$

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

$$
\begin{equation*}
\left\langle\Phi\left(\omega^{*}\right), f\right\rangle=\sum_{t=1}^{k} f\left(p_{\imath}\left(\omega^{*}\right)\right) \quad \text { for } \quad f \in \mathscr{F}^{+}(\boldsymbol{S}) \tag{2.2}
\end{equation*}
$$

when $\omega^{*}=\left(p_{1}\left(\omega^{*}\right), \cdots, p_{k}\left(\omega^{*}\right)\right) \in S^{k}(k \geq 1)$. Then we obtain a Poisson random measure $\Phi$ on $\boldsymbol{S}$ with intensity $M$ with respect to $\boldsymbol{P}^{*}$. We define

$$
\begin{equation*}
\Lambda^{*}\left(A, \omega^{*}\right)=v(A)+\iint_{A \times \boldsymbol{R}_{0}} x \Phi\left(d t d x, \omega^{*}\right)-\iint_{A \times J} x M(d t d x) \tag{2.3}
\end{equation*}
$$

$$
\left(A \in \mathscr{T}, \omega^{*} \in \Omega^{*}\right)
$$

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

Proposition 1. The process $\boldsymbol{\Lambda}^{*}=\left\{\Lambda^{*}(A) ; A \in \mathscr{T}\right\}$ is an ID random measure on $\boldsymbol{T}$ which is defined on $\left(\Omega^{*}, \mathscr{F}^{*}, \boldsymbol{P}^{*}\right)$ and characterized by $\boldsymbol{\Lambda}^{*}={ }^{d}$ [ $v, M]$.

Case (II): $M(\boldsymbol{S})=\infty$. On account of (1.2) we can choose a sequence $\left\{\boldsymbol{S}_{n} ; n \geq 1\right\} \subset \mathscr{S}$ of disjoint subsets of $\boldsymbol{S}$ satisfying $\boldsymbol{S}=\cup_{n=1}^{\infty} \boldsymbol{S}_{n}$ and 0 $<M\left(S_{n}\right)<\infty(n \geq 1)$. Let $\left\{M_{n} ; n \geq 1\right\}$ be a sequence of finite measures on $(\boldsymbol{S}, \mathscr{\&})$ defined by $M_{n}(U)=M\left(U \cap \boldsymbol{S}_{n}\right)$ for $U \in \mathscr{\&}$. Let us introduce an infinite product probability space

$$
\begin{equation*}
(\tilde{\Omega}, \tilde{\mathscr{F}}, \tilde{\boldsymbol{P}})=\prod_{n=1}^{\infty}\left(\Omega^{*}, \mathscr{F}^{*}, \boldsymbol{P}_{n}^{*}\right) \tag{2.4}
\end{equation*}
$$

where $\left(\Omega^{*}, \mathscr{F}^{*}, \boldsymbol{P}_{n}^{*}\right)$ is the basic canonical probability space associated with $\left(\boldsymbol{S}, \mathscr{\&}, M_{n}\right)$. We call ( $\left.\tilde{\Omega}, \tilde{\mathscr{F}}, \tilde{\boldsymbol{P}}\right)$ the canonical probability space assoacited with decomposition $M=\sum_{n=1}^{\infty} M_{n}$ on ( $\boldsymbol{S}, \boldsymbol{\&}$ ). Then we have Poisson random measures $\Psi_{n}=\sum_{c=1}^{n}\left(\Phi \circ \pi_{l}\right)$ and $\Psi=\sum_{n=1}^{\infty}\left(\Phi \circ \pi_{n}\right)$ on $\boldsymbol{S}$ respectively with intensities $M_{(n)}=\sum_{c=1}^{n} M_{c}$ and $M$. Here $\pi_{n}$ denotes the $n$-th projection map from $\tilde{\Omega}=\left(\Omega^{*}\right)^{\infty}$ onto $\Omega^{*}$. Now we define, for each $n \geq 1$,

$$
\begin{equation*}
\tilde{\Lambda}_{n}(A, \tilde{\omega})=v(A)+\iint_{A \times \boldsymbol{R}_{0}} x \Psi_{n}(d t d x, \tilde{\omega})-\iint_{A \times J} x M_{(n)}(d t d x) \tag{2.5}
\end{equation*}
$$

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
$\left\{\tilde{\Lambda}_{n}(A)\right\}$ converges almost surely as $n \rightarrow \infty$. Thus we have
Proposition 2. The process $\tilde{\boldsymbol{\Lambda}}_{\infty}=\left\{\tilde{\Lambda}_{\infty}(A) ; A \in \mathscr{T}\right\}$ is an ID random measure on $\boldsymbol{T}$ which is defined on $(\tilde{\Omega}, \tilde{\mathscr{F}}, \tilde{\boldsymbol{P}})$ and characterized by $\tilde{\boldsymbol{\Lambda}}_{\infty}={ }^{d}[v, M]$.

In the rest of this section we are concerned with a realization of $\boldsymbol{\Lambda}$ based on the space $\mathcal{N}=\mathcal{N}(\boldsymbol{S})$. We mean by $\left(\mathcal{N}, \mathscr{B}(\mathcal{N}), \boldsymbol{Q}^{M}\right)$ a probability space given by
(2.6) $\quad \boldsymbol{Q}^{M}=\left[\boldsymbol{P}^{*}\right]_{\Phi} \quad$ in Case (I) and $\quad \boldsymbol{Q}^{M}=[\tilde{\boldsymbol{P}}]_{\Psi} \quad$ in Case (II), where $\left[\boldsymbol{P}^{*}\right]_{\Phi}$ and $[\tilde{\boldsymbol{P}}]_{\Psi}$ stand for the images of $\boldsymbol{P}^{*}$ and $\tilde{\boldsymbol{P}}$ induced by $\Phi$ and $\Psi$ respectively. Then the identity map I on $\left(\mathcal{N}, \mathscr{B}(\mathcal{N}), \boldsymbol{Q}^{M}\right)$ is considered as a Poisson random measure on $\boldsymbol{S}$ with intensity $M$. Furthermore we can realize $\boldsymbol{\Lambda}={ }^{d}[v, M]$ in the space of $\boldsymbol{R}$-valued signed measures on $\mathscr{T}$ whenever the following conditions are satisfied.
(2.7) For each $A \in \mathscr{T}$, there exists $n \geq 1$ such that $A \subset \boldsymbol{T}_{n}$;

$$
\begin{equation*}
\iint_{A \times J}|x| M(d t d x) \equiv m(A)<\infty \quad(A \in \mathscr{T}) \tag{2.8}
\end{equation*}
$$

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

$$
\begin{align*}
H^{ \pm}(A, \nu) & =v^{ \pm}(A)+m(A)+\iint_{A \times \boldsymbol{R}_{0}} x^{ \pm} \nu(d t d x)-\iint_{A \times J} x^{ \pm} M(d t d x)  \tag{2.9}\\
H(A, \nu) & =v(A)+\iint_{A \times \boldsymbol{R}_{0}} x \nu(d t d x)-\iint_{A \times J} x M(d t d x) \tag{2.10}
\end{align*}
$$

Here $v=v^{+}-v^{-}$stands for the Jordan decomposition of $v$. We put $\boldsymbol{R}_{ \pm}=$ $\{ \pm x>0\}$ and $M_{ \pm}(U)=M\left(U \cap\left(\boldsymbol{T} \times \boldsymbol{R}_{ \pm}\right)\right)$for $U \in \&$ respectively. Then we have

Theorem 1. Assume (2.7) and (2.8). Then $\boldsymbol{H}^{ \pm}$and $\boldsymbol{H}$ are characterized by $\boldsymbol{H}^{ \pm}={ }^{d}\left[v^{ \pm}+m, M_{ \pm}\right]$and $\boldsymbol{H}={ }^{d}[v, M]$. Furthermore $\boldsymbol{H}^{+}$and $\boldsymbol{H}^{-}$are independent and also there exists a set $\mathcal{N}_{0} \in \mathscr{B}(\mathcal{N})$ with $\boldsymbol{Q}^{M}\left(\mathcal{N}_{0}\right)=1$ satisfying

$$
\begin{align*}
& H(A, \nu)=H^{+}(A, \nu)-H^{-}(A, \nu) \text { and }  \tag{2.11}\\
& \\
& 0 \leq H^{ \pm}(A, \nu)<\infty \quad\left(A \in \mathscr{T}, \nu \in \mathcal{N}_{0}\right) .
\end{align*}
$$

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

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

Theorem 2. Let $\boldsymbol{\Lambda}_{1}$ and $\boldsymbol{\Lambda}_{2}$ be ID random measures on $\boldsymbol{T}$ given by $\boldsymbol{\Lambda}_{j}={ }^{d}$ $\left[v_{j}, M^{(j)}\right](j=1,2)$. Then $\boldsymbol{P}_{\Lambda_{1}} \sim \boldsymbol{P}_{\Lambda_{2}}$ if the following three conditions hold simultaneously:
(E.1) $\quad M^{(1)} \sim M^{(2)}$,
(E.2) $\operatorname{dist}\left(M^{(1)}, M^{(2)}\right)<\infty$,

$$
\begin{equation*}
v_{1}(A)-v_{2}(A)=\iint_{A \times J} x\left\{M^{(1)}-M^{(2)}\right\}(d t d x) \quad(A \in \mathscr{T}) \tag{E.3}
\end{equation*}
$$

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

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

associated with decomposition $M^{(j)} \stackrel{n=1}{=} \sum_{n=1}^{\infty} M_{n}^{(j)}$ on (S, \&), where we put $M_{n}^{(j)}(U)=M^{(j)}\left(U \cap S_{n}\right)$ for $U \in \mathscr{A}$. 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) $\quad \Pi_{n=1}^{\infty} \varrho\left(\boldsymbol{P}_{n}^{*(1)}, \boldsymbol{P}_{n}^{*(2)}\right)=\exp \left[-(1 / 2) \operatorname{dist}\left(M^{(1)}, M^{(2)}\right)^{2}\right]>0$.

Therefore we obtain $\tilde{\boldsymbol{P}}^{n(1)} \sim \tilde{\boldsymbol{P}}^{(2)}$ by the Kakutani's theorem on the equivalence of infinite product probability measures (see [2]). By applying Proposition 2, we obtain stochastic processes $\tilde{\Lambda}_{\infty}^{(j)}=\left\{\tilde{\Lambda}_{\infty}^{(j)}(A) ; A \in \mathscr{T}\right\} \quad(j=1,2)$ satisfying the following two conditions.
(4.2) $\tilde{\boldsymbol{\Lambda}}_{\infty}^{(j)}$ is defined on ( $\tilde{\Omega}, \tilde{\mathscr{F}}, \tilde{\boldsymbol{P}}^{(j)}$ ) and characterized by $\tilde{\boldsymbol{\Lambda}}_{\infty}^{(j)}={ }^{d}\left[v_{j}, M^{(j)}\right]$; (4.3) For each $A \in \mathscr{T}$, the sequence $\left\{\tilde{\Lambda}_{n}^{(j)}(A) ; n \geq 1\right\}$ converges almost surely to $\tilde{\Lambda}_{\infty}^{(j)}(A)$ with respect to $\tilde{\boldsymbol{P}}^{(j)}$ as $n \rightarrow \infty$, where we put $M_{(n)}^{(j)}=\sum_{t=1}^{n} M_{\iota}^{(j)}$ and

$$
\begin{array}{r}
\tilde{\Lambda}_{n}^{(j)}(A, \tilde{\omega})=v_{j}(A)+\iint_{A \times \boldsymbol{R}_{0}} x \Psi_{n}(d t d x, \tilde{\omega})-\iint_{A \times J} x M_{(n)}^{(j)}(d t d x)  \tag{4.4}\\
(A \in \mathscr{T}, \tilde{\omega} \in \tilde{\Omega})
\end{array}
$$

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

$$
\begin{equation*}
\lim _{n \rightarrow \infty} \iint_{A \times J} x\left\{M_{(n)}^{(1)}-M_{(n)}^{(2)}\right\}(d t d x)=\iint_{A \times J} x\left\{M^{(1)}-M^{(2)}\right\}(d t d x) \tag{4.5}
\end{equation*}
$$

$$
\begin{equation*}
\tilde{\Lambda}_{n}^{(1)}(A, \tilde{\omega})-\tilde{\Lambda}_{n}^{(2)}(A, \tilde{\omega})=v_{1}(A)-v_{2}(A)-\iint_{A \times J} x\left\{M_{(n)}^{(1)}-M_{(n)}^{(2)}\right\}(d t d x) \tag{4.6}
\end{equation*}
$$ for $A \in \mathscr{T}, \tilde{\omega} \in \tilde{\Omega}$, 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 $\Theta(A, \tilde{\omega})=\tilde{\Lambda}_{\infty}^{(1)}(A, \tilde{\omega})$ for $A \in \mathscr{T}$ and $\tilde{\omega} \in \tilde{\Omega}$, we have a process $\boldsymbol{\Theta}=$ $\{\Theta(A) ; A \in \mathscr{T}\}$ which is characterized by

(4.7) $\quad \boldsymbol{\Theta}={ }^{d}\left[v_{j}, M^{(j)}\right] \quad$ with respect to $\tilde{\boldsymbol{P}}^{(j)} \quad(j=1,2)$.

This implies the equalities $\boldsymbol{P}_{\Lambda_{j}}=\left[\tilde{\boldsymbol{P}}^{(j)}\right]_{\boldsymbol{\theta}}(j=1,2)$, where $\left[\tilde{\boldsymbol{P}}^{(j)}\right]_{\theta_{\mathscr{T}}}$ stands for the image of $\tilde{\boldsymbol{P}}^{(j)}$ induced by the map $\Theta: \tilde{\Omega} \ni \tilde{\omega} \rightarrow \Theta(\cdot, \tilde{\omega}) \in \boldsymbol{R}^{\mathscr{T}}$. Thus we obtain the desired relation $\boldsymbol{P}_{\Lambda_{1}} \sim \boldsymbol{P}_{\Lambda_{2}}$ from $\tilde{\boldsymbol{P}}^{(1)} \sim \tilde{\boldsymbol{P}}^{(2)}$.

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