

## A LIMIT THEOREM ON $(J, X)$ -PROCESSES

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1. Let  $\{(J_n, X_n); n=0, 1, 2, \dots\}$  be a two-dimensional stochastic process with the state space  $I_r \times R$ , where  $I_r = \{1, 2, \dots, r\}$  and  $R = (-\infty, \infty)$ , and let  $\{Q_{jk}(\cdot); j, k=1, 2, \dots, r\}$  be a family of non-decreasing functions defined on  $R$ , where  $Q_{jk}(-\infty)=0$  for  $j, k=1, 2, \dots, r$  and  $\sum_{k=1}^r Q_{jk}(+\infty)=1$  for  $j=1, 2, \dots, r$ . If  $X_0 \equiv 0$  and

$$(1) \quad P\{J_n=k, X_n \leq x | (J_0, X_0), \dots, (J_{n-1}, X_{n-1})\} = Q_{J_{n-1}, k}(x) \quad (\text{a. s.})$$

for all  $(k, x) \in I_r \times R$ , then  $\{(J_n, X_n); n=0, 1, 2, \dots\}$  is called a  $(J, X)$ -process, which has been introduced by Pyke [2].  $f$  being a real-valued Baire function defined on  $I_r \times R$ , the random variable

$$\frac{1}{\sqrt{n}} \sum_{\nu=1}^n [f(J_\nu, X_\nu) - E\{f(J_\nu, X_\nu)\}]$$

is asymptotically normally distributed as  $n \rightarrow \infty$ . Taga [3] has proved this fact in the cases where  $f(k, x) \equiv x$  and  $f(k, x) \equiv \delta_{jk}x$ , respectively. In this paper, we shall give an alternative proof of this fact in a general form, which is also regarded as an extension of the consequence in section 3 of [1].

2. Firstly, consider the  $r \times r$  matrix  $P = (p_{jk})$  where  $p_{jk} \stackrel{\text{def}}{=} Q_{jk}(+\infty)$ . When there exists a natural number  $m$  such that every element of the matrix  $P^m$  is strictly positive, then we have  $z=1$  as a simple root of the equation  $\det(I - zP) = 0$ , where  $I$  is the  $r \times r$  identity matrix, and it is known that  $|\alpha_l| > 1$  ( $l=1, 2, \dots, k$ ), where  $\alpha_1, \dots, \alpha_k$  are the remaining roots of  $\det(I - zP) = 0$ . In what follows, we assume that the equation  $\det(I - zP) = 0$  has the simple roots  $1, \alpha_1, \dots, \alpha_{r-1}$ , where  $|\alpha_l| > 1$  ( $l=1, 2, \dots, r-1$ ). Secondly, we introduce a family

$$\{R_n(j, k, t); j \in I_r, k \in I_r, t \in R, n=1, 2, \dots\}$$

of real-valued random variables. And we set the following assumptions:

(i) the characteristic functions  $\chi_{jki}(\theta) \stackrel{\text{def}}{=} E\{e^{i\theta R_n(j, k, t)}\}$  of  $R_n(j, k, t)$ , where  $i = \sqrt{-1}$ , are independent of  $n$  and  $dQ_{jk}$ -measurable on  $t$  for any fixed  $(j, k, \theta)$ ,

(ii)  $\{R_n(j, k, t); n \geq 1, j \in I_r, k \in I_r, t \in R\}$  and  $\{(J_n, X_n); n \geq 0\}$  are mutually independent,

(iii)  $R_1(k, k_1, t_1), R_2(k_1, k_2, t_2), \dots$  are mutually independent for every  $(k, k_1, k_2, \dots; t_1, t_2, \dots)$ ,

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(iv)  $Y_n \stackrel{\text{def}}{=} R_n(J_{n-1}, J_n, X_n)$  is a random variable or a measurable function on the probability space for every  $n \geq 1$ , and

(v) the functions

$$\eta_{jk}(\theta) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} \chi_{jkt}(\theta) dQ_{jk}(t),$$

where  $j \in I_r$  and  $k \in I_r$ , have continuous derivatives of the 2nd order in a neighborhood of  $\theta=0$ , respectively. Then, we have the following

**THEOREM.** *Under the assumptions mentioned above,  $(Y_1 + \dots + Y_n - n\mu) / \sqrt{n}$  converges in distribution to a normal distribution, where  $\mu$  is a constant.*

*Proof.* Let  $\varphi_{kn}(\theta) \stackrel{\text{def}}{=} E\{e^{i\theta(Y_1 + \dots + Y_n)} | J_0 = k\}$  be the characteristic function of  $Y_1 + \dots + Y_n$  given that  $J_0 = k$ , where  $i = \sqrt{-1}$  and  $\theta$  is a real-valued variable. Then, from (1), (i), (ii), (iii) and (iv), we have

$$\begin{aligned} \varphi_{kn}(\theta) &= E\{e^{i\theta(Y_1 + \dots + Y_n)} | J_0 = k\} \\ &= \sum_{k_1, \dots, k_n=1}^r \int_{t_1=-\infty}^{\infty} \dots \int_{t_n=-\infty}^{\infty} P\{J_\nu = k_\nu, t_\nu \leq X_\nu < t_\nu + dt_\nu \ (\nu=1, 2, \dots, n) | J_0 = k\} \\ &\quad \times E\{e^{i\theta(R_1(k, k_1, t_1) + \dots + R_n(k_{n-1}, k_n, t_n))} | J_0 = k, J_\nu = k_\nu, X_\nu = t_\nu \ (\nu=1, 2, \dots, n)\} \\ (2) \quad &= \sum_{k_1, \dots, k_n} \int \dots \int dQ_{kk_1}(t_1) dQ_{k_1k_2}(t_2) \dots dQ_{k_{n-1}k_n}(t_n) \chi_{kk_1t_1}(\theta) \chi_{k_1k_2t_2}(\theta) \dots \chi_{k_{n-1}k_nt_n}(\theta) \\ &= \sum_{k_1, \dots, k_n} \eta_{kk_1}(\theta) \eta_{k_1k_2}(\theta) \dots \eta_{k_{n-1}k_n}(\theta). \end{aligned}$$

Introducing the  $r \times r$  matrix  $H(\theta) \stackrel{\text{def}}{=} (\eta_{jk}(\theta))$ , the  $r$ -dimensional vectors

$$\boldsymbol{\varphi}_n(\theta) \stackrel{\text{def}}{=} \begin{bmatrix} \varphi_{1n}(\theta) \\ \vdots \\ \varphi_{rn}(\theta) \end{bmatrix} \quad \text{and} \quad \boldsymbol{e} \stackrel{\text{def}}{=} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix},$$

(2) may be described as  $\boldsymbol{\varphi}_n(\theta) = H(\theta)^n \boldsymbol{e}$  and so

$$(3) \quad \boldsymbol{\varphi}_n(\theta) = H(\theta) \boldsymbol{\varphi}_{n-1}(\theta) \quad (n=1, 2, \dots),$$

where  $\boldsymbol{\varphi}_0(\theta) = \boldsymbol{e}$ . Since  $\lim_{\theta \rightarrow 0} H(\theta) = H(0) = P$ , the equation  $\det(I - zH(\theta)) = 0$  has the  $r$  simple roots  $\zeta_0(\theta), \zeta_1(\theta), \dots, \zeta_{r-1}(\theta)$ , which satisfy that

$$\zeta_0(\theta) \rightarrow 1, \quad \zeta_1(\theta) \rightarrow \alpha_1, \quad \dots, \quad \zeta_{r-1}(\theta) \rightarrow \alpha_{r-1} \quad \text{as } \theta \rightarrow 0$$

and

$$(4) \quad \left| \frac{\zeta_0(\theta)}{\zeta_l(\theta)} \right| < \rho < 1 \quad \text{for } |\theta| < \theta_0 \quad \text{and } l=1, 2, \dots, r-1,$$

where  $\rho$  and  $\theta_0$  are some positive constants. Then, applying the method used in section 3 of [1], we get

$$(5) \quad \varphi_n(\theta) = \sum_{l=0}^{r-1} \frac{\tau_l(\theta)}{\zeta_l(\theta)^n}$$

and

$$(6) \quad \tau_0(0) = e,$$

where  $\tau_l(\theta)$  ( $l=0, 1, \dots, r-1$ ) are  $r$ -dimensional vectors independent of  $z$ . Introducing the vector  $\pi = [\pi_1, \dots, \pi_r]$  of initial probabilities, where  $\pi_k = P\{J_0 = k\}$  ( $k=1, 2, \dots, r$ ), it follows that

$$(7) \quad \varphi_n(\theta) \stackrel{\text{def}}{=} E\{e^{i\theta(Y_1 + \dots + Y_n)}\} = \sum_{k=1}^r \pi_k \varphi_{kn}(\theta) = \pi \cdot \varphi_n(\theta) = \sum_{l=0}^{r-1} \frac{\tau_l(\theta)}{\zeta_l(\theta)^n}$$

and

$$(8) \quad \tau_0(0) = \pi \cdot \tau_0(0) = \pi \cdot e = 1,$$

where  $\tau_l(\theta) = \pi \cdot \tau_l(\theta)$  ( $l=0, 1, \dots, r-1$ ). Now, from the assumption (v), we know that  $\zeta_l(\theta)$  and  $\tau_l(\theta)$  ( $l=0, 1, \dots, r-1$ ) have continuous derivatives of the 2nd order in a neighborhood of  $\theta=0$ . Therefore, we have

$$(9) \quad \varphi_n'(\theta) = \sum_{l=0}^{r-1} \left\{ - \frac{n \zeta_l'(\theta) \tau_l(\theta)}{\zeta_l(\theta)^{n+1}} + \frac{\tau_l'(\theta)}{\zeta_l(\theta)^n} \right\},$$

which implies with (8) that

$$(10) \quad \begin{aligned} i \cdot E\{Y_1 + \dots + Y_n\} &= \varphi_n'(0) \\ &= -n \zeta_0'(0) + \tau_0'(0) + \sum_{l=0}^{r-1} \left\{ - \frac{n \zeta_l'(0) \tau_l(0)}{\alpha_l^{n+1}} + \frac{\tau_l'(0)}{\alpha_l^n} \right\} \\ &\doteq -n \zeta_0'(0) + \tau_0'(0) \quad \text{as } n \rightarrow \infty. \end{aligned}$$

Hence we know that  $\mu \stackrel{\text{def}}{=} i \cdot \zeta_0'(0)$  is a real number. Similarly, by deriving

$$(11) \quad \begin{aligned} \text{Var}(Y_1 + \dots + Y_n) &= E\{(Y_1 + \dots + Y_n)^2\} - (E\{Y_1 + \dots + Y_n\})^2 \\ &= -\varphi_n''(0) - (-i\varphi_n'(0))^2 \\ &\doteq n(\zeta_0''(0) - \zeta_0'(0)^2) + \tau_0'(0)^2 - \tau_0''(0) \end{aligned}$$

as  $n \rightarrow \infty$ , we know that  $\zeta_0''(0) + \mu^2$  is a non-negative number. Now, we shall consider the characteristic function  $\psi_n(\theta)$  of the random variable  $(Y_1 + \dots + Y_n - n\mu)/\sqrt{n}$ . By (7), we have

$$(12) \quad \begin{aligned} \psi_n(\theta) &= e^{-i\sqrt{n}\mu\theta} \varphi_n\left(\frac{\theta}{\sqrt{n}}\right) \\ &= \frac{1}{\{e^{i\mu\theta/\sqrt{n}} \zeta_0(\theta/\sqrt{n})\}^n} \left\{ \tau_0\left(\frac{\theta}{\sqrt{n}}\right) + \sum_{l=1}^{r-1} \left(\frac{\zeta_0(\theta/\sqrt{n})}{\zeta_l(\theta/\sqrt{n})}\right)^n \tau_l\left(\frac{\theta}{\sqrt{n}}\right) \right\} \end{aligned}$$

For any fixed  $\theta$ , we have  $|\theta/\sqrt{n}| < \theta_0$  for all sufficiently large  $n$ , so that it follows from (5) that

$$\left| \frac{\zeta_0(\theta/\sqrt{n})}{\zeta_l(\theta/\sqrt{n})} \right| < \rho < 1 \quad (l=1, 2, \dots, r-1)$$

and so

$$(13) \quad \left( \frac{\zeta_0(\theta/\sqrt{n})}{\zeta_l(\theta/\sqrt{n})} \right)^n \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

On the other hand, we have

$$\begin{aligned} \zeta_0(\theta) &= \zeta_0(0) + \zeta_0'(0)\theta + \frac{\zeta_0''(0)}{2}\theta^2 + o(\theta^2) \\ &= 1 - i\mu\theta + \frac{\zeta_0''(0)}{2}\theta^2 + o(\theta^2) \quad \text{as } \theta \rightarrow 0, \end{aligned}$$

which implies for any fixed  $\theta$  that

$$\begin{aligned} &e^{i\mu\theta/\sqrt{n}} \zeta_0\left(\frac{\theta}{\sqrt{n}}\right) \\ &= \left(1 + i\mu\frac{\theta}{\sqrt{n}} - \frac{\mu^2}{2}\frac{\theta^2}{n} + o\left(\frac{1}{n}\right)\right) \cdot \left(1 - i\mu\frac{\theta}{\sqrt{n}} + \frac{\zeta_0''(0)}{2}\frac{\theta^2}{n} + o\left(\frac{1}{n}\right)\right) \\ &= 1 + \frac{\theta^2}{2n}(\zeta_0''(0) + \mu^2) + o\left(\frac{1}{n}\right) \end{aligned}$$

and

$$(14) \quad \left\{ e^{i\mu\theta/\sqrt{n}} \zeta_0\left(\frac{\theta}{\sqrt{n}}\right) \right\}^n \rightarrow e^{(\zeta_0''(0) + \mu^2)\theta^2/2} \quad \text{as } n \rightarrow \infty.$$

Hence we have by (8), (12), (13) and (14) that

$$(15) \quad \psi_n(\theta) \rightarrow e^{-\zeta_0''(0) + \mu^2)\theta^2/2} \quad \text{as } n \rightarrow \infty,$$

which proves our theorem.

3. If it holds that  $Y_n \geq 0$  (a. s.) and  $Y_1 + Y_2 + \cdots + Y_n \rightarrow \infty$  (a. s.), we can define a random variable  $N(t)$  for every positive number  $t$  such that

$$(16) \quad Y_1 + \cdots + Y_{N(t)} < t \leq Y_1 + \cdots + Y_{N(t)+1}.$$

Noticing that  $P\{N(t) < n\} = P\{S_n > t\}$ , we have immediately the following corollary, which gives a property of renewal type. (See the proof of Theorem 4.2. in [3])

*COROLLARY. Under the same assumptions of the foregoing theorem,  $(N(t) - t/\mu)/\sqrt{t}$  converges in distribution to a normal distribution as  $t \rightarrow \infty$ .*

## REFERENCES

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