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§0. Introduction

The purpose of the present paper is to give a central limit theorem for subadditive process in the sense of J. F. C. Kingman (cf. [3], [4]).

Throughout this article (Ω, \mathcal{B}, P) denotes a probability space on which all random variables are defined. Let T be a measure preserving transformation in what follows. According to Kingman, a family $(x_{s,t}; s < t, s=0, 1, \dots, t=1, 2, \dots)$ of random variables is called a subadditive process, if the following three conditions S_1, S_2 and S_3 are satisfied.

$$S_1. \quad x_{s,u} \leq x_{s,t} + x_{t,u} \text{ for all } s < t < u.$$

$$S_2. \quad x_{s+1,t+1}(\omega) = x_{s,t}(T\omega) \text{ for all } s < t.$$

$S_3.$ The expectation $E(x_{0,t})$ exists and satisfies

$$E(x_{0,t}) \geq -At,$$

for all $t \geq 1$ with some constant A .

In §1 and §2 we show that the random variable

$$t^{-\frac{1}{2}}(x_{0,t} - E(x_{0,t}))$$

has an asymptotically normal distribution under suitable conditions. J. F. C. Kingman and D. L. Burkholder proved the decomposition of

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subadditive processes into additive ones (cf. [3], [4]). Hence we can reduce our problem to the central limit theorem for stationary sequences given in [2].

In §3 we treat an application to products of positive random matrices. H. Furstenberg and H. Kesten have already obtained a central limit theorem for those (cf. [1]). By using the result in §1, we can weaken their conditions on moments and "weak dependence".

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§1. Notations and Results

In order to state our results, we shall introduce some notations. Set $g_t = E(x_{0,t})$ and $\gamma = \lim_{t \rightarrow \infty} g_t/t$ (cf. [4], p. 883). Let $\{\mathcal{M}_a^b; a \leq b, a=0, 1, \dots, \infty, b=0, 1, \dots, \infty\}$ be a family of sub- σ -fields of \mathcal{B} satisfying the following two conditions \mathbf{P}_1 and \mathbf{P}_2 :

\mathbf{P}_1 . If $a \leq c \leq d \leq b$, then $\mathcal{M}_c^d \subset \mathcal{M}_a^b$.

\mathbf{P}_2 . For all $a \leq b$, $T^{-1}\mathcal{M}_a^b = \mathcal{M}_{a+1}^{b+1}$.

We define $\phi(n)$ and $\alpha(n)$ by

$$(1.1) \quad \phi(n) = \sup_{k \geq 0} \sup \left\{ \frac{|P(A \cap B) - P(A)P(B)|}{P(A)}; A \in \mathcal{M}_0^k, B \in \mathcal{M}_{k+n}^\infty, P(A) \neq 0 \right\},$$

and

$$(1.2) \quad \alpha(n) = \sup_{k \geq 0} \sup \{ |P(A \cap B) - P(A)P(B)|; A \in \mathcal{M}_0^k, B \in \mathcal{M}_{k+n}^\infty \}.$$

Let us define $\Phi_\sigma(z)$ by

$$\Phi_\sigma(z) = \frac{1}{\sqrt{2\pi\sigma}} \int_{-\infty}^z \exp[-t^2/2\sigma^2] dt$$

for $\sigma > 0$ and

$$\Phi_0(z) = \begin{cases} 1, & (z > 0), \\ 0, & (z \leq 0). \end{cases}$$

The notation $\|*\|_\theta$ ($\theta \geq 1$) stands for $[E|*|^\theta]^{1/\theta}$.

We are now in a position to state our results.

Theorem 1. *Let $(x_{s,t})$ be a subadditive process and $\{\mathcal{M}_n^b\}$ be a family of sub- σ -fields of \mathcal{B} satisfying \mathbf{P}_1 and \mathbf{P}_2 . Suppose that the following four conditions are satisfied:*

(1) $(g_t - t\gamma)/\sqrt{t} \rightarrow 0$, as $t \rightarrow \infty$,

(2) $\sum_{n=1}^{\infty} [\phi(n)]^{\frac{1}{2}} < +\infty$,

(3) *there exists a random variable Ψ such that $E|\Psi|^2 < +\infty$ and*

$$|x_{0,t} - x_{1,t}| \leq \Psi \quad (\text{a.e.})$$

for all $t \geq 1$,

(4) $\sum_{n=1}^{\infty} \sup_{t \geq 1} \|(x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^b)\|_2 < +\infty$.

Then we have, for some $\sigma \geq 0$,

$$\lim_{t \rightarrow \infty} P\left\{ \frac{1}{\sqrt{t}}(x_{0,t} - g_t) < z \right\} = \Phi_\sigma(z)$$

at every continuity point of $\Phi_\sigma(z)$.

If we assume a stronger condition than (3) of Theorem 1, we can weaken the conditions (2) and (4); namely we have the following remarks.

Remark 1. The conclusion of Theorem 1 remains valid if the conditions (2)–(4) of Theorem 1 are replaced by

(2') $\sum_{n=1}^{\infty} [\alpha(n)]^{\delta/(2+\delta)} < +\infty$ for some $\delta > 0$,

(3') *there exists a random variable Ψ such that $E|\Psi|^{2+\delta} < +\infty$ and*

$$|x_{0,t} - x_{1,t}| \leq \Psi \quad (\text{a.e.})$$

for all $t \geq 1$,

$$(4') \quad \sum_{n=1}^{\infty} \sup_{t \geq 1} \|(x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n)\|_{\theta} < +\infty,$$

where $\theta = (2 + \delta)/(1 + \delta)$.

Remark 2. The conclusion of Theorem 1 remains valid if the conditions (2)–(4) of Theorem 1 are replaced by

$$(2'') \quad \sum_{n=1}^{\infty} \alpha(n) < +\infty,$$

(3'') there exists an essentially bounded random variable Ψ such that

$$|x_{0,t} - x_{1,t}| \leq \Psi \quad (\text{a.e.})$$

for all $t \geq 1$,

$$(4'') \quad \sum_{n=1}^{\infty} \sup_{t \geq 1} \|(x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n)\|_1 < +\infty.$$

§2. Proof of Theorem 1

Our result can be easily obtained by combining the following two theorems. The first theorem is proved by J. F. C. Kingman and D. L. Burkholder (cf. [3], [4]). The second is a central limit theorem for stationary sequences of weakly dependent case, which can be found in [2].

Theorem A. *There is a sequence $n_1 < n_2 < \dots$ of positive integers and an integrable random variable y such that*

$$(2.1) \quad y = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} (x_{0,t} - x_{1,t}) \quad (\text{a.e.}).$$

Moreover we have $Ey = \gamma$ and

$$(2.2) \quad \sum_{k=s}^{t-1} y(T^k \omega) \leq x_{s,t}(\omega).$$

Theorem B (Theorem 18.6.1 in [2]). *Let y be a random variable and T be a measure preserving transformation. Let $\{\mathcal{M}_a^b\}$ be a family of sub- σ -fields satisfying \mathbf{P}_1 and \mathbf{P}_2 . Suppose that*

- (1) $\sum_{n=1}^{\infty} [\phi(n)]^2 < +\infty,$
 (2) $E|y|^2 < +\infty,$
 (3) $\sum_{n=1}^{\infty} \|y - E(y|\mathcal{M}_0^n)\|_2 < +\infty.$

Then there exists a nonnegative constant σ and we have

$$\lim_{n \rightarrow \infty} P \left\{ \frac{1}{\sqrt{n}} \left(\sum_{k=0}^{n-1} y(T^k \omega) - nE(y) \right) < z \right\} = \Phi_{\sigma}(z)$$

at every continuity point of $\Phi_{\sigma}(z)$.

Proof of Theorem 1. By virtue of Theorem A, we can get

$$\begin{aligned} & E \left| \frac{1}{\sqrt{t}} (x_{0,t}(\omega) - g_t) - \frac{1}{\sqrt{t}} \left(\sum_{k=0}^{t-1} y(T^k \omega) - t\gamma \right) \right| \\ & \leq \frac{1}{\sqrt{t}} (g_t - t\gamma) + E \left| \frac{1}{\sqrt{t}} (x_{0,t}(\omega) - \sum_{k=0}^{t-1} y(T^k \omega)) \right| = \frac{2}{\sqrt{t}} (g_t - t\gamma). \end{aligned}$$

On the other hand, from the condition (1) of Theorem 1, we have

$$\frac{2}{\sqrt{t}} (g_t - t\gamma) \longrightarrow 0$$

as $t \rightarrow \infty$. Hence our theorem can be deduced from the central limit theorem for the stationary sequence $\{y(T^k \omega); k=0, 1, \dots\}$.

The condition (2) of Theorem 1 immediately implies (1) of Theorem B, so we only need to check that y satisfies the conditions (2) and (3) of Theorem B. Using the condition (3) of Theorem 1 and (2.1), we have

$$E|y|^2 \leq E \left(\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} |x_{0,t} - x_{1,t}| \right)^2 \leq E|\Psi|^2 < +\infty,$$

which implies the condition (2) of Theorem B.

Moreover the condition (3) of Theorem 1 guarantees that

$$(2.3) \quad E(y|\mathcal{M}_0^n) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} E(x_{0,t} - x_{1,t}|\mathcal{M}_0^n)$$

and

$$(2.4) \quad \left| \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} \{ (x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n) \} \right| \\ \leq \Psi + E(\Psi | \mathcal{M}_0^n).$$

Therefore we obtain, from (2.3) and (2.4),

$$\|y - E(y | \mathcal{M}_0^n)\|_2 \\ = \lim_{N \rightarrow \infty} \left\| \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} \{ (x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n) \} \right\|_2 \\ (2.5) \quad \leq \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \frac{1}{n_j} \sum_{t=1}^{n_j} \| (x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n) \|_2 \\ \leq \sup_{t \geq 1} \| (x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n) \|_2.$$

Combining (2.5) and the condition (4) of Theorem 1, we get that the condition (3) of Theorem B holds for y . The proof is therefore complete.

Remark 3. Remarks 1 and 2 in §1 can be deduced from Theorems 18.6.2 and 18.6.3 in [2], instead of Theorem B, respectively. The proof is exactly similar to the proof of Theorem 1.

§3. Application to Products of Random Matrices

Let Z^1, Z^2, \dots be a stationary sequence of random $k \times k$ matrices with strictly positive elements. By the stationarity there is a measure preserving transformation T such that

$$Z^{i+1}(\omega) = Z^i(T\omega) \quad (i=1, 2, \dots).$$

Let us define a family of sub- σ -fields as follows:

$$\mathcal{M}_a^b = \sigma\{Z^{a+1}, Z^{a+2}, \dots, Z^{b+1}\},$$

where the notation $\sigma\{*\}$ stands for the σ -field generated by $*$. Clearly, this $\{\mathcal{M}_a^b\}$ satisfies \mathbf{P}_1 and \mathbf{P}_2 with respect to T . Associated with this

family, let $\phi(n)$ and $\alpha(n)$ be the quantities defined by (1.1) and (1.2) respectively. We denote by $(A)_{i,j}$ the (i, j) -th element of a matrix A . We write

$${}^t Y^s = Z^t \cdot Z^{t-1} \dots Z^s.$$

Under these notations, we can obtain the following

Theorem 2. *Suppose that $(Z^i; i=1, 2, \dots)$ is a stationary sequence of random matrices and the elements of these matrices are strictly positive. Suppose also that*

(1) *there exists a positive constant C such that*

$$1 \leq \max_{i,j} (Z^1)_{i,j} / \min_{i,j} (Z^1)_{i,j} \leq C \quad (\text{a.e.}),$$

(2) $\sum_{n=1}^{\infty} [\phi(n)]^{\frac{1}{2}} < +\infty,$

and

(3) $E|\log(Z^1)_{1,1}|^2 < +\infty.$

Then there is a nonnegative constant σ such that, for all $1 \leq i, j \leq k,$

$$\lim_{t \rightarrow \infty} P \left\{ \frac{1}{\sqrt{t}} [\log ({}^t Y^1)_{i,j} - E(\log ({}^t Y^1)_{i,j})] < z \right\} = \Phi_{\sigma}(z)$$

at every continuity point of $\Phi_{\sigma}(z).$

Proof. Combining (1) and (3) of Theorem 2, we have

$$E(|\log(Z^1)_{i,j}|) < +\infty$$

for all $1 \leq i, j \leq k.$ This fact and the positivity of elements show that the family of random variables

$$(3.1) \quad x_{s,t} = -\log ({}^t Y^{s+1})_{1,1}$$

is a subadditive process (cf. [4], pp. 891-892).

First, we prove the asymptotic normality of the random variables

$$\frac{1}{\sqrt{t}} \{x_{0,t} - E(x_{0,t})\},$$

using Theorem 1.

(a) Under the condition (1) of Theorem 2, it is already shown in [1] that

$$g_{t+1} - g_t = \gamma + O((1 - C^{-3})^t)$$

([1], p. 467), where $g_t = E(x_{0,t})$ and $\gamma = \lim_{t \rightarrow \infty} g_t/t$ as in §1. This fact allows us to deduce the condition (1) of Theorem 1.

(b) The condition (2) of Theorem 2 immediately implies (2) of Theorem 1.

(c) The obvious equality

$$\frac{({}^t Y^1)_{1,1}}{({}^t Y^2)_{1,1}} = \frac{\sum_{i=1}^k \sum_{j=1}^k ({}^t Y^3)_{1,i} (Z^2)_{i,j} (Z^1)_{j,1}}{\sum_{i=1}^k ({}^t Y^3)_{1,i} (Z^2)_{i,1}}$$

and the condition (1) of Theorem 2 show that the following evaluations

$$C^{-1} \sum_{j=1}^k (Z^1)_{j,1} \leq \frac{({}^t Y^1)_{1,1}}{({}^t Y^2)_{1,1}} \leq C \sum_{j=1}^k (Z^1)_{j,1}.$$

Hence we have

$$|x_{0,t} - x_{1,t}| \leq |\log(Z^1)_{1,1}| + \log kC^2.$$

This inequality and the condition (3) of Theorem 2 allow us to derive (3) of Theorem 1.

(d) If A, B and C are $k \times k$ matrices and if the notation $(A)_{*,j}$ stands for $\sum_{i=1}^k (A)_{i,j}$, then the equality

$$(3.2) \quad \frac{(ABC)_{1,1}}{(AB)_{1,1}} = \frac{\sum_{i=1}^k (B)_{*,i} (C)_{i,1}}{\sum_{i=1}^k (B)_{*,1}} + \frac{\sum_{i=1}^k \sum_{j=1}^k (A)_{1,j} \left[\frac{(B)_{j,i}}{(B)_{*,i}} - \frac{(B)_{j,1}}{(B)_{*,1}} \right] (B)_{*,i} (C)_{i,1}}{\sum_{j=1}^k (A)_{1,j} (B)_{j,1}}$$

can be easily shown. Now take $A = {}^{m+n} Y^{n+1}$, $B = {}^n Y^2$ and $C = Z^1$. With these substitutions we have

$$(3.3) \quad \frac{(m+nY^1)_{1,1}}{(m+nY^2)_{1,1}} = \sum_{i=1}^k \frac{({}^n Y^2)_{*,i}}{({}^n Y^2)_{*,1}} (Z^1)_{1,1} \\ + \frac{\sum_{i=1}^k \sum_{j=1}^k (m+nY^{n+1})_{1,j} \left[\frac{({}^n Y^2)_{j,i}}{({}^n Y^2)_{*,i}} - \frac{({}^n Y^2)_{j,1}}{({}^n Y^2)_{*,1}} \right] ({}^n Y^2)_{*,i} (Z^1)_{i,1}}{\sum_{j=1}^k (m+nY^{n+1})_{1,j} ({}^n Y^2)_{j,1}}.$$

We denote by α the first term in the right hand side of (3.3) and by β the second term.

The obvious identity

$$\frac{(m+nY^m)_{i_1, j_1}}{(m+nY^m)_{i_2, j_2}} = \frac{\sum_{r,s} (Z^{m+n})_{i_1, r} ({}^{m+n-1} Y^{m+1})_{r, s} (Z^m)_{s, j_1}}{\sum_{r,s} (Z^{m+n})_{i_2, r} ({}^{m+n-1} Y^{m+1})_{r, s} (Z^m)_{s, j_2}}$$

and the condition (1) of Theorem 2 allow us to derive

$$(3.4) \quad C^{-2} \leq (m+nY^m)_{i_1, j_1} / (m+nY^m)_{i_2, j_2} \leq C^2,$$

as was pointed out in [1] (cf. [1], Lemma 2). From (3.4) it clearly follows that

$$C^{-2} \leq ({}^n Y^2)_{*,i} / ({}^n Y^2)_{*,1} \leq C^2.$$

Therefore, using the condition (1) of Theorem 2, we have

$$(3.5) \quad kC^{-3}(Z^1)_{1,1} \leq \alpha \leq kC^3(Z^1)_{1,1}.$$

To deal with β , we derive

$$(3.6) \quad \left| \frac{({}^n Y^2)_{j,i}}{({}^n Y^2)_{*,i}} - \frac{({}^n Y^2)_{j,1}}{({}^n Y^2)_{*,1}} \right| \leq (1 - C^{-3})^{n-2}$$

from (3.4) and the condition (1) of Theorem 2. The proof of (3.6) is all the same as the proof of Lemma 3 in [1], p. 463, so we omit it. Inequality (3.4) guarantees that

$$(3.7) \quad \max_i ({}^n Y^2)_{*,i} / \min_j ({}^n Y^2)_{j,1} \leq kC^2.$$

Combining (3.6) and (3.7) allows us to deduce that

$$(3.8) \quad \beta \leq kC^2(1 - C^{-3})^{n-2} \sum_{i=1}^k (Z^1)_{i,1} \leq k^2 C^3 (1 - C^{-3})^{n-2} (Z^1)_{1,1}.$$

Using (3·5) and (3·8), we can get

$$\beta/\alpha \leq kC^6(1-C^{-3})^{n-2}.$$

Hence, from the identity

$$\log(\alpha + \beta) = \log \alpha + \log(1 + (\beta/\alpha)) = \log \alpha + O(\beta/\alpha)$$

as $\beta/\alpha \rightarrow 0$, we have

$$(3\cdot9) \quad \begin{aligned} x_{0,m+n} - x_{1,m+n} &= -\log [(^{m+n}Y^1)_{1,1}/(^{m+n}Y^2)_{1,1}] \\ &= -\log \alpha + O((1-C^{-3})^n) \end{aligned}$$

uniformly in m and ω .

Since $\log \alpha$ is \mathcal{M}_0^n -measurable, we can easily obtain

$$\begin{aligned} & |(x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n)| \\ &= \begin{cases} 0, & (t \leq n), \\ O((1-C^{-3})^n), & (t > n), \end{cases} \end{aligned}$$

uniformly in t and ω . This shows that the condition (4) of Theorem 1 holds. Therefore we can apply Theorem 1 to the random variables $x_{0,t}$ defined by (3·1). Consequently, we get the result of Theorem 2 for $i=j=1$.

To deal with the other values of (i, j) , note first that the inequality

$$|\log(^tY^1)_{i,j} - \log(^tY^1)_{1,1}| \leq 2 \log C$$

follows from (3·4), as was pointed out in [1]. Therefore random variables

$$\frac{1}{\sqrt{t}} \{\log(^tY^1)_{1,1} - E(\log(^tY^1)_{1,1})\}$$

and

$$\frac{1}{\sqrt{t}} \{\log(^tY^1)_{i,j} - E(\log(^tY^1)_{i,j})\}, \quad (1 \leq i, j \leq k),$$

have an asymptotically same distribution. This completes the proof.

By using Remarks 1 and 2 in §1 instead of Theorem 1, we can get the following Remarks 4 and 5 respectively. Their proofs can be copied from the previous one.

Remark 4. The conclusion of Theorem 2 remains valid if the conditions (2) and (3) of Theorem 2 are replaced by

$$(2') \quad \sum_{n=1}^{\infty} [\alpha(n)]^{\frac{\delta}{2+\delta}} < +\infty \text{ for some } \delta > 0,$$

$$(3') \quad E|\log(Z^1)_{1,1}|^{2+\delta} < +\infty.$$

Remark 5. The conclusion of Theorem 2 remains valid again if the conditions (2) and (3) of Theorem 2 are replaced by

$$(2'') \quad \sum_{n=1}^{\infty} \alpha(n) < +\infty,$$

$$(3'') \quad \log(Z^1)_{1,1} \text{ is an essentially bounded random variable.}$$

Remark 6. It is easy to see that the assumption AII in [1], p. 464, is stronger than our condition (2) of Theorem 2, so that our results cover Furstenberg and Kesten's central limit theorem in [1].

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