

# A Central Limit Theorem for the Subadditive Process and Its Application to Products of Random Matrices

By

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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  $(\Omega, \mathcal{B}, P)$  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- $\sigma$ -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

$$(1 \cdot 1) \quad \phi(n) = \sup_{k \geq 0} \sup \left\{ \frac{|P(A \cap B) - P(A)P(B)|}{P(A)}; A \in \mathcal{M}_0^k, B \in \mathcal{M}_{k+n}^\infty, P(A) \neq 0 \right\},$$

and

$$(1 \cdot 2) \quad \alpha(n) = \sup_{k \geq 0} \sup \{ |P(A \cap B) - P(A)P(B)|; A \in \mathcal{M}_0^k, B \in \mathcal{M}_{k+n}^\infty \}.$$

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  $\{\mathcal{M}_a^b\}$  be a family of sub- $\sigma$ -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  $\Psi$  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)  $\sum_{n=1}^\infty \sup_{t \geq 1} \|(x_{0,t} - x_{1,t}) - E(x_{0,t} - x_{1,t} | \mathcal{M}_0^n)\|_2 < +\infty$ .

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  $\Psi$  such that  $E|\Psi|^{2+\delta} < +\infty$  and

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

for all  $t \geq 1$ ,

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

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  $\Psi$  such that

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

for all  $t \geq 1$ ,

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

## §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- $\sigma$ -fields satisfying  $\mathbf{P}_1$  and  $\mathbf{P}_2$ . Suppose that*

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

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

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

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

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

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

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

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

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- $\sigma$ -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  $\sigma$ -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  $\sigma$  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

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

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

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

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

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

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



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

We denote by  $\alpha$  the first term in the right hand side of (3.3) and by  $\beta$  the second term.

The obvious identity

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

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

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

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  $\beta$ , we derive

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

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

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

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

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

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

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

uniformly in  $m$  and  $\omega$ .

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  $\omega$ . 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

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

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

**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].

### References

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