

**ON A CENTRAL LIMIT THEOREM FOR
A STATIONARY MULTIVARIATE LINEAR
PROCESS GENERATED BY LINEARLY POSITIVE
QUADRANT DEPENDENT RANDOM VECTORS**

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ABSTRACT. For a stationary multivariate linear process of the form $\mathbb{X}_t = \sum_{j=0}^{\infty} A_j \mathbb{Z}_{t-j}$, where $\{\mathbb{Z}_t : t = 0, \pm 1, \pm 2, \dots\}$ is a sequence of stationary linearly positive quadrant dependent m -dimensional random vectors with $E(\mathbb{Z}_t) = \mathbb{0}$ and $E\|\mathbb{Z}_t\|^2 < \infty$, we prove a central limit theorem.

1. Introduction

Lehmann [8] introduced a simple and natural definition of positive dependence : A sequence $\{Y_t : t = 0, 1, 2, \dots\}$ of random variables is said to be pairwise positive quadrant dependent (pairwise PQD) if for any real α_i, α_j and $i \neq j$ $P\{Y_i > \alpha_i, Y_j > \alpha_j\} \geq P\{Y_i > \alpha_i\}P\{Y_j > \alpha_j\}$. A concept stronger than PQD was introduced by Newman [10]: A sequence $\{Y_t\}$ of random variables is said to be linearly positive quadrant dependent (LPQD) if for any disjoint A, B and positive r_j 's, $\sum_{i \in A} r_i Y_i$ and

$\sum_{j \in B} r_j Y_j$ are PQD.

Two m -variate random vectors $\mathbb{Z}_1, \mathbb{Z}_2$ are said to be positive quadrant dependent (PQD) if Z_{1i}, Z_{2j} are PQD for all $i, j = 1, \dots, m$, where Z_{1i}, Z_{2j} are components of $\mathbb{Z}_1, \mathbb{Z}_2$, respectively.

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Let (Z_1, Z_2, \dots, Z_t) be m -variate random vectors. We say that (Z_1, Z_2, \dots, Z_t) is linearly positive quadrant dependent if for any disjoint $A, B \subset \{1, \dots, t\}$ and for any real vectors a_r with nonnegative components,

$$(1) \quad \sum_{s \in A} a_s Z_s \text{ and } \sum_{r \in B} a_r Z_r \text{ are PQD.}$$

Let $\{X_t, t = 0, \pm 1, \dots\}$ be an m -variate linear process of the form

$$(2) \quad X_t = \sum_{u=0}^{\infty} A_u Z_{t-u}$$

defined on a probability space (Ω, \mathcal{F}, P) , where $\{Z_t\}$ is a sequence of stationary m -variate LPQD random vectors with $EZ_t = \mathbb{O}$, $E\|Z_t\|^2 < \infty$ and positive definite covariance matrix $\Gamma_{m \times m}$. Throughout this paper we shall assume that

$$(3) \quad \sum_{u=0}^{\infty} \|A_u\| < \infty \text{ and } \sum_{u=0}^{\infty} A_u \neq \mathbb{O}_{m \times m},$$

where for any $m \times m$, $m \geq 1$, matrix $A = (a_{ij})$, $\|A\| = \sum_{i=1}^m \sum_{j=1}^m |a_{ij}|$ and

$\mathbb{O}_{m \times m}$ denotes the $m \times m$ zero matrix. Further, let

$$T = \left(\sum_{j=0}^{\infty} A_j \right) \Gamma \left(\sum_{j=0}^{\infty} A_j \right)',$$

where the prime denotes transpose, and the matrix $\Gamma = [\sigma_{kj}]$ with

$$(4) \quad \sigma_{kj} = E(Z_{1k}Z_{1j}) + \sum_{t=2}^{\infty} (E(Z_{1k}Z_{tj}) + E(Z_{1j}Z_{tk})).$$

Further, let $S_n = \sum_{t=1}^n X_t$, ($n \geq 0$; $S_0 = \mathbb{O}$).

Fakhre-Zakeri and Lee [4] proved a central limit theorem for multivariate linear processes generated by independent multivariate random vectors and Fakhre-Zakeri and Lee [5] also derived a functional central limit theorem for multivariate linear processes generated by multivariate random vectors with martingale difference sequence.

In this note we prove a central limit theorem for an m -variate linear process generated by m -variate LPQD random vectors.

THEOREM 1.1. *Let $\{Z_t, t = 0, \pm 1, \dots\}$ be a strictly stationary LPQD sequence of m -dimensional random vectors with $E(Z_t) = \mathbb{O}$, $E\|Z_t\|^2 < \infty$ and positive definite covariance matrix Γ as in (4). Let $\{X_t\}$ be an m -variate linear process defined as in (2). Assume that*

$$(5) \quad E\|Z_1\|^2 + 2 \sum_{t=2}^{\infty} \sum_{i=1}^m E(Z_{1i}Z_{ti}) = \sigma^2 < \infty,$$

$$(6) \quad \sum_{t=n+1}^{\infty} E\|Z_{1i}Z_{ti}\| = O(n^{-\rho}) \quad \text{for some } \rho > 0,$$

and

$$(7) \quad E\|Z_t\|^s < \infty \quad \text{for some } s > 2.$$

Then, the multivariate linear process $\{X_t\}$ fulfills the central limit theorem, that is, $n^{-\frac{1}{2}}S_n \xrightarrow{\mathcal{D}} N(\mathbb{O}, T)$.

REMARK. For $m = 1$, Kim and Baek [7] showed that the central limit theorem holds for the linear processes generated by an LPQD process.

2. Proofs

Note that Newman [10] has proved the central limit theorem for LPQD random variables (see Theorem 12 of [10]). Thus by means of the simple device due to Cramer Wold the following result holds.

LEMMA 2.1. *Let $\{Z_t\}$ be a sequence of stationary LPQD m -variate random vectors with $E(Z_t) = \mathbb{O}$ and $E\|Z_t\|^2 < \infty$. If (5) holds then*

$$n^{-\frac{1}{2}} \sum_{t=1}^n Z_t \xrightarrow{\mathcal{D}} N(\mathbb{O}, \Gamma),$$

where $\Gamma = [\sigma_{kj}]$ is defined as in (4); that is, $\{Z_t\}$ satisfies the central limit theorem.

LEMMA 2.2. *Let $\{Z_t\}$ be a sequence of stationary LPQD random vectors with $E(Z_t) = \mathbb{O}$, $E\|Z_t\|^2 < \infty$. Let $\tilde{X}_t = (\sum_{j=0}^{\infty} A_j)Z_t$ and $\tilde{S}_k =$*

$\sum_{t=1}^k \tilde{\mathbb{X}}_t$. Assume that (5), (6) and (7) hold. Then

$$(8) \quad n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|\tilde{\mathbb{S}}_k - \mathbb{S}_k\| = o_p(1).$$

Proof. See Appendix. \square

Proof of Theorem 1.1. As in Lemma 2.2, set $\tilde{\mathbb{X}}_t = (\sum_{j=0}^{\infty} A_j)Z_t$ and

$\tilde{\mathbb{S}}_n = \sum_{t=1}^n \tilde{\mathbb{X}}_t$. First note that

$$(9) \quad \begin{aligned} & E\|\tilde{\mathbb{X}}_1\|^2 + 2 \sum_{t=2}^{\infty} \sum_{i=1}^m E(\tilde{X}_{1i}\tilde{X}_{ti}) \\ &= \left(\sum_{j=1}^{\infty} A_j\right)^2 (E\|Z_1\|^2 + 2 \sum_{t=2}^{\infty} \sum_{i=1}^m E(Z_{1i}Z_{ti})). \end{aligned}$$

Since $\tilde{\mathbb{X}}_t$ is LPQD, by Lemma 2.1 $\{\tilde{\mathbb{X}}_t\}$ satisfies the central limit theorem, that is,

$$(10) \quad n^{-\frac{1}{2}} \tilde{\mathbb{S}}_n \xrightarrow{\mathcal{D}} N(\mathbb{O}, T).$$

According to Lemma 2.2 we also have

$$(11) \quad n^{-\frac{1}{2}} |\tilde{\mathbb{S}}_n - \mathbb{S}_n| = o_p(1).$$

Hence from (10) and (11) the desired conclusion follows by Theorem 4.1 of [1]. \square

Appendix

Proof of Lemma 2.2. We prove Lemma 2.2 using the ideas in the proof of Lemma 3 of [5] and Lemma 2 of [7]. First observe that

$$(A.1) \quad \sum_{t=n+1}^{\infty} E(\tilde{X}_{1i}\tilde{X}_{ti}) = \left(\sum_{j=0}^{\infty} A_j\right)^2 \sum_{t=n+1}^{\infty} \sum_{i=1}^m E(Z_{1i}Z_{ti}) = O(n^{-\rho})$$

and that

$$(A.2) \quad E\|\tilde{\mathbb{X}}_t\|^s = \left(\sum_{j=0}^{\infty} A_j\right)^s E\|Z_t\|^s < \infty \quad \text{for some } s > 2.$$

By Lemma 3 of [7], it follows from (A.1) and (A.2) that

$$(A.3) \quad E(\max_{1 \leq k < n} \|\tilde{X}_1 + \dots + \tilde{X}_k\|^r) \geq Bn^{\frac{r}{2}}$$

for some $r > 2$ and a constant B .

Next, we observe that

$$\begin{aligned} \tilde{S}_k &= \sum_{t=1}^k \left(\sum_{j=0}^{k-t} A_j \right) Z_t + \sum_{t=1}^k \left(\sum_{j=k-t+1}^{\infty} A_j \right) Z_t \\ &= \sum_{t=1}^k \left(\sum_{j=0}^{t-1} A_j Z_{t-j} \right) + \sum_{t=1}^k \left(\sum_{j=k-t+1}^{\infty} A_j \right) Z_t \end{aligned}$$

and thus

$$\begin{aligned} \tilde{S}_k - S_k &= - \sum_{t=1}^k \sum_{j=t}^{\infty} A_j Z_{t-j} + \sum_{t=1}^k \left(\sum_{j=k-t+1}^{\infty} A_j \right) Z_t \\ &= I_1 + I_2 \text{ (say)}. \end{aligned}$$

To prove

$$(A.4) \quad n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|I_1\| \xrightarrow{P} 0,$$

we observe that for $r > 2$

$$\begin{aligned} &n^{-\frac{r}{2}} E \max_{1 \leq k \leq n} \left\| \sum_{t=1}^k \sum_{j=t}^{\infty} A_j Z_{t-j} \right\|^r \\ &= n^{-\frac{r}{2}} E \max_{1 \leq k \leq n} \left\| \sum_{j=1}^{\infty} \sum_{t=1}^{j \wedge k} A_j Z_{t-j} \right\|^r \\ &\leq n^{-\frac{r}{2}} \left(\sum_{j=1}^{\infty} \|A_j\| \left\{ E \max_{1 \leq k \leq n} \left\| \sum_{t=1}^{j \wedge k} Z_{t-j} \right\|^r \right\}^{\frac{1}{r}} \right)^r \\ &\leq K \left[\sum_{j=1}^{\infty} \|A_j\| \left(\frac{j \wedge k}{n} \right)^{\frac{1}{2}} \right]^r \end{aligned}$$

for some positive constant K , where we have used Lemma 2 in [7] for LPQD random variables. By the dominated convergence theorem the last term above tends to zero as $n \rightarrow \infty$ from which (A4) follows.

Next, we show that

$$(A.5) \quad n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|I_2\| = o_p(1).$$

Write

$$I_2 = I_{21} + I_{22}, \text{ where}$$

$$I_{21} = A_1 \mathbb{Z}_k + A_2(\mathbb{Z}_k + \mathbb{Z}_{k-1}) + \cdots + A_k(\mathbb{Z}_k + \cdots + \mathbb{Z}_1)$$

and

$$I_{22} = (A_{k+1} + A_{k+2} + \cdots) (\mathbb{Z}_k + \cdots + \mathbb{Z}_1).$$

Let p_n be a sequence of positive integers such that

$$(A.6) \quad p_n \rightarrow \infty \text{ and } p_n/n \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Note that

$$\begin{aligned} n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|I_{22}\| &\leq \left(\sum_{i=0}^{\infty} \|A_i\| \right) n^{-\frac{1}{2}} \max_{1 \leq k \leq p_n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_k\| \\ &\quad + \left(\sum_{i > p_n} \|A_i\| \right) n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_k\| \\ &= III + IV \text{ (say)}. \end{aligned}$$

It follows from (3) and (A.6) that for some $r > 2$

$$III \leq \left(\sum_{i=0}^{\infty} \|A_i\| \right)^r B_1 (p_n/n)^{\frac{r}{2}} \xrightarrow{P} 0$$

and

$$IV \leq \left(\sum_{i > p_n} \|A_i\| \right)^r B_2 \xrightarrow{P} 0,$$

by Lemma 2 of [7]. It remains to prove that

$$Y_n := n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|I_{21}\| = o_p(1).$$

To this end, define for each $l \geq 1$

$$I_{21,l} = B_1 \mathbb{Z}_k + B_2(\mathbb{Z}_k + \mathbb{Z}_{k+1}) + \cdots + B_k(\mathbb{Z}_k + \cdots + \mathbb{Z}_1),$$

where

$$B_k = \begin{cases} A_k, & k \leq l \\ \mathbb{O}_{m \times m}, & k > l. \end{cases}$$

Let $Y_{n,l} = n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \|I_{21,l}\|$. Clearly, for each $l \geq 1$,

$$(A.7) \quad Y_{n,l} = o_p(1).$$

On the other hand,

$$\begin{aligned} (Y_{n,l} - Y_n) &\leq n^{-\frac{1}{2}} \max_{1 \leq k \leq n} \left\| \sum_{i=1}^k (A_i - B_i) (\mathbb{Z}_k + \cdots + \mathbb{Z}_{k-i+1}) \right\| \\ &\leq n^{-\frac{1}{2}} \max_{l < k \leq n} \left(\sum_{i=l+1}^k \|A_i\| \max_{l < i \leq n} \|\mathbb{Z}_k + \cdots + \mathbb{Z}_{k-i+1}\| \right) \\ &\leq n^{-\frac{1}{2}} \sum_{i>l} \|A_i\| \max_{l < k \leq n} \max_{l < i \leq k} (\|\mathbb{Z}_1 + \cdots + \mathbb{Z}_k\| + \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_{k-i}\|) \\ &\leq n^{-\frac{1}{2}} \sum_{i>l} \|A_i\| \left(\max_{l < k \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_k\| \right. \\ &\quad \left. + \max_{l < k \leq n} \max_{l < i \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_{k-i}\| \right) \\ &\leq n^{-\frac{1}{2}} \sum_{i>l} \|A_i\| \left(\max_{1 \leq j \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_j\| + \max_{1 \leq k \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_j\| \right) \\ &= 2n^{-\frac{1}{2}} \sum_{i>l} \|A_i\| \max_{1 \leq j \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_j\|. \end{aligned}$$

From this result and Lemma 2 of [7], for any $\delta > 0$,

$$\begin{aligned} (A.8) \quad &\lim_{l \rightarrow \infty} \limsup_{n \rightarrow \infty} P(|Y_{n,l} - Y_n|^2 > \delta) \\ &\leq \lim_{l \rightarrow \infty} 2^r \delta^{-r} \left(\sum_{i>l} \|A_i\| \right)^r \lim_{n \rightarrow \infty} n^{-\frac{r}{2}} \max_{1 \leq j \leq n} \|\mathbb{Z}_1 + \cdots + \mathbb{Z}_j\|^r \\ &\leq B \lim_{l \rightarrow \infty} \delta^{-r} 2^r \left(\sum_{i>l} \|A_i\| \right)^r = 0. \end{aligned}$$

In view of (A.7) and (A.8), it follows from Theorem 4.2 of [1, p.25] that $Y_n = o_p(1)$. This completes the proof of Lemma 2.2. \square

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