

A CENTRAL LIMIT THEOREM FOR GENERAL WEIGHTED SUMS OF LPQD RANDOM VARIABLES AND ITS APPLICATION

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ABSTRACT. In this paper we derive the central limit theorem for $\sum_{i=1}^n a_{ni}\xi_i$, where $\{a_{ni}, 1 \leq i \leq n\}$ is a triangular array of non-negative numbers such that $\sup_n \sum_{i=1}^n a_{ni}^2 < \infty$, $\max_{1 \leq i \leq n} a_{ni} \rightarrow 0$ as $n \rightarrow \infty$ and ξ_i 's are a linearly positive quadrant dependent sequence. We also apply this result to consider a central limit theorem for a partial sum of a generalized linear process of the form $X_n = \sum_{j=-\infty}^{\infty} a_{k+j}\xi_j$.

1. Introduction and results

Lehmann [7] introduced a simple and natural definition of positive dependence: A sequence $\{\xi_i, 1 \leq i \leq n\}$ of random variables is said to be pairwise positive quadrant dependent (pairwise PQD) if for any real α_i, α_j and $i \neq j$ $P(\xi_i > \alpha_i, \xi_j > \alpha_j) \geq P(\xi_i > \alpha_i)P(\xi_j > \alpha_j)$. Much stronger concepts than PQD was considered by Esary, Proschan and Walkup [4]: A sequence $\{\xi_i, 1 \leq i \leq n\}$ of random variables is said to be associated if for any real coordinatewise increasing functions f, g on \mathbb{R}^n , $\text{Cov}(f(\xi_1, \dots, \xi_n), g(\xi_1, \dots, \xi_n)) \geq 0$.

Instead of association, Newman's [9] central limit theorem requires only that positive linear combinations of the random variables are PQD. The definition of positive dependence introduced by Newman [9] is the following: A sequence $\{\xi_i, 1 \leq i \leq n\}$ of random variables is said to be linearly positive quadrant dependent(LPQD) if for every pair of disjoint

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subsets $A, B \subset \{1, 2, \dots, n\}$ and positive r'_j s

$$(1.1) \quad \sum_{i \in A} r_i \xi_i \quad \text{and} \quad \sum_{j \in B} r_j \xi_j \quad \text{are PQD.}$$

Let us remark that LPQD is between pairwise PQD and association and it is well known (see, for example, Newman [9, p.131] that association implies LPQD and LPQD implies PQD.

Newman [9] established the central limit theorem for a strictly stationary LPQD process and Birkel [2] also obtained a functional central limit theorem for LPQD processes. Kim and Baek [6] extended this result to a stationary linear process of the form $Y_k = \sum_{j=0}^{\infty} a_j \xi_{k-j}$, where $\{a_j\}$ is a sequence of real numbers with $\sum_{j=0}^{\infty} |a_j| < \infty$ and $\{\xi_k\}$ is a strictly stationary LPQD process with $E\xi_i = 0$ and $0 < E\xi_i^2 < \infty$.

In this paper we derive a central limit theorem for a linearly positive quadrant dependent sequence in a double array, replacing the strictly stationarity assumption with uniform integrability (see Theorem 1.1 below). We apply this result to obtain a central limit theorem for a partial sum of a linear process $X_n = \sum_{j=-\infty}^{\infty} a_{k+j} \xi_j$ generated by linearly positive quadrant dependent sequence $\{\xi_j\}$ (see Theorem 1.2 below).

Newman [9] proved that the following central limit theorem for strictly stationary LPQD sequence:

THEOREM A. [9] *Let $\{\xi_i, i \geq 1\}$ be a strictly stationary sequence of LPQD random variables with $E\xi_i = 0$ and $E\xi_i^2 < \infty$. If we assume that*

$$\sigma^2 = \sum_{i=1}^{\infty} \text{Cov}(\xi_1, \xi_i) < \infty,$$

then

$$(\sigma^2 n)^{-\frac{1}{2}} \sum_{i=1}^n \xi_i \rightarrow^D N(0, 1) \text{ as } n \rightarrow \infty.$$

The next theorem extends Newman's central limit theorem for a strictly stationary LPQD sequence (Theorem A) from equal weights to general weights, while at the same time weakening the assumption of stationarity.

THEOREM 1.1. *Let $\{a_{ni}, 1 \leq i \leq n\}$ be a triangular array of non-negative numbers such that*

$$(1.2) \quad \sup_n \sum_{i=1}^n a_{ni}^2 < \infty,$$

and

$$(1.3) \quad \max_{1 \leq i \leq n} a_{ni} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Let $\{\xi_i\}$ be a centered sequence of linearly positive quadrant dependent random variables such that

$$(1.4) \quad \{\xi_i^2\} \text{ is an uniformly integrable family,}$$

$$(1.5) \quad \text{Var}\left(\sum_{i=1}^n a_{ni}\xi_i\right) = 1$$

and

$$(1.6) \quad \sum_{j:|k-j|\geq u} \text{Cov}(\xi_k, \xi_j) \rightarrow 0 \text{ as } u \rightarrow \infty \text{ uniformly in } k \geq 1$$

(see Cox and Grimmet [3]). Then

$$\sum_{i=1}^n a_{ni}\xi_i \xrightarrow{\mathcal{D}} N(0, 1) \text{ as } n \rightarrow \infty.$$

COROLLARY 1.1. Let $\{\xi_i\}$ be a centered sequence of linearly positive quadrant dependent random variables such that $\{\xi_i^2\}$ is an uniformly integrable family and let $\{a_{ni}, 1 \leq i \leq n\}$ be a triangular array of nonnegative numbers such that

$$(1.7) \quad \sup_n \sum_{i=1}^n \frac{a_{ni}^2}{\sigma_n^2} < \infty,$$

$$(1.8) \quad \max_{1 \leq i \leq n} \frac{a_{ni}}{\sigma_n} \rightarrow 0 \text{ as } n \rightarrow \infty,$$

where $\sigma_n^2 = \text{Var}(\sum_{i=1}^n a_{ni}\xi_i)$. If (1.6) holds then, as $n \rightarrow \infty$

$$(1.9) \quad \frac{1}{\sigma_n} \sum_{i=1}^n a_{ni}\xi_i \xrightarrow{\mathcal{D}} N(0, 1).$$

THEOREM 1.2. Let $\{a_j, j \in Z\}$ be a sequence of nonnegative numbers such that $\sum_j a_j < \infty$ and let $\{\xi_j, j \in Z\}$ be a centered sequence of linearly positive quadrant dependent random variables which is uniformly integrable in L_2 and satisfies (1.6). Let

$$X_k = \sum_{j=-\infty}^{\infty} a_{k+j}\xi_j \quad \text{and} \quad S_n = \sum_{i=1}^n X_i.$$

Assume

$$(1.10) \quad \inf_{n \geq 1} n^{-1} \sigma_n^2 > 0,$$

where $\sigma_n^2 = \text{Var}(S_n)$. Then

$$(1.11) \quad \frac{S_n}{\sigma_n} \xrightarrow{\mathcal{D}} N(0, 1) \text{ as } n \rightarrow \infty.$$

This result extends Theorem 18.6.5 in Ibragimov and Linnik [5] from the i.i.d. case to the linearly positive quadrant dependence case by adding condition (1.6) and improves on the central limit theorem of Kim and Baek [6] for linear process generated by LPQD sequence.

2. Proofs

We start with the following lemma.

LEMMA 2.1. [8] *Let $\{Z_i, 1 \leq i \leq n\}$ be a sequence of linearly positive quadrant dependent random variables with finite second moments. Then*

$$\left| E \exp(it \sum_{j=1}^n Z_j) - \prod_{j=1}^n E \exp(it Z_j) \right| \leq Ct^2 \left| \text{Var}(\sum_{j=1}^n Z_j) - \sum_{j=1}^n \text{Var}(Z_j) \right|$$

for all $t \in \mathbb{R}$, where $C > 0$ is an arbitrary constant, not depending on n .

Proof of Theorem 1.1. Without loss of generality we assume that $a_{ni} = 0$ for all $i > n$. For every $1 \leq a < b \leq n$ and $1 \leq u \leq b - a$ we have, after simple manipulations,

$$(2.1) \quad \begin{aligned} 0 &\leq \sum_{i=a}^{b-u} a_{ni} \sum_{j=i+u}^b a_{nj} \text{Cov}(\xi_i, \xi_j) \\ &\leq \sup_k \left(\sum_{j:|k-j| \geq u} \text{Cov}(\xi_k, \xi_j) \right) \left(\sum_{i=a}^b a_{ni}^2 \right). \end{aligned}$$

In particular by (1.6), we have

$$(2.2) \quad \sup_k \left(\sum_{j:|k-j| \geq 1} \text{Cov}(\xi_k, \xi_j) \right) < \infty.$$

Without restricting the generality we can assume that $\sup_k E \xi_k^2 = 1$.

Hence, there exists a constant $C > 0$ such that for every $1 \leq a \leq b \leq n$,

$$\begin{aligned}
 \text{Var} \left(\sum_{i=a}^b a_{ni} \xi_i \right) &\leq \sum_{i=a}^b a_{ni}^2 \text{Var}(\xi_i) + 2 \sum_{i=a}^{b-1} a_{ni} \sum_{j=i+1}^b a_{nj} \text{Cov}(\xi_i, \xi_j) \\
 &\leq \sum_{i=a}^b a_{ni}^2 \text{Var}(\xi_i) \\
 (2.3) \qquad &+ 2 \sup_k \left(\sum_{j:|k-j| \geq 1}^{\infty} \text{Cov}(\xi_k, \xi_j) \right) \left(\sum_{i=a}^b a_{ni}^2 \right) \\
 &\leq C \left(\sum_{i=a}^b a_{ni}^2 \right)
 \end{aligned}$$

by (2.1) and (2.2).

We shall construct now a triangular array of random variables $\{Z_{ni}, 1 \leq i \leq n\}$ for which we shall make use of Lemma 2.1. Fix a small positive ϵ and find a positive integer $u = u_\epsilon$ such that, for every $n \geq u + 1$

$$(2.4) \qquad 0 \leq \left(\sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^n a_{nj} \text{Cov}(\xi_i, \xi_j) \right) \leq \epsilon.$$

This is possible because of (2.1) and (1.6). Denote by $[x]$ the integer part of x and define

$$\begin{aligned}
 K &= \left[\frac{1}{\epsilon} \right], \\
 Y_{nj} &= \sum_{i=u_j+1}^{u(j+1)} a_{ni} \xi_i, \quad j = 0, 1, \dots, \\
 A_j &= \left\{ i : 2Kj \leq i < 2Kj + K, \text{Cov}(Y_{ni}, Y_{n,i+1}) \right. \\
 &\qquad \left. \leq \frac{2}{K} \sum_{i=2Kj}^{2Kj+K} \text{Var}(Y_{ni}) \right\}.
 \end{aligned}$$

Since $2\text{Cov}(Y_{ni}, Y_{n,i+1}) \leq \text{Var}(Y_{ni}) + \text{Var}(Y_{n,i+1})$, we get that for every j the set A_j is not empty. Now we define the integers m_1, m_2, \dots, m_n recursively. Let $m_0 = 0$ and

$$m_{j+1} = \min\{m : m > m_j, m \in A_j\}$$

and define

$$Z_{nj} = \sum_{i=m_j+1}^{m_{j+1}} Y_{ni}, \quad j = 0, 1, \dots,$$

$$\Delta_j = \{u(m_j + 1) + 1, \dots, u(m_{j+1} + 1)\}.$$

We observe that

$$Z_{nj} = \sum_{k \in \Delta_j} a_{nk} \xi_k, \quad j = 0, 1, \dots$$

By the definition of LPQD the sequence $\{Z_{nj}\}$ is LPQD. From the fact that $m_j \geq 2K(j-1)$ and $m_{j+1} \leq K(2j+1)$ every set Δ_j contains no more than $3Ku$ elements and $m_{j+1}/m_j \rightarrow 1$ as $j \rightarrow \infty$. Hence, for every fixed positive ϵ by (1.2)–(1.5) the array $\{Z_{ni} : i = 0, 1, \dots, n; n \geq 1\}$ satisfies the Lindeberg condition (see Petrov [10], Theorem 22, p.100), that is, $\{Z_{nj}\}$ satisfies

$$(2.5) \quad \sigma_n^{-1} \sum_{j=1}^n E Z_{nj}^2 I(|Z_{nj}| > \epsilon \sigma_n) \rightarrow 0 \text{ as } n \rightarrow \infty,$$

where $\sigma_n^2 = \text{Var}(\sum_{j=1}^n Z_{nj})$. We can observe that by Lemma 2.1 and the construction,

$$(2.6) \quad \left| E \exp(it \sum_{j=1}^n Z_{nj}) - \prod_{j=1}^n E \exp(it Z_{nj}) \right|$$

$$\leq Ct^2 \left\{ \text{Var} \left(\sum_{j=1}^n Z_{nj} \right) - \sum_{j=1}^n \text{Var} (Z_{nj}) \right\}$$

$$\leq Ct^2 \left\{ 2 \left(\sum_{i=1}^n \text{Cov} (Z_{ni}, Z_{n,i+1}) \right) + 2 \left(\sum_{i=1}^{n-2} \sum_{j=i+2}^n \text{Cov} (Z_{ni}, Z_{nj}) \right) \right\}$$

$$\leq Ct^2 \left[\left\{ 2 \sum_{j=1}^n \text{Cov} (Y_{n,m_j}, Y_{n,m_j+1}) + 2 \sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^n a_{nj} \text{Cov} (\xi_i, \xi_j) \right\} \right.$$

$$\quad \left. + \left\{ 2 \sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^n a_{nj} \text{Cov} (\xi_i, \xi_j) \right\} \right]$$

$$= Ct^2 \left\{ 4 \sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^n a_{nj} \text{Cov} (\xi_i, \xi_j) + 2 \sum_{j=1}^n \text{Cov} (Y_{n,m_j}, Y_{n,m_j+1}) \right\}$$

$$\begin{aligned}
 &\leq Ct^2 \left\{ 4\epsilon + \frac{8}{K} \sum_{i=1}^n \text{Var}(Y_{ni}) \right\} \\
 &= Ct^2 \left\{ 4\epsilon + \frac{8}{K} \sum_{j=1}^n \text{Var} \left(\sum_{i=u_{j+1}}^{u^{(j+1)}} a_{ni} \xi_i \right) \right\} \\
 &\leq Ct^2 \left\{ 4\epsilon + \frac{8M}{K} \sum_{j=1}^n \sum_{i=u_{j+1}}^{u^{(j+1)}} a_{ni}^2 \right\} \text{ (by (2.3))} \\
 &\leq C_1 t^2 \epsilon \left\{ 1 + \sup_n \sum_{i=1}^n a_{ni}^2 \right\} \\
 &\leq C_2 t^2 \epsilon \text{ for every positive } \epsilon.
 \end{aligned}$$

Therefore the problem is now reduced to the study of the central limit theorem of a decoupled sequence $\{\tilde{Z}_{nj}\}$ of independent random variables such that for each n and j , the variable \tilde{Z}_{nj} is distributed as Z_{nj} .

By (2.5), $\{\tilde{Z}_{nj}\}$ also satisfies the Lindeberg condition, that is, $\{\tilde{Z}_{nj}\}$ satisfies $\tilde{\sigma}_n^{-1} \sum_{j=1}^n E \tilde{Z}_{nj}^2 I(|\tilde{Z}_{nj}| > \epsilon \tilde{\sigma}_n) \rightarrow 0$ as $n \rightarrow \infty$, where $\tilde{\sigma}_n^2 = \text{Var}(\sum_{j=1}^n \tilde{Z}_{nj})$, and hence by Theorem 7.2 of Billingsley [1]

$$(2.7) \quad \tilde{\sigma}_n^{-1} \sum_{j=1}^n \tilde{Z}_{nj} \xrightarrow{\mathcal{D}} N(0, 1) \text{ as } n \rightarrow \infty,$$

where $\tilde{\sigma}_n^2 = \text{Var}(\sum_{j=1}^n \tilde{Z}_{nj})$. It follows from (2.5), (2.6), and (2.7) that

$$(2.8) \quad \sigma_n^{-1} \sum_{j=1}^n Z_{nj} \xrightarrow{\mathcal{D}} N(0, 1) \text{ as } n \rightarrow \infty,$$

where $\sigma_n^2 = \text{Var}(\sum_{j=1}^n Z_{nj})$, and now the proof is complete by (2.7), (2.8), and Theorem 4.2 in Billingsley [1]. □

Proof of Corollary 1.1. Let $A_{ni} = \frac{a_{ni}}{\sigma_n}$. Then we have

$$\begin{aligned}
 &\max_{1 \leq i \leq n} A_{ni} \rightarrow 0 \text{ as } n \rightarrow \infty, \\
 &\sup_n \sum_{i=1}^n A_{ni}^2 < \infty, \\
 &\text{Var} \left(\sum_{i=1}^n A_{ni} \xi_i \right) = 1.
 \end{aligned}$$

Hence, by Theorem 1.1 the desired result (1.11) follows. □

Proof of Theorem 1.2. First note that $\sum_j a_j^2 < \infty$ and, without loss of generality, we can assume that $\sup E\xi_k^2 = 1$. Let

$$S_n = \sum_{k=1}^n X_k = \sum_{j=-\infty}^{\infty} \left(\sum_{k=1}^n a_{k+j} \right) \xi_j.$$

In order to apply Theorem 1.1, fix W_n such that $\sum_{|j|>W_n} a_j^2 < n^{-3}$ and take $k_n = W_n + n$. Then

$$\frac{S_n}{\sigma_n} = \sum_{|j| \leq k_n} \left(\sum_{k=1}^n a_{k+j} \right) \frac{\xi_j}{\sigma_n} + \sum_{|j| > k_n} \left(\sum_{k=1}^n a_{k+j} \right) \frac{\xi_j}{\sigma_n} = T_n + U_n.$$

By the Cauchy Schwarz inequality and the assumptions we have the following estimate

$$\begin{aligned} \text{Var}(U_n) &\leq C \sum_{|j| > k_n} \text{Var} \left(\sum_{k=1}^n a_{k+j} \frac{\xi_j}{\sigma_n} \right) \\ &\leq C \sum_{|j| > k_n} \left(\sum_{k=1}^n a_{k+j} / \sigma_n \right)^2 E\xi_j^2 \\ &\leq Cn\sigma_n^{-2} \sum_{|j| > k_n} \left(\sum_{k=1}^n a_{k+j}^2 \right) \\ &\leq Cn^2\sigma_n^{-2} \sum_{|j| > k_n - n} a_j^2 \\ &\leq Cn^2\sigma_n^{-2} \sum_{|j| > W_n} a_j^2 \\ &\leq Cn^{-1}\sigma_n^{-2} \rightarrow 0 \text{ as } n \rightarrow \infty \end{aligned}$$

which yields

$$(2.9) \quad U_n \rightarrow 0 \text{ in probability as } n \rightarrow \infty.$$

By Theorem 4.1 of Billingsley [1], it remains to prove that $T_n \xrightarrow{\mathcal{D}} N(0, 1)$. Put

$$(2.10) \quad a_{nk} = \frac{\sum_{j=1}^n a_{k+j}}{\sigma_n}.$$

From the assumption $\sum_j a_j < \infty$ ($a_j > 0$), (1.10), and (2.10) we obtain

$$\frac{\sup_{-\infty < k < \infty} \sum_{j=1}^n a_{k+j}}{\sigma_n} \rightarrow 0 \text{ as } n \rightarrow \infty,$$

$$\max_{1 \leq k \leq n} a_{nk} \rightarrow 0 \text{ as } n \rightarrow \infty,$$

$$\sup_n \sum_{k=1}^n a_{nk}^2 < \infty. \text{ (see Remark below.)}$$

Hence, by Theorem 1.1

$$(2.11) \quad T_n \xrightarrow{\mathcal{D}} N(0, 1)$$

and from (2.9) and (2.11) the desired result (1.10) follows. □

REMARK. In the proof of Theorem 1.2, let us suppose on the contrary that for some $\epsilon > 0$ there exists a subsequence $(j', n'), n' \rightarrow \infty$ such that $\sum_{k=1}^{n'} a_{k+j'} > \epsilon \sigma_{n'}$. Denote by $A = \sup_{-\infty < k < \infty} a_k$ and notice that for $r > j'$

$$\sum_{k=1}^{n'} a_{k+r} > \epsilon \sigma_{n'} - 2A(r - j').$$

Hence

$$\begin{aligned} \frac{\sigma_{n'}^2}{b} &\geq \sum_{i=j'}^{j'+w} \left(\sum_{k=1}^{n'} a_{k+j} \right)^2 \geq W \epsilon^2 \sigma_{n'}^2 - 4A \sigma_{n'} \epsilon \left(\sum_{i=j'}^{j'+W} (i - j') \right) \\ &\geq W \epsilon^2 \sigma_{n'}^2 - 4A \sigma_{n'} \epsilon W^2. \end{aligned}$$

Taking W to be the least integer greater than or equal to $\frac{3}{b\epsilon^2}$ and because $\sigma_{n'} \rightarrow \infty$, we obtain for n' sufficiently large,

$$\frac{\sigma_{n'}^2}{b} \geq \frac{3\sigma_{n'}^2}{b} - \sigma_{n'} \frac{36A}{b^2\epsilon^3} \geq \frac{2\sigma_{n'}^2}{b}$$

which is a contradiction. □

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